

# Package ‘TidyDensity’

May 28, 2024

**Title** Functions for Tidy Analysis and Generation of Random Data

**Version** 1.5.0

**Description** To make it easy to generate random numbers based upon the underlying stats distribution functions. All data is returned in a tidy and structured format making working with the data simple and straight forward. Given that the data is returned in a tidy 'tibble' it lends itself to working with the rest of the 'tidyverse'.

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**Encoding** UTF-8

**RoxygenNote** 7.3.1

**URL** <https://github.com/spsanderson/TidyDensity>

**BugReports** <https://github.com/spsanderson/TidyDensity/issues>

**Depends** R (>= 4.1.0)

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**Suggests** rmarkdown, knitr, EnvStats

**VignetteBuilder** knitr

**NeedsCompilation** no

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bootstrap\_density\_augment

*Bootstrap Density Tibble*

---

## Description

Add density information to the output of `tidy_bootstrap()`, and `bootstrap_unnest_tbl()`.

## Usage

```
bootstrap_density_augment(.data)
```

## Arguments

`.data` The data that is passed from the `tidy_bootstrap()` or `bootstrap_unnest_tbl()` functions.

**Details**

This function takes as input the output of the `tidy_bootstrap()` or `bootstrap_unnest_tbl()` and returns an augmented tibble that has the following columns added to it: `x`, `y`, `dx`, and `dy`.

It looks for an attribute that comes from using `tidy_bootstrap()` or `bootstrap_unnest_tbl()` so it will not work unless the data comes from one of those functions.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Bootstrap: [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_augment\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [bootstrap\\_stat\\_plot\(\)](#), [bootstrap\\_unnest\\_tbl\(\)](#), [tidy\\_bootstrap\(\)](#)

Other Augment Function: [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_q\\_augment\(\)](#)

**Examples**

```
x <- mtcars$mpg

tidy_bootstrap(x) |>
  bootstrap_density_augment()

tidy_bootstrap(x) |>
  bootstrap_unnest_tbl() |>
  bootstrap_density_augment()
```

---

`bootstrap_p_augment`    *Augment Bootstrap P*

---

**Description**

Takes a numeric vector and will return the ecdf probability.

**Usage**

```
bootstrap_p_augment(.data, .value, .names = "auto")
```

**Arguments**

<code>.data</code>	The data being passed that will be augmented by the function.
<code>.value</code>	This is passed <a href="#">rlang::enquo()</a> to capture the vectors you want to augment.
<code>.names</code>	The default is "auto"

**Details**

Takes a numeric vector and will return the ecdf probability of that vector. This function is intended to be used on its own in order to add columns to a tibble.

**Value**

A augmented tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Augment Function: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_q\\_augment\(\)](#)

Other Bootstrap: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_augment\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [bootstrap\\_stat\\_plot\(\)](#), [bootstrap\\_unnest\\_tbl\(\)](#), [tidy\\_bootstrap\(\)](#)

**Examples**

```
x <- mtcars$mpg
tidy_bootstrap(x) |>
  bootstrap_unnest_tbl() |>
  bootstrap_p_augment(y)
```

---

bootstrap\_p\_vec

*Compute Bootstrap P of a Vector*

---

**Description**

This function takes in a vector as it's input and will return the ecdf probability of a vector.

**Usage**

```
bootstrap_p_vec(.x)
```

**Arguments**

.x                    A numeric

**Details**

A function to return the ecdf probability of a vector.

**Value**

A vector

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Bootstrap: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_q\\_augment\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [bootstrap\\_stat\\_plot\(\)](#), [bootstrap\\_unnest\\_tbl\(\)](#), [tidy\\_bootstrap\(\)](#)

Other Vector Function: [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

**Examples**

```
x <- mtcars$mpg
bootstrap_p_vec(x)
```

---

bootstrap\_q\_augment     *Augment Bootstrap Q*

---

**Description**

Takes a numeric vector and will return the quantile.

**Usage**

```
bootstrap_q_augment(.data, .value, .names = "auto")
```

**Arguments**

<code>.data</code>	The data being passed that will be augmented by the function.
<code>.value</code>	This is passed <a href="#">rlang::enquo()</a> to capture the vectors you want to augment.
<code>.names</code>	The default is "auto"

**Details**

Takes a numeric vector and will return the quantile of that vector. This function is intended to be used on its own in order to add columns to a tibble.

**Value**

A augmented tibble

**Author(s)**

Steven P. Sanderson II, MPH



**See Also**

Other Augment Function: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_augment\(\)](#)

Other Bootstrap: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [bootstrap\\_stat\\_plot\(\)](#), [bootstrap\\_unnest\\_tbl\(\)](#), [tidy\\_bootstrap\(\)](#)

**Examples**

```
x <- mtcars$mpg

tidy_bootstrap(x) |>
  bootstrap_unnest_tbl() |>
  bootstrap_q_augment(y)
```

---

bootstrap\_q\_vec

*Compute Bootstrap Q of a Vector*

---

**Description**

This function takes in a vector as it's input and will return the quantile of a vector.

**Usage**

```
bootstrap_q_vec(.x)
```

**Arguments**

.x                    A numeric

**Details**

A function to return the quantile of a vector.

**Value**

A vector

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Bootstrap: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_augment\(\)](#), [bootstrap\\_stat\\_plot\(\)](#), [bootstrap\\_unnest\\_tbl\(\)](#), [tidy\\_bootstrap\(\)](#)

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

## Examples

```
x <- mtcars$mpg  
  
bootstrap_q_vec(x)
```

---

bootstrap\_stat\_plot *Bootstrap Stat Plot*

---

## Description

This function produces a plot of a cumulative statistic function applied to the bootstrap variable from `tidy_bootstrap()` or after `bootstrap_unnest_tbl()` has been applied to it.

## Usage

```
bootstrap_stat_plot(  
  .data,  
  .value,  
  .stat = "cmean",  
  .show_groups = FALSE,  
  .show_ci_labels = TRUE,  
  .interactive = FALSE  
)
```

## Arguments

<code>.data</code>	The data that comes from either <code>tidy_bootstrap()</code> or after <code>bootstrap_unnest_tbl()</code> is applied to it.
<code>.value</code>	The value column that the calculations are being applied to.
<code>.stat</code>	The cumulative statistic function being applied to the <code>.value</code> column. It must be quoted. The default is "cmean".
<code>.show_groups</code>	The default is FALSE, set to TRUE to get output of all simulations of the bootstrap data.
<code>.show_ci_labels</code>	The default is TRUE, this will show the last value of the upper and lower quantile.
<code>.interactive</code>	The default is FALSE, set to TRUE to get a plotly plot object back.

## Details

This function will take in data from either `tidy_bootstrap()` directly or after apply `bootstrap_unnest_tbl()` to its output. There are several different cumulative functions that can be applied to the data. The accepted values are:

- "cmean" - Cumulative Mean

- "chmean" - Cumulative Harmonic Mean
- "cgmean" - Cumulative Geometric Mean
- "csum" = Cumulative Sum
- "cmedian" = Cumulative Median
- "cmax" = Cumulative Max
- "cmin" = Cumulative Min
- "cprod" = Cumulative Product
- "csd" = Cumulative Standard Deviation
- "cvar" = Cumulative Variance
- "cskewness" = Cumulative Skewness
- "ckurtosis" = Cumulative Kurtosis

**Value**

A plot either `ggplot2` or `plotly`.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Bootstrap: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_augment\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [bootstrap\\_unnest\\_tbl\(\)](#), [tidy\\_bootstrap\(\)](#)

Other Autoplot: [tidyautoplot\(\)](#), [tidy\\_combinedautoplot\(\)](#), [tidy\\_fourautoplot\(\)](#), [tidy\\_multi\\_distautoplot\(\)](#), [tidy\\_random\\_walkautoplot\(\)](#)

**Examples**

```
x <- mtcars$mpg

tidy_bootstrap(x) |>
  bootstrap_stat_plot(y, "cmean")

tidy_bootstrap(x, .num_sims = 10) |>
  bootstrap_stat_plot(y,
    .stat = "chmean", .show_groups = TRUE,
    .show_ci_label = FALSE
  )
```

---

bootstrap\_unnest\_tbl *Unnest Tidy Bootstrap Tibble*

---

### Description

Unnest the data output from `tidy_bootstrap()`.

### Usage

```
bootstrap_unnest_tbl(.data)
```

### Arguments

`.data`            The data that is passed from the `tidy_bootstrap()` function.

### Details

This function takes as input the output of the `tidy_bootstrap()` function and returns a two column tibble. The columns are `sim_number` and `y`

It looks for an attribute that comes from using `tidy_bootstrap()` so it will not work unless the data comes from that function.

### Value

A tibble

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Bootstrap: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_augment\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [bootstrap\\_stat\\_plot\(\)](#), [tidy\\_bootstrap\(\)](#)

### Examples

```
tb <- tidy_bootstrap(.x = mtcars$mpg)
bootstrap_unnest_tbl(tb)

bootstrap_unnest_tbl(tb) |>
  tidy_distribution_summary_tbl(sim_number)
```

---

cgmean	<i>Cumulative Geometric Mean</i>
--------	----------------------------------

---

## Description

A function to return the cumulative geometric mean of a vector.

## Usage

```
cgmean(.x)
```

## Arguments

`.x` A numeric vector

## Details

A function to return the cumulative geometric mean of a vector.  $\exp(\text{cummean}(\log(.x)))$

## Value

A numeric vector

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

## Examples

```
x <- mtcars$mpg  
  
cgmean(x)
```

---

check\_duplicate\_rows *Check for Duplicate Rows in a Data Frame*

---

### Description

This function checks for duplicate rows in a data frame.

### Usage

```
check_duplicate_rows(.data)
```

### Arguments

`.data` A data frame.

### Details

This function checks for duplicate rows by comparing each row in the data frame to every other row. If a row is identical to another row, it is considered a duplicate.

### Value

A logical vector indicating whether each row is a duplicate or not.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

[duplicated](#), [anyDuplicated](#)

Other Utility: [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

### Examples

```
data <- data.frame(  
  x = c(1, 2, 3, 1),  
  y = c(2, 3, 4, 2),  
  z = c(3, 2, 5, 3)  
)
```

```
check_duplicate_rows(data)
```

---

chmean

*Cumulative Harmonic Mean*

---

### Description

A function to return the cumulative harmonic mean of a vector.

### Usage

```
chmean(.x)
```

### Arguments

`.x` A numeric vector

### Details

A function to return the cumulative harmonic mean of a vector.  $1 / (\text{cumsum}(1 / .x))$

### Value

A numeric vector

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

### Examples

```
x <- mtcars$mpg  
  
chmean(x)
```

---

ci_hi	<i>Confidence Interval Generic</i>
-------	------------------------------------

---

**Description**

Gets the upper 97.5% quantile of a numeric vector.

**Usage**

```
ci_hi(.x, .na_rm = FALSE)
```

**Arguments**

.x	A vector of numeric values
.na_rm	A Boolean, defaults to FALSE. Passed to the quantile function.

**Details**

Gets the upper 97.5% quantile of a numeric vector.

**Value**

A numeric value.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Statistic: [ci\\_lo\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_range\\_statistic\(\)](#), [tidy\\_skewness\\_vec\(\)](#), [tidy\\_stat\\_tbl\(\)](#)

**Examples**

```
x <- mtcars$mpg
ci_hi(x)
```



---

ci_lo	<i>Confidence Interval Generic</i>
-------	------------------------------------

---

**Description**

Gets the lower 2.5% quantile of a numeric vector.

**Usage**

```
ci_lo(.x, .na_rm = FALSE)
```

**Arguments**

`.x` A vector of numeric values  
`.na_rm` A Boolean, defaults to FALSE. Passed to the quantile function.

**Details**

Gets the lower 2.5% quantile of a numeric vector.

**Value**

A numeric value.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Statistic: [ci\\_hi\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_range\\_statistic\(\)](#), [tidy\\_skewness\\_vec\(\)](#), [tidy\\_stat\\_tbl\(\)](#)

**Examples**

```
x <- mtcars$mpg  
ci_lo(x)
```

---

`ckurtosis`*Cumulative Kurtosis*

---

**Description**

A function to return the cumulative kurtosis of a vector.

**Usage**

```
ckurtosis(.x)
```

**Arguments**

`.x` A numeric vector

**Details**

A function to return the cumulative kurtosis of a vector.

**Value**

A numeric vector

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

**Examples**

```
x <- mtcars$mpg  
  
ckurtosis(x)
```

---

cmean	<i>Cumulative Mean</i>
-------	------------------------

---

### Description

A function to return the cumulative mean of a vector.

### Usage

```
cmean(.x)
```

### Arguments

`.x` A numeric vector

### Details

A function to return the cumulative mean of a vector. It uses `dplyr::cummean()` as the basis of the function.

### Value

A numeric vector

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

### Examples

```
x <- mtcars$mpg  
  
cmean(x)
```

---

`cmedian`*Cumulative Median*

---

**Description**

A function to return the cumulative median of a vector.

**Usage**

```
cmedian(.x)
```

**Arguments**

`.x` A numeric vector

**Details**

A function to return the cumulative median of a vector.

**Value**

A numeric vector

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

**Examples**

```
x <- mtcars$mpg
```

```
cmedian(x)
```

---

color_blind	<i>Provide Colorblind Compliant Colors</i>
-------------	--

---

**Description**

8 Hex RGB color definitions suitable for charts for colorblind people.

**Usage**

```
color_blind()
```

---

convert_to_ts	<i>Convert Data to Time Series Format</i>
---------------	---

---

**Description**

This function converts data in a data frame or tibble into a time series format. It is designed to work with data generated from `tidy_` distribution functions. The function can return time series data, pivot it into long format, or both.

**Usage**

```
convert_to_ts(.data, .return_ts = TRUE, .pivot_longer = FALSE)
```

**Arguments**

<code>.data</code>	A data frame or tibble to be converted into a time series format.
<code>.return_ts</code>	A logical value indicating whether to return the time series data. Default is TRUE.
<code>.pivot_longer</code>	A logical value indicating whether to pivot the data into long format. Default is FALSE.

**Details**

The function takes a data frame or tibble as input and processes it based on the specified options. It performs the following actions:

1. Checks if the input is a data frame or tibble; otherwise, it raises an error.
2. Checks if the data comes from a `tidy_` distribution function; otherwise, it raises an error.
3. Converts the data into a time series format, grouping it by "sim\_number" and transforming the "y" column into a time series.
4. Returns the result based on the chosen options:
  - If `ret_ts` is set to TRUE, it returns the time series data.
  - If `pivot_longer` is set to TRUE, it pivots the data into long format.
  - If both options are set to FALSE, it returns the data as a tibble.

**Value**

The function returns the processed data based on the chosen options:

- If `ret_ts` is set to `TRUE`, it returns time series data.
- If `pivot_longer` is set to `TRUE`, it returns the data in long format.
- If both options are set to `FALSE`, it returns the data as a tibble.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

**Examples**

```
# Example 1: Convert data to time series format without returning time series data
x <- tidy_normal()
result <- convert_to_ts(x, FALSE)
head(result)

# Example 2: Convert data to time series format and pivot it into long format
x <- tidy_normal()
result <- convert_to_ts(x, FALSE, TRUE)
head(result)

# Example 3: Convert data to time series format and return the time series data
x <- tidy_normal()
result <- convert_to_ts(x)
head(result)
```

**Description**

A function to return the cumulative standard deviation of a vector.

**Usage**

```
csd(.x)
```

**Arguments**

.x                    A numeric vector

**Details**

A function to return the cumulative standard deviation of a vector.

**Value**

A numeric vector. Note: The first entry will always be NaN.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

**Examples**

```
x <- mtcars$mpg  
  
csd(x)
```

---

cskewness

*Cumulative Skewness*

---

**Description**

A function to return the cumulative skewness of a vector.

**Usage**

```
cskewness(.x)
```

**Arguments**

.x                    A numeric vector

**Details**

A function to return the cumulative skewness of a vector.

**Value**

A numeric vector

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

**Examples**

```
x <- mtcars$mpg
cskewness(x)
```

---

cvar

*Cumulative Variance*

---

**Description**

A function to return the cumulative variance of a vector.

**Usage**

```
cvar(.x)
```

**Arguments**

`.x` A numeric vector

**Details**

A function to return the cumulative variance of a vector. `exp(cummean(log(.x)))`

**Value**

A numeric vector. Note: The first entry will always be NaN.

**Author(s)**

Steven P. Sanderson II, MPH



**See Also**

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

**Examples**

```
x <- mtcars$mpg
cvar(x)
```

---

dist\_type\_extractor     *Extract Distribution Type from Tidy Distribution Object*

---

**Description**

Get the distribution name in title case from the tidy\_ distribution function.

**Usage**

```
dist_type_extractor(.x)
```

**Arguments**

.x                    The attribute list passed from a tidy\_ distribution function.

**Details**

This will extract the distribution type from a tidy\_ distribution function output using the attributes of that object. You must pass the attribute directly to the function. It is meant really to be used internally.

You should be passing if using manually the \$tibble\_type attribute.

**Value**

A character string

**Author(s)**

Steven P. Sanderson II,

**Examples**

```
tn <- tidy_normal()
atb <- attributes(tn)
dist_type_extractor(atb$tibble_type)
```

---

quantile\_normalize      *Perform quantile normalization on a numeric matrix/data.frame*

---

### Description

This function will perform quantile normalization on two or more distributions of equal length. Quantile normalization is a technique used to make the distribution of values across different samples more similar. It ensures that the distributions of values for each sample have the same quantiles. This function takes a numeric matrix as input and returns a quantile-normalized matrix.

### Usage

```
quantile_normalize(.data, .return_tibble = FALSE)
```

### Arguments

`.data`                    A numeric matrix where each column represents a sample.  
`.return_tibble`        A logical value that determines if the output should be a tibble. Default is 'FALSE'.

### Details

This function performs quantile normalization on a numeric matrix by following these steps:

1. Sort each column of the input matrix.
2. Calculate the mean of each row across the sorted columns.
3. Replace each column's sorted values with the row means.
4. Unsort the columns to their original order.

### Value

A list object that has the following:

1. A numeric matrix that has been quantile normalized.
2. The row means of the quantile normalized matrix.
3. The sorted data
4. The ranked indices

### Author(s)

Steven P. Sanderson II, MPH

**See Also**

`rowMeans`: Calculate row means.

`apply`: Apply a function over the margins of an array.

`order`: Order the elements of a vector.

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

**Examples**

```
# Create a sample numeric matrix
data <- matrix(rnorm(20), ncol = 4)

# Perform quantile normalization
normalized_data <- quantile_normalize(data)
normalized_data

as.data.frame(normalized_data$normalized_data) |>
  sapply(function(x) quantile(x, probs = seq(0, 1, 1 / 4)))

quantile_normalize(
  data.frame(rnorm(30),
            rnorm(30)),
  .return_tibble = TRUE)
```

---



*Provide Colorblind Compliant Colors*

---

**Description**

Provide Colorblind Compliant Colors

**Usage**

```
td_scale_color_colorblind(..., theme = "td")
```

**Arguments**

...	Data passed to the function
theme	This defaults to <code>td</code> and that is the only allowed value

---



*Provide Colorblind Compliant Colors*


---

### Description

Provide Colorblind Compliant Colors

### Usage

```
td_scale_fill_colorblind(..., theme = "td")
```

### Arguments

...	Data passed to the function
theme	This defaults to td and that is the only allowed value

---

tidy\_autoplot

*Automatic Plot of Density Data*


---

### Description

This is an auto plotting function that will take in a tidy\_ distribution function and a few arguments, one being the plot type, which is a quoted string of one of the following:

- density
- quantile
- probablity
- qq
- mcmc

If the number of simulations exceeds 9 then the legend will not print. The plot subtitle is put together by the attributes of the table passed to the function.

### Usage

```
tidy_autoplot(
  .data,
  .plot_type = "density",
  .line_size = 0.5,
  .geom_point = FALSE,
  .point_size = 1,
  .geom_rug = FALSE,
  .geom_smooth = FALSE,
  .geom_jitter = FALSE,
  .interactive = FALSE
)
```

**Arguments**

<code>.data</code>	The data passed in from a tidy_distribution function like tidy_normal()
<code>.plot_type</code>	This is a quoted string like 'density'
<code>.line_size</code>	The size param ggplot
<code>.geom_point</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return a plot with ggplot2::geom_point()
<code>.point_size</code>	The point size param for ggplot
<code>.geom_rug</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of ggplot2::geom_rug()
<code>.geom_smooth</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of ggplot2::geom_smooth() The aes parameter of group is set to FALSE. This ensures a single smoothing band returned with SE also set to FALSE. Color is set to 'black' and linetype is 'dashed'.
<code>.geom_jitter</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of ggplot2::geom_jitter()
<code>.interactive</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return an interactive plotly plot.

**Details**

This function will spit out one of the following plots:

- density
- quantile
- probability
- qq
- mcmc

**Value**

A ggplot or a plotly plot.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Autoplot: [bootstrap\\_stat\\_plot\(\)](#), [tidy\\_combinedautoplot\(\)](#), [tidy\\_fourautoplot\(\)](#), [tidy\\_multi\\_distautoplot\(\)](#), [tidy\\_random\\_walkautoplot\(\)](#)

**Examples**

```
tidy_normal(.num_sims = 5) |>
  tidyautoplot()

tidy_normal(.num_sims = 20) |>
  tidyautoplot(.plot_type = "qq")
```

---

`tidy_bernoulli`*Tidy Randomly Generated Bernoulli Distribution Tibble*

---

### Description

This function will generate `n` random points from a Bernoulli distribution with a user provided, `.prob`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

### Usage

```
tidy_bernoulli(.n = 50, .prob = 0.1, .num_sims = 1, .return_tibble = TRUE)
```

### Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.prob</code>	The probability of success/failure.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

### Details

This function uses the `rbinom()`, and its underlying `p`, `d`, and `q` functions. The *Bernoulli* distribution is a special case of the *Binomial* distribution with `size = 1` hence this is why the `binom` functions are used and set to `size = 1`.

### Value

A tibble of randomly generated data.

### Author(s)

Steven P. Sanderson II, MPH

**See Also**

[https://en.wikipedia.org/wiki/Bernoulli\\_distribution](https://en.wikipedia.org/wiki/Bernoulli_distribution)

Other Discrete Distribution: `tidy_binomial()`, `tidy_hypergeometric()`, `tidy_negative_binomial()`, `tidy_poisson()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `tidy_zero_truncated_poisson()`

Other Bernoulli: `util_bernoulli_param_estimate()`, `util_bernoulli_stats_tbl()`

**Examples**

```
tidy_bernoulli()
```

---

tidy\_beta

*Tidy Randomly Generated Beta Distribution Tibble*

---

**Description**

This function will generate  $n$  random points from a beta distribution with a user provided, `.shape1`, `.shape2`, `.ncp` or non-centrality parameter, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_beta(
  .n = 50,
  .shape1 = 1,
  .shape2 = 1,
  .ncp = 0,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

### Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.shape1</code>	A non-negative parameter of the Beta distribution.
<code>.shape2</code>	A non-negative parameter of the Beta distribution.
<code>.ncp</code>	The non-centrality parameter of the Beta distribution.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

### Details

This function uses the underlying `stats::rbeta()`, and its underlying `p`, `d`, and `q` functions. For more information please see [stats::rbeta\(\)](#)

### Value

A tibble of randomly generated data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

<https://statisticsglobe.com/beta-distribution-in-r-dbeta-pbeta-qbeta-rbeta>

[https://en.wikipedia.org/wiki/Beta\\_distribution](https://en.wikipedia.org/wiki/Beta_distribution)

Other Continuous Distribution: [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Beta: [tidy\\_generalized\\_beta\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#)

### Examples

```
tidy_beta()
```



---

`tidy_binomial`*Tidy Randomly Generated Binomial Distribution Tibble*

---

## Description

This function will generate  $n$  random points from a binomial distribution with a user provided, `.size`, `.prob`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

## Usage

```
tidy_binomial(  
  .n = 50,  
  .size = 0,  
  .prob = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

## Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.size</code>	Number of trials, zero or more.
<code>.prob</code>	Probability of success on each trial.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

## Details

This function uses the underlying `stats::rbinom()`, and its underlying `p`, `d`, and `q` functions. For more information please see `stats::rbinom()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda366i.htm>

Other Discrete Distribution: `tidy_bernoulli()`, `tidy_hypergeometric()`, `tidy_negative_binomial()`, `tidy_poisson()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `tidy_zero_truncated_poisson()`

Other Binomial: `tidy_negative_binomial()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `util_binomial_param_estimate()`, `util_binomial_stats_tbl()`, `util_negative_binomial_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_negative_binomial_stats_tbl()`

**Examples**

```
tidy_binomial()
```

---

tidy\_bootstrap

*Bootstrap Empirical Data*

---

**Description**

Takes an input vector of numeric data and produces a bootstrapped nested tibble by simulation number.

**Usage**

```
tidy_bootstrap(
  .x,
  .num_sims = 2000,
  .proportion = 0.8,
  .distribution_type = "continuous"
)
```

**Arguments**

`.x` The vector of data being passed to the function. Must be a numeric vector.

`.num_sims` The default is 2000, can be set to anything desired. A warning will pass to the console if the value is less than 2000.

`.proportion` How much of the original data do you want to pass through to the sampling function. The default is 0.80 (80%)

`.distribution_type` This can either be 'continuous' or 'discrete'

## Details

This function will take in a numeric input vector and produce a tibble of bootstrapped values in a list. The table that is output will have two columns: `sim_number` and `bootstrap_samples`

The `sim_number` corresponds to how many times you want the data to be resampled, and the `bootstrap_samples` column contains a list of the bootstrapped resampled data.

## Value

A nested tibble

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Bootstrap: [bootstrap\\_density\\_augment\(\)](#), [bootstrap\\_p\\_augment\(\)](#), [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_augment\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [bootstrap\\_stat\\_plot\(\)](#), [bootstrap\\_unnest\\_tbl\(\)](#)

## Examples

```
x <- mtcars$mpg
tidy_bootstrap(x)
```

---

tidy\_burr

*Tidy Randomly Generated Burr Distribution Tibble*

---

## Description

This function will generate `n` random points from a Burr distribution with a user provided, `.shape1`, `.shape2`, `.scale`, `.rate`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_burr(
  .n = 50,
  .shape1 = 1,
  .shape2 = 1,
  .rate = 1,
  .scale = 1/.rate,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape1</code>	Must be strictly positive.
<code>.shape2</code>	Must be strictly positive.
<code>.rate</code>	An alternative way to specify the <code>.scale</code> .
<code>.scale</code>	Must be strictly positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rburr()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rburr\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Burr: [tidy\\_inverse\\_burr\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#)

## Examples

```
tidy_burr()
```

---

tidy_cauchy	<i>Tidy Randomly Generated Cauchy Distribution Tibble</i>
-------------	---

---

## Description

This function will generate  $n$  random points from a cauchy distribution with a user provided, `.location`, `.scale`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

## Usage

```
tidy_cauchy(  
  .n = 50,  
  .location = 0,  
  .scale = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

## Arguments

- |                             |   |
|-----------------------------|---|
| <code>.n</code>             | The number of randomly generated points you want.                                     |
| <code>.location</code>      | The location parameter.   |
| <code>.scale</code>         | The scale parameter, must be greater than or equal to 0.                              |
| <code>.num_sims</code>      | The number of randomly generated simulations you want.                                |
| <code>.return_tibble</code> | A logical value indicating whether to return the result as a tibble. Default is TRUE. |

**Details**

This function uses the underlying `stats::rcauchy()`, and its underlying `p`, `d`, and `q` functions. For more information please see `stats::rcauchy()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3663.htm>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Cauchy: `util_cauchy_param_estimate()`, `util_cauchy_stats_tbl()`

**Examples**

```
tidy_cauchy()
```

---

tidy_chisquare	<i>Tidy Randomly Generated Chisquare (Non-Central) Distribution Tibble</i>
----------------	--

---

**Description**

This function will generate `n` random points from a chisquare distribution with a user provided, `.df`, `.ncp`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_chisquare(  
  .n = 50,  
  .df = 1,  
  .ncp = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.df</code>	Degrees of freedom (non-negative but can be non-integer)
<code>.ncp</code>	Non-centrality parameter, must be non-negative.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rchisq()`, and its underlying `p`, `d`, and `q` functions. For more information please see [stats::rchisq\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3666.htm>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Chisquare: [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#)

**Examples**

```
tidy_chisquare()
```

---

tidy\_combined\_autoplot

*Automatic Plot of Combined Multi Dist Data*


---

## Description

This is an auto plotting function that will take in a `tidy_` distribution function and a few arguments, one being the plot type, which is a quoted string of one of the following:

- density
- quantile
- probablity
- qq
- mcmc

If the number of simulations exceeds 9 then the legend will not print. The plot subtitle is put together by the attributes of the table passed to the function.

## Usage

```
tidy_combined_autoplot(
  .data,
  .plot_type = "density",
  .line_size = 0.5,
  .geom_point = FALSE,
  .point_size = 1,
  .geom_rug = FALSE,
  .geom_smooth = FALSE,
  .geom_jitter = FALSE,
  .interactive = FALSE
)
```

## Arguments

<code>.data</code>	The data passed in from a the function <code>tidy_multi_dist()</code>
<code>.plot_type</code>	This is a quoted string like 'density'
<code>.line_size</code>	The size param <code>ggplot</code>
<code>.geom_point</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return a plot with <code>ggplot2::geom_point()</code>
<code>.point_size</code>	The point size param for <code>ggplot</code>
<code>.geom_rug</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_rug()</code>
<code>.geom_smooth</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_smooth()</code> The <code>aes</code> parameter of <code>group</code> is set to FALSE. This ensures a single smoothing band returned with <code>SE</code> also set to FALSE. Color is set to 'black' and <code>linetype</code> is 'dashed'.



- `.geom_jitter` A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of `ggplot2::geom_jitter()`
- `.interactive` A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return an interactive plotly plot.

### Details

This function will spit out one of the following plots:

- density
- quantile
- probability
- qq
- mcmc

### Value

A ggplot or a plotly plot.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Autoplot: [bootstrap\\_stat\\_plot\(\)](#), [tidyautoplot\(\)](#), [tidy\\_fourautoplot\(\)](#), [tidy\\_multi\\_distautoplot\(\)](#), [tidy\\_random\\_walkautoplot\(\)](#)

### Examples

```
combined_tbl <- tidy_combine_distributions(  
  tidy_normal(),  
  tidy_gamma(),  
  tidy_beta()  
)  
  
combined_tbl  
  
combined_tbl |>  
  tidy_combinedautoplot()  
  
combined_tbl |>  
  tidy_combinedautoplot(.plot_type = "qq")
```

---

`tidy_combine_distributions`*Combine Multiple Tidy Distributions of Different Types*

---

**Description**

This allows a user to specify any n number of tidy\_ distributions that can be combined into a single tibble. This is the preferred method for combining multiple distributions of different types, for example a Gaussian distribution and a Beta distribution.

This generates a single tibble with an added column of dist\_type that will give the distribution family name and its associated parameters.

**Usage**

```
tidy_combine_distributions(...)
```

**Arguments**

...                   The ... is where you can place your different distributions

**Details**

Allows a user to generate a tibble of different tidy\_ distributions

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Multiple Distribution: [tidy\\_multi\\_single\\_dist\(\)](#)

**Examples**

```
tn <- tidy_normal()
tb <- tidy_beta()
tc <- tidy_cauchy()

tidy_combine_distributions(tn, tb, tc)

## OR

tidy_combine_distributions(
```

```
tidy_normal(),
tidy_beta(),
tidy_cauchy(),
tidy_logistic()
)
```

---

tidy\_distribution\_comparison

*Compare Empirical Data to Distributions*

---

### Description

Compare some empirical data set against different distributions to help find the distribution that could be the best fit.

### Usage

```
tidy_distribution_comparison(
  .x,
  .distribution_type = "continuous",
  .round_to_place = 3
)
```

### Arguments

<code>.x</code>	The data set being passed to the function
<code>.distribution_type</code>	What kind of data is it, can be one of continuous or discrete
<code>.round_to_place</code>	How many decimal places should the parameter estimates be rounded off to for distribution construction. The default is 3

### Details

The purpose of this function is to take some data set provided and to try to find a distribution that may fit the best. A parameter of `.distribution_type` must be set to either continuous or discrete in order for this the function to try the appropriate types of distributions.

The following distributions are used:

Continuous:

- tidy\_beta
- tidy\_cauchy
- tidy\_chisquare
- tidy\_exponential
- tidy\_gamma

- tidy\_logistic
- tidy\_lognormal
- tidy\_normal
- tidy\_pareto
- tidy\_uniform
- tidy\_weibull

Discrete:

- tidy\_binomial
- tidy\_geometric
- tidy\_hypergeometric
- tidy\_poisson

The function itself returns a list output of tibbles. Here are the tibbles that are returned:

- comparison\_tbl
- deviance\_tbl
- total\_deviance\_tbl
- aic\_tbl
- kolmogorov\_smirnov\_tbl
- multi\_metric\_tbl

The `comparison_tbl` is a long tibble that lists the values of the density function against the given data.

The `deviance_tbl` and the `total_deviance_tbl` just give the simple difference from the actual density to the estimated density for the given estimated distribution.

The `aic_tbl` will provide the AIC for likelihood of the distribution.

The `kolmogorov_smirnov_tbl` for now provides a `two.sided` estimate of the `ks.test` of the estimated density against the empirical.

The `multi_metric_tbl` will summarise all of these metrics into a single tibble.

## Value

An invisible list object. A tibble is printed.

## Author(s)

Steven P. Sanderson II, MPH

## Examples

```
xc <- mtcars$mpg
output_c <- tidy_distribution_comparison(xc, "continuous")

xd <- trunc(xc)
output_d <- tidy_distribution_comparison(xd, "discrete")

output_c
output_d
```

---

tidy\_distribution\_summary\_tbl

*Tidy Distribution Summary Statistics Tibble*

---

## Description

This function returns a summary statistics tibble. It will use the y column from the tidy\_ distribution function.

## Usage

```
tidy_distribution_summary_tbl(.data, ...)
```

## Arguments

.data            The data that is going to be passed from a a tidy\_ distribution function.  
...             This is the grouping variable that gets passed to `dplyr::group_by()` and `dplyr::select()`.

## Details

This function takes in a tidy\_ distribution table and will return a tibble of the following information:

- sim\_number
- mean\_val
- median\_val
- std\_val
- min\_val
- max\_val
- skewness
- kurtosis
- range
- iqr
- variance
- ci\_hi
- ci\_lo

The kurtosis and skewness come from the package `healthyR.ai`

**Value**

A summary stats tibble

**Author(s)**

Steven P. Sanderson II, MPH

**Examples**

```
library(dplyr)

tn <- tidy_normal(.num_sims = 5)
tb <- tidy_beta(.num_sims = 5)

tidy_distribution_summary_tbl(tn)
tidy_distribution_summary_tbl(tn, sim_number)

data_tbl <- tidy_combine_distributions(tn, tb)

tidy_distribution_summary_tbl(data_tbl)
tidy_distribution_summary_tbl(data_tbl, dist_type)
```

---

tidy_empirical	<i>Tidy Empirical</i>
----------------	-----------------------

---

**Description**

This function takes in a single argument of `.x` a vector and will return a tibble of information similar to the `tidy_distribution` functions. The `y` column is set equal to `dy` from the density function.

**Usage**

```
tidy_empirical(.x, .num_sims = 1, .distribution_type = "continuous")
```

**Arguments**

<code>.x</code>	A vector of numbers
<code>.num_sims</code>	How many simulations should be run, defaults to 1.
<code>.distribution_type</code>	A string of either "continuous" or "discrete". The function will default to "continuous"

**Details**

This function takes in a single argument of `.x` a vector

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**Examples**

```
x <- mtcars$mpg
tidy_empirical(.x = x, .distribution_type = "continuous")
tidy_empirical(.x = x, .num_sims = 10, .distribution_type = "continuous")
```

---

tidy_exponential	<i>Tidy Randomly Generated Exponential Distribution Tibble</i>
------------------	--

---

**Description**

This function will generate  $n$  random points from a exponential distribution with a user provided, `.rate`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_exponential(.n = 50, .rate = 1, .num_sims = 1, .return_tibble = TRUE)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.rate</code>	A vector of rates
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rexp()`, and its underlying `p`, `d`, and `q` functions. For more information please see `stats::rexp()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3667.htm>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Exponential: `tidy_inverse_exponential()`, `util_exponential_param_estimate()`, `util_exponential_stats_`

**Examples**

```
tidy_exponential()
```

---

tidy\_f

*Tidy Randomly Generated F Distribution Tibble*

---

**Description**

This function will generate `n` random points from a `rf` distribution with a user provided, `df1`, `df2`, and `ncp`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.



**Usage**

```
tidy_f(  
  .n = 50,  
  .df1 = 1,  
  .df2 = 1,  
  .ncp = 0,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

**Arguments**

.n	The number of randomly generated points you want.
.df1	Degrees of freedom, Inf is allowed.
.df2	Degrees of freedom, Inf is allowed.
.ncp	Non-centrality parameter.
.num_sims	The number of randomly generated simulations you want.
.return_tibble	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rf()`, and its underlying p, d, and q functions. For more information please see [stats::rf\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3665.htm>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other F Distribution: [util\\_f\\_param\\_estimate\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#)

**Examples**

```
tidy_f()
```

---

tidy\_four\_autoplot      *Automatic Plot of Density Data*


---

### Description

This is an auto plotting function that will take in a `tidy_` distribution function and a few arguments, one being the plot type, which is a quoted string of one of the following:

- density
- quantile
- probablity
- qq
- mcmc

If the number of simulations exceeds 9 then the legend will not print. The plot subtitle is put together by the attributes of the table passed to the function.

### Usage

```
tidy_four_autoplot(
  .data,
  .line_size = 0.5,
  .geom_point = FALSE,
  .point_size = 1,
  .geom_rug = FALSE,
  .geom_smooth = FALSE,
  .geom_jitter = FALSE,
  .interactive = FALSE
)
```

### Arguments

<code>.data</code>	The data passed in from a <code>tidy_distribution</code> function like <code>tidy_normal()</code>
<code>.line_size</code>	The size param <code>ggplot</code>
<code>.geom_point</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return a plot with <code>ggplot2::geom_point()</code>
<code>.point_size</code>	The point size param for <code>ggplot</code>
<code>.geom_rug</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_rug()</code>
<code>.geom_smooth</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_smooth()</code> The aes parameter of <code>group</code> is set to FALSE. This ensures a single smoothing band returned with <code>SE</code> also set to FALSE. Color is set to 'black' and <code>linetype</code> is 'dashed'.
<code>.geom_jitter</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_jitter()</code>
<code>.interactive</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return an interactive <code>plotly</code> plot.

## Details

This function will spit out one of the following plots:

- density
- quantile
- probability
- qq
- mcmc

## Value

A ggplot or a plotly plot.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Autoplot: [bootstrap\\_stat\\_plot\(\)](#), [tidy\\_autoplot\(\)](#), [tidy\\_combined\\_autoplot\(\)](#), [tidy\\_multi\\_dist\\_autoplot\(\)](#), [tidy\\_random\\_walk\\_autoplot\(\)](#)

## Examples

```
tidy_normal(.num_sims = 5) |>
  tidy_four_autoplot()
```

---

tidy\_gamma

*Tidy Randomly Generated Gamma Distribution Tibble*

---

## Description

This function will generate  $n$  random points from a gamma distribution with a user provided, `.shape`, `.scale`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_gamma(  
  .n = 50,  
  .shape = 1,  
  .scale = 0.3,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	This is strictly 0 to infinity.
<code>.scale</code>	The standard deviation of the randomly generated data. This is strictly from 0 to infinity.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rgamma()`, and its underlying `p`, `d`, and `q` functions. For more information please see [stats::rgamma\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.statology.org/fit-gamma-distribution-to-dataset-in-r/>

[https://en.wikipedia.org/wiki/Gamma\\_distribution](https://en.wikipedia.org/wiki/Gamma_distribution)

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Gamma: [tidy\\_inverse\\_gamma\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#)

**Examples**

```
tidy_gamma()
```

---

tidy\_generalized\_beta *Tidy Randomly Generated Generalized Beta Distribution Tibble*

---

### Description

This function will generate  $n$  random points from a generalized beta distribution with a user provided, `.shape1`, `.shape2`, `.shape3`, `.rate`, and/or `.scale`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

### Usage

```
tidy_generalized_beta(
  .n = 50,
  .shape1 = 1,
  .shape2 = 1,
  .shape3 = 1,
  .rate = 1,
  .scale = 1/.rate,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

### Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.shape1</code>	A non-negative parameter of the Beta distribution.
<code>.shape2</code>	A non-negative parameter of the Beta distribution.
<code>.shape3</code>	A non-negative parameter of the Beta distribution.
<code>.rate</code>	An alternative way to specify the <code>.scale</code> parameter.
<code>.scale</code>	Must be strictly positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rbeta()`, and its underlying `p`, `d`, and `q` functions. For more information please see `stats::rbeta()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://statisticsglobe.com/beta-distribution-in-r-dbeta-pbeta-qbeta-rbeta>

[https://en.wikipedia.org/wiki/Beta\\_distribution](https://en.wikipedia.org/wiki/Beta_distribution)

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Beta: `tidy_beta()`, `util_beta_param_estimate()`, `util_beta_stats_tbl()`

**Examples**

```
tidy_generalized_beta()
```

---

```
tidy_generalized_pareto
```

*Tidy Randomly Generated Generalized Pareto Distribution Tibble*

---

**Description**

This function will generate `n` random points from a generalized Pareto distribution with a user provided, `.shape1`, `.shape2`, `.rate` or `.scale` and number of `#'` random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.

- y The randomly generated data point.
- dx The x value from the `stats::density()` function.
- dy The y value from the `stats::density()` function.
- p The values from the resulting `p_` function of the distribution family.
- q The values from the resulting `q_` function of the distribution family.

### Usage

```
tidy_generalized_pareto(  
  .n = 50,  
  .shape1 = 1,  
  .shape2 = 1,  
  .rate = 1,  
  .scale = 1/.rate,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

### Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.shape1</code>	Must be positive.
<code>.shape2</code>	Must be positive.
<code>.rate</code>	An alternative way to specify the <code>.scale</code> argument
<code>.scale</code>	Must be positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

### Details

This function uses the underlying `actuar::rgenpareto()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rgenpareto\(\)](#)

### Value

A tibble of randomly generated data.

### Author(s)

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Pareto: `tidy_inverse_pareto()`, `tidy_pareto()`, `tidy_pareto1()`, `util_pareto1_aic()`, `util_pareto1_param_estimate()`, `util_pareto1_stats_tbl()`, `util_pareto_param_estimate()`, `util_pareto_stats_tbl()`

**Examples**

```
tidy_generalized_pareto()
```

---

tidy\_geometric

*Tidy Randomly Generated Geometric Distribution Tibble*

---

**Description**

This function will generate `n` random points from a geometric distribution with a user provided, `.prob`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_geometric(.n = 50, .prob = 1, .num_sims = 1, .return_tibble = TRUE)
```



**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.prob</code>	A probability of success in each trial $0 < \text{prob} \leq 1$ .
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rgeom()`, and its underlying `p`, `d`, and `q` functions. For more information please see [stats::rgeom\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

[https://en.wikipedia.org/wiki/Geometric\\_distribution](https://en.wikipedia.org/wiki/Geometric_distribution)

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Geometric: `tidy_zero_truncated_geometric()`, `util_geometric_param_estimate()`, `util_geometric_stats_tbl()`

**Examples**

```
tidy_geometric()
```

**Description**

This function will generate  $n$  random points from a hypergeometric distribution with a user provided,  $m, nn$ , and  $k$ , and number of random simulations to be produced. The function returns a tibble with the simulation number column the  $x$  column which corresponds to the  $n$  randomly generated points, the  $d_$ ,  $p_$  and  $q_$  data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting  $p_$  function of the distribution family.
- `q` The values from the resulting  $q_$  function of the distribution family.

**Usage**

```
tidy_hypergeometric(
  .n = 50,
  .m = 0,
  .nn = 0,
  .k = 0,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.m</code>	The number of white balls in the urn
<code>.nn</code>	The number of black balls in the urn
<code>.k</code>	The number of balls drawn fro the urn.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rhyper()`, and its underlying  $p$ ,  $d$ , and  $q$  functions. For more information please see [stats::rhyper\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

[https://en.wikipedia.org/wiki/Hypergeometric\\_distribution](https://en.wikipedia.org/wiki/Hypergeometric_distribution)

Other Discrete Distribution: `tidy_bernoulli()`, `tidy_binomial()`, `tidy_negative_binomial()`, `tidy_poisson()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `tidy_zero_truncated_poisson()`

Other Hypergeometric: `util_hypergeometric_param_estimate()`, `util_hypergeometric_stats_tbl()`

**Examples**

```
tidy_hypergeometric()
```

---

tidy_inverse_burr	<i>Tidy Randomly Generated Inverse Burr Distribution Tibble</i>
-------------------	---

---

**Description**

This function will generate  $n$  random points from an Inverse Burr distribution with a user provided, `.shape1`, `.shape2`, `.scale`, `.rate`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_inverse_burr(
  .n = 50,
  .shape1 = 1,
  .shape2 = 1,
  .rate = 1,
  .scale = 1/.rate,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape1</code>	Must be strictly positive.
<code>.shape2</code>	Must be strictly positive.
<code>.rate</code>	An alternative way to specify the <code>.scale</code> .
<code>.scale</code>	Must be strictly positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rinvburr()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rinvburr\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Burr: [tidy\\_burr\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#)

Other Inverse Distribution: [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#)

**Examples**

```
tidy_inverse_burr()
```

---

tidy\_inverse\_exponential

*Tidy Randomly Generated Inverse Exponential Distribution Tibble*


---

## Description

This function will generate  $n$  random points from an inverse exponential distribution with a user provided, `.rate` or `.scale` and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

## Usage

```
tidy_inverse_exponential(
  .n = 50,
  .rate = 1,
  .scale = 1/.rate,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

## Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.rate</code>	An alternative way to specify the <code>.scale</code>
<code>.scale</code>	Must be strictly positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

## Details

This function uses the underlying `actuar::rinvexp()`, and its underlying `p`, `d`, and `q` functions. For more information please see `actuar::rinvexp()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Exponential: `tidy_exponential()`, `util_exponential_param_estimate()`, `util_exponential_stats_tbl()`

Other Inverse Distribution: `tidy_inverse_burr()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`

**Examples**

```
tidy_inverse_exponential()
```

---

`tidy_inverse_gamma`      *Tidy Randomly Generated Inverse Gamma Distribution Tibble*

---

**Description**

This function will generate  $n$  random points from an inverse gamma distribution with a user provided, `.shape`, `.rate`, `.scale`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_inverse_gamma(
  .n = 50,
  .shape = 1,
  .rate = 1,
  .scale = 1/.rate,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	Must be strictly positive.
<code>.rate</code>	An alternative way to specify the <code>.scale</code>
<code>.scale</code>	Must be strictly positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rinvgamma()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rinvgamma\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Gamma: [tidy\\_gamma\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#)

Other Inverse Distribution: [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#)

**Examples**

```
tidy_inverse_gamma()
```

---

```
tidy_inverse_normal
```

*Tidy Randomly Generated Inverse Gaussian Distribution Tibble*

---

**Description**

This function will generate  $n$  random points from an Inverse Gaussian distribution with a user provided, `.mean`, `.shape`, `.dispersion`. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_inverse_normal(
  .n = 50,
  .mean = 1,
  .shape = 1,
  .dispersion = 1/.shape,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

- |                             |   |
|-----------------------------|---|
| <code>.n</code>             | The number of randomly generated points you want.                                     |
| <code>.mean</code>          | Must be strictly positive.  |
| <code>.shape</code>         | Must be strictly positive.  |
| <code>.dispersion</code>    | An alternative way to specify the <code>.shape</code> .                               |
| <code>.num_sims</code>      | The number of randomly generated simulations you want.                                |
| <code>.return_tibble</code> | A logical value indicating whether to return the result as a tibble. Default is TRUE. |



**Details**

This function uses the underlying actuar::rinvgauss(). For more information please see [rinvgauss\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Gaussian: [tidy\\_normal\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#)

Other Inverse Distribution: [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#)

**Examples**

```
tidy_inverse_normal()
```

---

tidy\_inverse\_pareto     *Tidy Randomly Generated Inverse Pareto Distribution Tibble*

---

**Description**

This function will generate n random points from an inverse pareto distribution with a user provided, .shape, .scale, and number of random simulations to be produced. The function returns a tibble with the simulation number column the x column which corresponds to the n randomly generated points, the d\_, p\_ and q\_ data points as well.

The data is returned un-grouped.

The columns that are output are:

- sim\_number The current simulation number.
- x The current value of n for the current simulation.
- y The randomly generated data point.
- dx The x value from the [stats::density\(\)](#) function.
- dy The y value from the [stats::density\(\)](#) function.
- p The values from the resulting p\_ function of the distribution family.
- q The values from the resulting q\_ function of the distribution family.

**Usage**

```
tidy_inverse_pareto(
  .n = 50,
  .shape = 1,
  .scale = 1,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	Must be positive.
<code>.scale</code>	Must be positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rinvpareto()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rinvpareto\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Pareto: [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#)

Other Inverse Distribution: [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_weibull\(\)](#)

**Examples**

```
tidy_inverse_pareto()
```

---

tidy\_inverse\_weibull *Tidy Randomly Generated Inverse Weibull Distribution Tibble*

---

**Description**

This function will generate  $n$  random points from a weibull distribution with a user provided, `.shape`, `.scale`, `.rate`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_inverse_weibull(
  .n = 50,
  .shape = 1,
  .rate = 1,
  .scale = 1/.rate,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	Must be strictly positive.
<code>.rate</code>	An alternative way to specify the <code>.scale</code> .
<code>.scale</code>	Must be strictly positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rinvweibull()`, and its underlying p, d, and q functions. For more information please see `actuar::rinvweibull()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Weibull: `tidy_weibull()`, `util_weibull_param_estimate()`, `util_weibull_stats_tbl()`

Other Inverse Distribution: `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`

**Examples**

```
tidy_inverse_weibull()
```

---

tidy_kurtosis_vec	<i>Compute Kurtosis of a Vector</i>
-------------------	-------------------------------------

---

**Description**

This function takes in a vector as it's input and will return the kurtosis of that vector. The length of this vector must be at least four numbers. The kurtosis explains the sharpness of the peak of a distribution of data.

$$\left(\frac{1}{n} * \sum(x - \mu)^4\right) / \left(\left(\frac{1}{n} * \sum(x - \mu)^2\right)^2\right)$$

**Usage**

```
tidy_kurtosis_vec(.x)
```

**Arguments**

`.x` A numeric vector of length four or more.

**Details**

A function to return the kurtosis of a vector.

**Value**

The kurtosis of a vector

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://en.wikipedia.org/wiki/Kurtosis>

Other Vector Function: `bootstrap_p_vec()`, `bootstrap_q_vec()`, `cgmean()`, `chmean()`, `ckurtosis()`, `cmean()`, `cmedian()`, `csd()`, `cskewness()`, `cvar()`, `tidy_scale_zero_one_vec()`, `tidy_skewness_vec()`

Other Statistic: `ci_hi()`, `ci_lo()`, `tidy_range_statistic()`, `tidy_skewness_vec()`, `tidy_stat_tbl()`

Other Vector Function: `bootstrap_p_vec()`, `bootstrap_q_vec()`, `cgmean()`, `chmean()`, `ckurtosis()`, `cmean()`, `cmedian()`, `csd()`, `cskewness()`, `cvar()`, `tidy_scale_zero_one_vec()`, `tidy_skewness_vec()`

**Examples**

```
tidy_kurtosis_vec(rnorm(100, 3, 2))
```

---

tidy\_logistic

*Tidy Randomly Generated Logistic Distribution Tibble*

---

**Description**

This function will generate `n` random points from a logistic distribution with a user provided, `.location`, `.scale`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_logistic(  
  .n = 50,  
  .location = 0,  
  .scale = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.location</code>	The location parameter
<code>.scale</code>	The scale parameter
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rlogis()`, and its underlying `p`, `d`, and `q` functions. For more information please see [stats::rlogis\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

[https://en.wikipedia.org/wiki/Logistic\\_distribution](https://en.wikipedia.org/wiki/Logistic_distribution)

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Logistic: [tidy\\_paralogistic\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#)

**Examples**

```
tidy_logistic()
```

tidy\_lognormal

*Tidy Randomly Generated Lognormal Distribution Tibble***Description**

This function will generate  $n$  random points from a lognormal distribution with a user provided, `.meanlog`, `.sdlog`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_lognormal(
  .n = 50,
  .meanlog = 0,
  .sdlog = 1,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.meanlog</code>	Mean of the distribution on the log scale with default 0
<code>.sdlog</code>	Standard deviation of the distribution on the log scale with default 1
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rlnorm()`, and its underlying `p`, `d`, and `q` functions. For more information please see `stats::rlnorm()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3669.htm>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Lognormal: `util_lognormal_param_estimate()`, `util_lognormal_stats_tbl()`

**Examples**

```
tidy_lognormal()
```

---

tidy\_mcmc\_sampling      *Tidy MCMC Sampling*

---

**Description**

This function performs Markov Chain Monte Carlo (MCMC) sampling on the input data and returns tidy data and a plot representing the results.

**Usage**

```
tidy_mcmc_sampling(.x, .fns = "mean", .cum_fns = "cmean", .num_sims = 2000)
```

**Arguments**

<code>.x</code>	The data vector for MCMC sampling.
<code>.fns</code>	The function(s) to apply to each MCMC sample. Default is "mean".
<code>.cum_fns</code>	The function(s) to apply to the cumulative MCMC samples. Default is "cmean".
<code>.num_sims</code>	The number of simulations. Default is 2000.



**Details**

Perform MCMC sampling and return tidy data and a plot.

The function takes a data vector as input and performs MCMC sampling with the specified number of simulations. It applies user-defined functions to each MCMC sample and to the cumulative MCMC samples. The resulting data is formatted in a tidy format, suitable for further analysis. Additionally, a plot is generated to visualize the MCMC samples and cumulative statistics.

**Value**

A list containing tidy data and a plot.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

**Examples**

```
# Generate MCMC samples
set.seed(123)
data <- rnorm(100)
result <- tidy_mcmc_sampling(data, "median", "cmedian", 500)
result
```

---

tidy\_mixture\_density *Tidy Mixture Data*

---

**Description**

Create mixture model data and resulting density and line plots.

**Usage**

```
tidy_mixture_density(...)
```

## Arguments

... The random data you want to pass. Example `rnorm(50,0,1)` or something like `tidy_normal(.mean = 5, .sd = 1)`

## Details

This function allows you to make mixture model data. It allows you to produce density data and plots for data that is not strictly of one family or of one single type of distribution with a given set of parameters.

For example this function will allow you to mix say `tidy_normal(.mean = 0, .sd = 1)` and `tidy_normal(.mean = 5, .sd = 1)` or you can mix and match distributions.

The output is a list object with three components.

### 1. Data

- `input_data` (The random data passed)
- `dist_tbl` (A tibble of the passed random data)
- `density_tbl` (A tibble of the x and y data from `stats::density()`)

### 1. Plots

- `line_plot` - Plots the `dist_tbl`
- `dens_plot` - Plots the `density_tbl`

### 1. Input Functions

- `input_fns` - A list of the functions and their parameters passed to the function itself

## Value

A list object

## Author(s)

Steven P. Sanderson II, MPH

## Examples

```
output <- tidy_mixture_density(rnorm(100, 0, 1), tidy_normal(.mean = 5, .sd = 1))
```

```
output$data
```

```
output$plots
```

```
output$input_fns
```

---

tidy\_multi\_dist\_autoplot

*Automatic Plot of Multi Dist Data*


---

## Description

This is an auto plotting function that will take in a `tidy_` distribution function and a few arguments, one being the plot type, which is a quoted string of one of the following:

- density
- quantile
- probablity
- qq
- mcmc

If the number of simulations exceeds 9 then the legend will not print. The plot subtitle is put together by the attributes of the table passed to the function.

## Usage

```
tidy_multi_dist_autoplot(
  .data,
  .plot_type = "density",
  .line_size = 0.5,
  .geom_point = FALSE,
  .point_size = 1,
  .geom_rug = FALSE,
  .geom_smooth = FALSE,
  .geom_jitter = FALSE,
  .interactive = FALSE
)
```

## Arguments

<code>.data</code>	The data passed in from a the function <code>tidy_multi_dist()</code>
<code>.plot_type</code>	This is a quoted string like 'density'
<code>.line_size</code>	The size param <code>ggplot</code>
<code>.geom_point</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return a plot with <code>ggplot2::geom_point()</code>
<code>.point_size</code>	The point size param for <code>ggplot</code>
<code>.geom_rug</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_rug()</code>
<code>.geom_smooth</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_smooth()</code> The <code>aes</code> parameter of <code>group</code> is set to FALSE. This ensures a single smoothing band returned with <code>SE</code> also set to FALSE. Color is set to 'black' and <code>linetype</code> is 'dashed'.

<code>.geom_jitter</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of <code>ggplot2::geom_jitter()</code>
<code>.interactive</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return an interactive plotly plot.

### Details

This function will spit out one of the following plots:

- density
- quantile
- probability
- qq
- mcmc

### Value

A ggplot or a plotly plot.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Autoplot: [bootstrap\\_stat\\_plot\(\)](#), [tidy\\_autoplot\(\)](#), [tidy\\_combined\\_autoplot\(\)](#), [tidy\\_four\\_autoplot\(\)](#), [tidy\\_random\\_walk\\_autoplot\(\)](#)

### Examples

```
tn <- tidy_multi_single_dist(  
  .tidy_dist = "tidy_normal",  
  .param_list = list(  
    .n = 100,  
    .mean = c(-2, 0, 2),  
    .sd = 1,  
    .num_sims = 5,  
    .return_tibble = TRUE  
  )  
)  
  
tn |>  
  tidy_multi_dist_autoplot()  
  
tn |>  
  tidy_multi_dist_autoplot(.plot_type = "qq")
```

---

`tidy_multi_single_dist`*Generate Multiple Tidy Distributions of a single type*

---

**Description**

Generate multiple distributions of data from the same tidy\_ distribution function.

**Usage**

```
tidy_multi_single_dist(.tidy_dist = NULL, .param_list = list())
```

**Arguments**

<code>.tidy_dist</code>	The type of tidy_ distribution that you want to run. You can only choose one.
<code>.param_list</code>	This must be a <code>list()</code> object of the parameters that you want to pass through to the TidyDensity tidy_ distribution function.

**Details**

Generate multiple distributions of data from the same tidy\_ distribution function. This allows you to simulate multiple distributions of the same family in order to view how shapes change with parameter changes. You can then visualize the differences however you choose.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Multiple Distribution: [tidy\\_combine\\_distributions\(\)](#)

**Examples**

```
tidy_multi_single_dist(  
  .tidy_dist = "tidy_normal",  
  .param_list = list(  
    .n = 50,  
    .mean = c(-1, 0, 1),  
    .sd = 1,  
    .num_sims = 3,  
    .return_tibble = TRUE  
  )  
)
```

```

)

tidy_multi_single_dist(
  .tidy_dist = "tidy_normal",
  .param_list = list(
    .n = 50,
    .mean = c(-1, 0, 1),
    .sd = 1,
    .num_sims = 3,
    .return_tibble = FALSE
  )
)

```

---

tidy\_negative\_binomial

*Tidy Randomly Generated Negative Binomial Distribution Tibble*

---

## Description

This function will generate  $n$  random points from a negative binomial distribution with a user provided, `.size`, `.prob`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

## Usage

```

tidy_negative_binomial(
  .n = 50,
  .size = 1,
  .prob = 0.1,
  .num_sims = 1,
  .return_tibble = TRUE
)

```

## Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.size</code>	target for number of successful trials, or dispersion parameter (the shape parameter of the gamma mixing distribution). Must be strictly positive, need not be integer.
<code>.prob</code>	Probability of success on each trial where $0 < .prob \leq 1$ .
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

## Details

This function uses the underlying `stats::rnbinom()`, and its underlying `p`, `d`, and `q` functions. For more information please see `stats::rnbinom()`

## Value

A tibble of randomly generated data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Discrete Distribution: `tidy_bernoulli()`, `tidy_binomial()`, `tidy_hypergeometric()`, `tidy_poisson()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `tidy_zero_truncated_poisson()`

Other Binomial: `tidy_binomial()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `util_binomial_param_estimate()`, `util_binomial_stats_tbl()`, `util_negative_binomial_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_negative_binomial_stats_`

## Examples

```
tidy_negative_binomial()
```

---

`tidy_normal`*Tidy Randomly Generated Gaussian Distribution Tibble*

---

### Description

This function will generate `n` random points from a Gaussian distribution with a user provided, `.mean`, `.sd` - standard deviation and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `dnorm`, `pnorm` and `qnorm` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

### Usage

```
tidy_normal(.n = 50, .mean = 0, .sd = 1, .num_sims = 1, .return_tibble = TRUE)
```

### Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.mean</code>	The mean of the randomly generated data.
<code>.sd</code>	The standard deviation of the randomly generated data.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

### Details

This function uses the underlying `stats::rnorm()`, `stats::pnorm()`, and `stats::qnorm()` functions to generate data from the given parameters. For more information please see [stats::rnorm\(\)](#)

### Value

A tibble of randomly generated data.

### Author(s)

Steven P. Sanderson II, MPH



**See Also**

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Gaussian: `tidy_inverse_normal()`, `util_normal_param_estimate()`, `util_normal_stats_tbl()`

**Examples**

```
tidy_normal()
```

---

<code>tidy_paralogistic</code>	<i>Tidy Randomly Generated Paralogistic Distribution Tibble</i>
--------------------------------	---

---

**Description**

This function will generate  $n$  random points from a paralogistic distribution with a user provided, `.shape`, `.rate`, `.scale` and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_paralogistic(
  .n = 50,
  .shape = 1,
  .rate = 1,
  .scale = 1/.rate,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

### Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	Must be strictly positive.
<code>.rate</code>	An alternative way to specify the <code>.scale</code>
<code>.scale</code>	Must be strictly positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

### Details

This function uses the underlying `actuar::rparalogis()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rparalogis\(\)](#)

### Value

A tibble of randomly generated data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

[https://en.wikipedia.org/wiki/Logistic\\_distribution](https://en.wikipedia.org/wiki/Logistic_distribution)

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Logistic: [tidy\\_logistic\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#)

### Examples

```
tidy_paralogistic()
```

tidy\_pareto

*Tidy Randomly Generated Pareto Distribution Tibble***Description**

This function will generate  $n$  random points from a pareto distribution with a user provided, `.shape`, `.scale`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_pareto(
  .n = 50,
  .shape = 10,
  .scale = 0.1,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	Must be positive.
<code>.scale</code>	Must be positive.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rpareto()`, and its underlying `p`, `d`, and `q` functions. For more information please see `actuar::rpareto()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Pareto: `tidy_generalized_pareto()`, `tidy_inverse_pareto()`, `tidy_pareto1()`, `util_pareto1_aic()`, `util_pareto1_param_estimate()`, `util_pareto1_stats_tbl()`, `util_pareto_param_estimate()`, `util_pareto_stats_tbl()`

**Examples**

```
tidy_pareto()
```

---

tidy_pareto1	<i>Tidy Randomly Generated Pareto Single Parameter Distribution Tibble</i>
--------------	--

---

**Description**

This function will generate  $n$  random points from a single parameter pareto distribution with a user provided, `.shape`, `.min`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_pareto1(
  .n = 50,
  .shape = 1,
  .min = 1,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	Must be positive.
<code>.min</code>	The lower bound of the support of the distribution.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rpareto1()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rpareto1\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Pareto: [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_pareto\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#)

**Examples**

```
tidy_pareto1()
```

---

`tidy_poisson`*Tidy Randomly Generated Poisson Distribution Tibble*

---

## Description

This function will generate `n` random points from a Poisson distribution with a user provided, `.lambda`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

## Usage

```
tidy_poisson(.n = 50, .lambda = 1, .num_sims = 1, .return_tibble = TRUE)
```

## Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.lambda</code>	A vector of non-negative means.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

## Details

This function uses the underlying `stats::rpois()`, and its underlying `p`, `d`, and `q` functions. For more information please see `stats::rpois()`

## Value

A tibble of randomly generated data.

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

<https://r-coder.com/poisson-distribution-r/>

[https://en.wikipedia.org/wiki/Poisson\\_distribution](https://en.wikipedia.org/wiki/Poisson_distribution)

Other Poisson: `tidy_zero_truncated_poisson()`, `util_poisson_param_estimate()`, `util_poisson_stats_tbl()`, `util_zero_truncated_poisson_param_estimate()`, `util_zero_truncated_poisson_stats_tbl()`

Other Discrete Distribution: `tidy_bernoulli()`, `tidy_binomial()`, `tidy_hypergeometric()`, `tidy_negative_binomial()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `tidy_zero_truncated_poisson()`

**Examples**

```
tidy_poisson()
```

---

tidy_random_walk	<i>Tidy Random Walk</i>
------------------	-------------------------

---

**Description**

Takes in the data from a `tidy_` distribution function and applies a random walk calculation of either `cum_prod` or `cum_sum` to `y`.

**Usage**

```
tidy_random_walk(
  .data,
  .initial_value = 0,
  .sample = FALSE,
  .replace = FALSE,
  .value_type = "cum_prod"
)
```

**Arguments**

<code>.data</code>	The data that is being passed from a <code>tidy_</code> distribution function.
<code>.initial_value</code>	The default is 0, this can be set to whatever you want.
<code>.sample</code>	This is a boolean value TRUE/FALSE. The default is FALSE. If set to TRUE then the <code>y</code> value from the <code>tidy_</code> distribution function is sampled.
<code>.replace</code>	This is a boolean value TRUE/FALSE. The default is FALSE. If set to TRUE AND <code>.sample</code> is set to TRUE then the <code>replace</code> parameter of the <code>sample</code> function will be set to TRUE.
<code>.value_type</code>	This can take one of three different values for now. These are the following: <ul style="list-style-type: none"> <li>"cum_prod" - This will take the cumprod of <code>y</code></li> <li>"cum_sum" - This will take the cumsum of <code>y</code></li> </ul>

## Details

Monte Carlo simulations were first formally designed in the 1940's while developing nuclear weapons, and since have been heavily used in various fields to use randomness solve problems that are potentially deterministic in nature. In finance, Monte Carlo simulations can be a useful tool to give a sense of how assets with certain characteristics might behave in the future. While there are more complex and sophisticated financial forecasting methods such as ARIMA (Auto-Regressive Integrated Moving Average) and GARCH (Generalised Auto-Regressive Conditional Heteroskedasticity) which attempt to model not only the randomness but underlying macro factors such as seasonality and volatility clustering, Monte Carlo random walks work surprisingly well in illustrating market volatility as long as the results are not taken too seriously.

## Value

An ungrouped tibble.

## Author(s)

Steven P. Sanderson II, MPH

## Examples

```
tidy_normal(.sd = .1, .num_sims = 25) %>%  
  tidy_random_walk()
```

---

tidy\_random\_walk\_autoplot

*Automatic Plot of Random Walk Data*

---

## Description

This is an auto-plotting function that will take in a tidy\_ distribution function and a few arguments with regard to the output of the visualization.

If the number of simulations exceeds 9 then the legend will not print. The plot subtitle is put together by the attributes of the table passed to the function.

## Usage

```
tidy_random_walk_autoplot(  
  .data,  
  .line_size = 0.5,  
  .geom_rug = FALSE,  
  .geom_smooth = FALSE,  
  .interactive = FALSE  
)
```



### Arguments

<code>.data</code>	The data passed in from a tidy_distribution function like tidy_normal()
<code>.line_size</code>	The size param ggplot
<code>.geom_rug</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of ggplot2::geom_rug()
<code>.geom_smooth</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return the use of ggplot2::geom_smooth() The aes parameter of group is set to FALSE. This ensures a single smoothing band returned with SE also set to FALSE. Color is set to 'black' and linetype is 'dashed'.
<code>.interactive</code>	A Boolean value of TRUE/FALSE, FALSE is the default. TRUE will return an interactive plotly plot.

### Details

This function will produce a simple random walk plot from a tidy\_ distribution function.

### Value

A ggplot or a plotly plot.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Autoplot: [bootstrap\\_stat\\_plot\(\)](#), [tidy\\_autoplot\(\)](#), [tidy\\_combined\\_autoplot\(\)](#), [tidy\\_four\\_autoplot\(\)](#), [tidy\\_multi\\_dist\\_autoplot\(\)](#)

### Examples

```
tidy_normal(sd = .1, .num_sims = 5) |>
  tidy_random_walk(.value_type = "cum_sum") |>
  tidy_random_walk_autoplot()
```

```
tidy_normal(sd = .1, .num_sims = 20) |>
  tidy_random_walk(.value_type = "cum_sum", .sample = TRUE, .replace = TRUE) |>
  tidy_random_walk_autoplot()
```

---

tidy\_range\_statistic *Get the range statistic*

---

### Description

Takes in a numeric vector and returns back the range of that vector

### Usage

```
tidy_range_statistic(.x)
```

### Arguments

`.x`                    A numeric vector

### Details

Takes in a numeric vector and returns the range of that vector using the `diff` and `range` functions.

### Value

A single number, the range statistic

### Author(s)

Steven P. Sandeson II, MPH

### See Also

Other Statistic: [ci\\_hi\(\)](#), [ci\\_lo\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#), [tidy\\_stat\\_tbl\(\)](#)

### Examples

```
tidy_range_statistic(seq(1:10))
```

---

`tidy_scale_zero_one_vec`*Vector Function Scale to Zero and One*

---

**Description**

Takes a numeric vector and will return a vector that has been scaled from  $[0, 1]$

**Usage**

```
tidy_scale_zero_one_vec(.x)
```

**Arguments**

`.x` A numeric vector to be scaled from  $[0, 1]$  inclusive.

**Details**

Takes a numeric vector and will return a vector that has been scaled from  $[0, 1]$  The input vector must be numeric. The computation is fairly straightforward. This may be helpful when trying to compare the distributions of data where a distribution like beta which requires data to be between 0 and 1

$$y[h] = (x - \min(x)) / (\max(x) - \min(x))$$

**Value**

A numeric vector

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_skewness\\_vec\(\)](#)

**Examples**

```
vec_1 <- rnorm(100, 2, 1)
vec_2 <- tidy_scale_zero_one_vec(vec_1)

dens_1 <- density(vec_1)
dens_2 <- density(vec_2)
max_x <- max(dens_1$x, dens_2$x)
max_y <- max(dens_1$y, dens_2$y)
plot(dens_1,
```

```

  asp = max_y / max_x, main = "Density vec_1 (Red) and vec_2 (Blue)",
  col = "red", xlab = "", ylab = "Density of Vec 1 and Vec 2"
)
lines(dens_2, col = "blue")

```

---

tidy\_skewness\_vec      *Compute Skewness of a Vector*

---

### Description

This function takes in a vector as it's input and will return the skewness of that vector. The length of this vector must be at least four numbers. The skewness explains the 'tailedness' of the distribution of data.

$$\frac{((1/n) * \sum(x - \mu)^3)}{(((1/n) * \sum(x - \mu)^2)^{3/2})}$$

### Usage

```
tidy_skewness_vec(.x)
```

### Arguments

.x                    A numeric vector of length four or more.

### Details

A function to return the skewness of a vector.

### Value

The skewness of a vector

### Author(s)

Steven P. Sanderson II, MPH

### See Also

<https://en.wikipedia.org/wiki/Skewness>

Other Statistic: [ci\\_hi\(\)](#), [ci\\_lo\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_range\\_statistic\(\)](#), [tidy\\_stat\\_tbl\(\)](#)

Other Vector Function: [bootstrap\\_p\\_vec\(\)](#), [bootstrap\\_q\\_vec\(\)](#), [cgmean\(\)](#), [chmean\(\)](#), [ckurtosis\(\)](#), [cmean\(\)](#), [cmedian\(\)](#), [csd\(\)](#), [cskewness\(\)](#), [cvar\(\)](#), [tidy\\_kurtosis\\_vec\(\)](#), [tidy\\_scale\\_zero\\_one\\_vec\(\)](#)

### Examples

```
tidy_skewness_vec(rnorm(100, 3, 2))
```

---

tidy_stat_tbl	<i>Tidy Stats of Tidy Distribution</i>
---------------	--

---

## Description

A function to return the stat function values of a given tidy\_ distribution output.

## Usage

```
tidy_stat_tbl(
  .data,
  .x = y,
  .fns,
  .return_type = "vector",
  .use_data_table = FALSE,
  ...
)
```

## Arguments

<code>.data</code>	The input data coming from a tidy_ distribution function.
<code>.x</code>	The default is <code>y</code> but can be one of the other columns from the input data.
<code>.fns</code>	The default is <code>IQR</code> , but this can be any stat function like <code>quantile</code> or <code>median</code> etc.
<code>.return_type</code>	The default is "vector" which returns an <code>sapply</code> object.
<code>.use_data_table</code>	The default is <code>FALSE</code> , <code>TRUE</code> will use <code>data.table</code> under the hood and still return a tibble. If this argument is set to <code>TRUE</code> then the <code>.return_type</code> parameter will be ignored.
<code>...</code>	Addition function arguments to be supplied to the parameters of <code>.fns</code>

## Details

A function to return the value(s) of a given tidy\_ distribution function output and chosen column from it. This function will only work with tidy\_ distribution functions.

There are currently three different output types for this function. These are:

- "vector" - which gives an `sapply()` output
- "list" - which gives an `lapply()` output, and
- "tibble" - which returns a tibble in long format.

Currently you can pass any stat function that performs an operation on a vector input. This means you can pass things like `IQR`, `quantile` and their associated arguments in the `...` portion of the function.

This function also by default will rename the value column of the tibble to the name of the function. This function will also give the column name of `sim_number` for the tibble output with the corresponding simulation numbers as the values.

For the `sapply` and `lapply` outputs the column names will also give the simulation number information by making column names like `sim_number_1` etc.

There is an option of `.use_data_table` which can greatly enhance the speed of the calculations performed if used while still returning a tibble. The calculations are performed after turning the input data into a `data.table` object, performing the necessary calculation and then converting back to a tibble object.

### Value

A return of object of either `sapply` `lapply` or tibble based upon user input.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Statistic: `ci_hi()`, `ci_lo()`, `tidy_kurtosis_vec()`, `tidy_range_statistic()`, `tidy_skewness_vec()`

### Examples

```
tn <- tidy_normal(.num_sims = 3)

p <- c(0.025, 0.25, 0.5, 0.75, 0.95)

tidy_stat_tbl(tn, y, quantile, "vector", probs = p, na.rm = TRUE)
tidy_stat_tbl(tn, y, quantile, "list", probs = p)
tidy_stat_tbl(tn, y, quantile, "tibble", probs = p)
tidy_stat_tbl(tn, y, quantile, .use_data_table = TRUE, probs = p, na.rm = TRUE)
```

---

tidy\_t

*Tidy Randomly Generated T Distribution Tibble*

---

### Description

This function will generate `n` random points from a `rt` distribution with a user provided, `df`, `nrep`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.

- x The current value of n for the current simulation.
- y The randomly generated data point.
- dx The x value from the `stats::density()` function.
- dy The y value from the `stats::density()` function.
- p The values from the resulting p\_ function of the distribution family.
- q The values from the resulting q\_ function of the distribution family.

### Usage

```
tidy_t(.n = 50, .df = 1, .ncp = 0, .num_sims = 1, .return_tibble = TRUE)
```

### Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.df</code>	Degrees of freedom, Inf is allowed.
<code>.ncp</code>	Non-centrality parameter.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

### Details

This function uses the underlying `stats::rt()`, and its underlying p, d, and q functions. For more information please see `stats::rt()`

### Value

A tibble of randomly generated data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3664.htm>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_triangular()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other T Distribution: `util_t_stats_tbl()`

### Examples

```
tidy_t()
```

---

tidy\_triangular      *Generate Tidy Data from Triangular Distribution*

---

### Description

This function generates tidy data from the triangular distribution.

### Usage

```
tidy_triangular(  
  .n = 50,  
  .min = 0,  
  .max = 1,  
  .mode = 1/2,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

### Arguments

<code>.n</code>	The number of x values for each simulation.
<code>.min</code>	The minimum value of the triangular distribution.
<code>.max</code>	The maximum value of the triangular distribution.
<code>.mode</code>	The mode (peak) value of the triangular distribution.
<code>.num_sims</code>	The number of simulations to perform.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

### Details

The function takes parameters for the triangular distribution (minimum, maximum, mode), the number of x values (`n`), the number of simulations (`num_sims`), and an option to return the result as a tibble (`return_tibble`). It performs various checks on the input parameters to ensure validity. The result is a data frame or tibble with tidy data for further analysis.

### Value

A tibble of randomly generated data.

### Author(s)

Steven P. Sanderson II, MPH



**See Also**

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_uniform()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Triangular: `util_triangular_param_estimate()`, `util_triangular_stats_tbl()`

**Examples**

```
tidy_triangular(.return_tibble = TRUE)
```

---

tidy\_uniform

*Tidy Randomly Generated Uniform Distribution Tibble*


---

**Description**

This function will generate `n` random points from a uniform distribution with a user provided, `.min` and `.max` values, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the `n` randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of `n` for the current simulation.
- `y` The randomly generated data point.
- `dx` The `x` value from the `stats::density()` function.
- `dy` The `y` value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_uniform(.n = 50, .min = 0, .max = 1, .num_sims = 1, .return_tibble = TRUE)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.min</code>	A lower limit of the distribution.
<code>.max</code>	An upper limit of the distribution
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::runif()`, and its underlying p, d, and q functions. For more information please see `stats::runif()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3662.htm>

Other Continuous Distribution: `tidy_beta()`, `tidy_burr()`, `tidy_cauchy()`, `tidy_chisquare()`, `tidy_exponential()`, `tidy_f()`, `tidy_gamma()`, `tidy_generalized_beta()`, `tidy_generalized_pareto()`, `tidy_geometric()`, `tidy_inverse_burr()`, `tidy_inverse_exponential()`, `tidy_inverse_gamma()`, `tidy_inverse_normal()`, `tidy_inverse_pareto()`, `tidy_inverse_weibull()`, `tidy_logistic()`, `tidy_lognormal()`, `tidy_normal()`, `tidy_paralogistic()`, `tidy_pareto()`, `tidy_pareto1()`, `tidy_t()`, `tidy_triangular()`, `tidy_weibull()`, `tidy_zero_truncated_geometric()`

Other Uniform: `util_uniform_param_estimate()`, `util_uniform_stats_tbl()`

**Examples**

```
tidy_uniform()
```

---

tidy\_weibull

*Tidy Randomly Generated Weibull Distribution Tibble*

---

**Description**

This function will generate n random points from a weibull distribution with a user provided, `.shape`, `.scale`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the x column which corresponds to the n randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of n for the current simulation.
- `y` The randomly generated data point.
- `dx` The x value from the `stats::density()` function.
- `dy` The y value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_weibull(  
  .n = 50,  
  .shape = 1,  
  .scale = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

**Arguments**

<code>.n</code>	The number of randomly generated points you want.
<code>.shape</code>	Shape parameter defaults to 0.
<code>.scale</code>	Scale parameter defaults to 1.
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `stats::rweibull()`, and its underlying `p`, `d`, and `q` functions. For more information please see [stats::rweibull\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://www.itl.nist.gov/div898/handbook/eda/section3/eda3669.htm>

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#)

Other Weibull: [tidy\\_inverse\\_weibull\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#)

**Examples**

```
tidy_weibull()
```

---

`tidy_zero_truncated_binomial`*Tidy Randomly Generated Binomial Distribution Tibble*

---

## Description

This function will generate  $n$  random points from a zero truncated binomial distribution with a user provided, `.size`, `.prob`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

## Usage

```
tidy_zero_truncated_binomial(  
  .n = 50,  
  .size = 1,  
  .prob = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

## Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.size</code>	Number of trials, zero or more.
<code>.prob</code>	Probability of success on each trial $0 \leq \text{prob} \leq 1$ .
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

## Details

This function uses the underlying `actuar::rztbinom()`, and its underlying `p`, `d`, and `q` functions. For more information please see `actuar::rztbinom()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Discrete Distribution: `tidy_bernoulli()`, `tidy_binomial()`, `tidy_hypergeometric()`, `tidy_negative_binomial()`, `tidy_poisson()`, `tidy_zero_truncated_negative_binomial()`, `tidy_zero_truncated_poisson()`

Other Binomial: `tidy_binomial()`, `tidy_negative_binomial()`, `tidy_zero_truncated_negative_binomial()`, `util_binomial_param_estimate()`, `util_binomial_stats_tbl()`, `util_negative_binomial_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_negative_binomial_stats_tbl()`

Other Zero Truncated Distribution: `tidy_zero_truncated_geometric()`, `tidy_zero_truncated_poisson()`, `util_zero_truncated_binomial_param_estimate()`

**Examples**

```
tidy_zero_truncated_binomial()
```

---

```
tidy_zero_truncated_geometric
```

*Tidy Randomly Generated Zero Truncated Geometric Distribution Tib-  
ble*

---

**Description**

This function will generate  $n$  random points from a zero truncated Geometric distribution with a user provided, `.prob`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

**Usage**

```
tidy_zero_truncated_geometric(
  .n = 50,
  .prob = 1,
  .num_sims = 1,
  .return_tibble = TRUE
)
```

**Arguments**

`.n` The number of randomly generated points you want.

`.prob` A probability of success in each trial  $0 < \text{prob} \leq 1$ .

`.num_sims` The number of randomly generated simulations you want.

`.return_tibble` A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rztgeom()`, and its underlying p, d, and q functions. For more information please see [actuar::rztgeom\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Geometric: [tidy\\_geometric\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#)

Other Continuous Distribution: [tidy\\_beta\(\)](#), [tidy\\_burr\(\)](#), [tidy\\_cauchy\(\)](#), [tidy\\_chisquare\(\)](#), [tidy\\_exponential\(\)](#), [tidy\\_f\(\)](#), [tidy\\_gamma\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_geometric\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [tidy\\_inverse\\_gamma\(\)](#), [tidy\\_inverse\\_normal\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_inverse\\_weibull\(\)](#), [tidy\\_logistic\(\)](#), [tidy\\_lognormal\(\)](#), [tidy\\_normal\(\)](#), [tidy\\_paralogistic\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [tidy\\_t\(\)](#), [tidy\\_triangular\(\)](#), [tidy\\_uniform\(\)](#), [tidy\\_weibull\(\)](#)

Other Zero Truncated Distribution: [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_poisson\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#)

**Examples**

```
tidy_zero_truncated_geometric()
```

---

`tidy_zero_truncated_negative_binomial`*Tidy Randomly Generated Binomial Distribution Tibble*

---

## Description

This function will generate  $n$  random points from a zero truncated binomial distribution with a user provided, `.size`, `.prob`, and number of random simulations to be produced. The function returns a tibble with the simulation number column the `x` column which corresponds to the  $n$  randomly generated points, the `d_`, `p_` and `q_` data points as well.

The data is returned un-grouped.

The columns that are output are:

- `sim_number` The current simulation number.
- `x` The current value of  $n$  for the current simulation.
- `y` The randomly generated data point.
- `dx` The  $x$  value from the `stats::density()` function.
- `dy` The  $y$  value from the `stats::density()` function.
- `p` The values from the resulting `p_` function of the distribution family.
- `q` The values from the resulting `q_` function of the distribution family.

## Usage

```
tidy_zero_truncated_negative_binomial(  
  .n = 50,  
  .size = 0,  
  .prob = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

## Arguments

<code>.n</code>	The number of randomly generated points you want.
<code>.size</code>	Number of trials, zero or more.
<code>.prob</code>	Probability of success on each trial $0 \leq \text{prob} \leq 1$ .
<code>.num_sims</code>	The number of randomly generated simulations you want.
<code>.return_tibble</code>	A logical value indicating whether to return the result as a tibble. Default is TRUE.

## Details

This function uses the underlying `actuar::rztbninom()`, and its underlying `p`, `d`, and `q` functions. For more information please see `actuar::rztbninom()`

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Discrete Distribution: [tidy\\_bernoulli\(\)](#), [tidy\\_binomial\(\)](#), [tidy\\_hypergeometric\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_poisson\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_poisson\(\)](#)

Other Binomial: [tidy\\_binomial\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

Other Zero Truncated Negative Distribution: [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#)

**Examples**

```
tidy_zero_truncated_negative_binomial()
```

---

```
tidy_zero_truncated_poisson
```

*Tidy Randomly Generated Zero Truncated Poisson Distribution Tibble*

---

**Description**

This function will generate n random points from a Zero Truncated Poisson distribution with a user provided, .lambda, and number of random simulations to be produced. The function returns a tibble with the simulation number column the x column which corresponds to the n randomly generated points, the d\_, p\_ and q\_ data points as well.

The data is returned un-grouped.

The columns that are output are:

- sim\_number The current simulation number.
- x The current value of n for the current simulation.
- y The randomly generated data point.
- dx The x value from the [stats::density\(\)](#) function.
- dy The y value from the [stats::density\(\)](#) function.
- p The values from the resulting p\_ function of the distribution family.
- q The values from the resulting q\_ function of the distribution family.



**Usage**

```
tidy_zero_truncated_poisson(  
  .n = 50,  
  .lambda = 1,  
  .num_sims = 1,  
  .return_tibble = TRUE  
)
```

**Arguments**

`.n` The number of randomly generated points you want.

`.lambda` A vector of non-negative means.

`.num_sims` The number of randomly generated simulations you want.

`.return_tibble` A logical value indicating whether to return the result as a tibble. Default is TRUE.

**Details**

This function uses the underlying `actuar::rztpois()`, and its underlying `p`, `d`, and `q` functions. For more information please see [actuar::rztpois\(\)](#)

**Value**

A tibble of randomly generated data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

<https://openacttexts.github.io/Loss-Data-Analytics/ChapSummaryDistributions.html>

Other Poisson: [tidy\\_poisson\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

Other Zero Truncated Distribution: [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#)

Other Discrete Distribution: [tidy\\_bernoulli\(\)](#), [tidy\\_binomial\(\)](#), [tidy\\_hypergeometric\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_poisson\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_negative](#)

**Examples**

```
tidy_zero_truncated_poisson()
```

---

triangle_plot	<i>Triangle Distribution PDF Plot</i>
---------------	---------------------------------------

---

### Description

This function generates a probability density function (PDF) plot for the triangular distribution.

### Usage

```
triangle_plot(.data, .interactive = FALSE)
```

### Arguments

<code>.data</code>	Tidy data from the <code>tidy_triangular</code> function.
<code>.interactive</code>	A logical value indicating whether to return an interactive plot using plotly. Default is FALSE.

### Details

The function checks if the input data is a data frame or tibble, and if it comes from the `tidy_triangular` function. It then extracts necessary attributes for the plot and creates a PDF plot using `ggplot2`. The plot includes data points and segments to represent the triangular distribution.

### Value

The function returns a `ggplot2` object representing the probability density function plot for the triangular distribution.

### Author(s)

Steven P. Sanderson II, MPH

### Examples

```
# Example: Generating a PDF plot for the triangular distribution
data <- tidy_triangular(.n = 50, .min = 0, .max = 1, .mode = 1/2, .num_sims = 1,
  .return_tibble = TRUE)
triangle_plot(data)
```

---

`util_bernoulli_param_estimate`*Estimate Bernoulli Parameters*

---

## Description

This function will attempt to estimate the Bernoulli prob parameter given some vector of values `.x`. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated Bernoulli data.

## Usage

```
util_bernoulli_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function. Must be non-negative integers.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will see if the given vector `.x` is a numeric vector. It will attempt to estimate the prob parameter of a Bernoulli distribution.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#),

```
util_uniform_param_estimate(), util_weibull_param_estimate(), util_zero_truncated_binomial_param_estimate(),  
util_zero_truncated_geometric_param_estimate(), util_zero_truncated_negative_binomial_param_estimate(),  
util_zero_truncated_poisson_param_estimate()
```

Other Bernoulli: `tidy_bernoulli()`, `util_bernoulli_stats_tbl()`

## Examples

```
library(dplyr)  
library(ggplot2)  
  
tb <- tidy_bernoulli(.prob = .1) |> pull(y)  
output <- util_bernoulli_param_estimate(tb)  
  
output$parameter_tbl  
  
output$combined_data_tbl |>  
  tidy_combined_autoplot()
```

---

util\_bernoulli\_stats\_tbl

*Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_bernoulli_stats_tbl(.data)
```

## Arguments

`.data` The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Bernoulli: `tidy_bernoulli()`, `util_bernoulli_param_estimate()`

Other Distribution Statistics: `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_bernoulli() |>
  util_bernoulli_stats_tbl() |>
  glimpse()
```

---

 util\_beta\_aic

---

*Calculate Akaike Information Criterion (AIC) for Beta Distribution*


---

**Description**

This function estimates the parameters of a beta distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_beta_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a beta distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a beta distribution fitted to the provided data.

Initial parameter estimates: The choice of initial values can impact the convergence of the optimization.

Optimization method: You might explore different optimization methods within `optim` for potentially better performance. Data transformation: Depending on your data, you may need to apply transformations (e.g., scaling to  $[0, 1]$  interval) before fitting the beta distribution.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

### Value

The AIC value calculated based on the fitted beta distribution to the provided data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

### Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rbeta(30, 1, 1)
util_beta_aic(x)
```

---

util\_beta\_param\_estimate

*Estimate Beta Parameters*

---

### Description

This function will automatically scale the data from 0 to 1 if it is not already. This means you can pass a vector like `mtcars$mpg` and not worry about it.

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated beta data.

Three different methods of shape parameters are supplied:

- Bayes
- NIST mme
- EnvStats mme, see [EnvStats::ebeta\(\)](#)

**Usage**

```
util_beta_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function. Must be numeric, and all values must be  $0 \leq x \leq 1$

`.auto_gen_empirical` This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will attempt to estimate the beta `shape1` and `shape2` parameters given some vector of values.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Beta: [tidy\\_beta\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

x <- mtcars$mpg
output <- util_beta_param_estimate(x)
```

```
output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

tb <- rbeta(50, 2.5, 1.4)
util_beta_param_estimate(tb)$parameter_tbl
```

---

util\_beta\_stats\_tbl     *Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_beta_stats_tbl(.data)
```

## Arguments

`.data`             The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Beta: [tidy\\_beta\(\)](#), [tidy\\_generalized\\_beta\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#),



```
util_poisson_stats_tbl(), util_t_stats_tbl(), util_triangular_stats_tbl(), util_uniform_stats_tbl(),  
util_weibull_stats_tbl(), util_zero_truncated_binomial_stats_tbl(), util_zero_truncated_geometric_stats_tbl(),  
util_zero_truncated_negative_binomial_stats_tbl(), util_zero_truncated_poisson_stats_tbl()
```

### Examples

```
library(dplyr)  
  
tidy_beta() |>  
  util_beta_stats_tbl() |>  
  glimpse()
```

---

util_binomial_aic	<i>Calculate Akaike Information Criterion (AIC) for Binomial Distribution</i>
-------------------	---

---

### Description

This function estimates the size and probability parameters of a binomial distribution from the provided data and then calculates the AIC value based on the fitted distribution.

### Usage

```
util_binomial_aic(.x)
```

### Arguments

`.x` A numeric vector containing the data to be fitted to a binomial distribution.

### Details

This function calculates the Akaike Information Criterion (AIC) for a binomial distribution fitted to the provided data.

This function fits a binomial distribution to the provided data. It estimates the size and probability parameters of the binomial distribution from the data. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the size and probability parameters of the binomial distribution.

Optimization method: Since the parameters are directly calculated from the data, no optimization is needed.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

### Value

The AIC value calculated based on the fitted binomial distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rbinom(30, size = 10, prob = 0.2)
util_binomial_aic(x)
```

---

```
util_binomial_param_estimate
```

```
Estimate Binomial Parameters
```

---

**Description**

This function will check to see if some given vector `.x` is either a numeric vector or a factor vector with at least two levels then it will cause an error and the function will abort. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated binomial data.

**Usage**

```
util_binomial_param_estimate(.x, .size = NULL, .auto_gen_empirical = TRUE)
```

**Arguments**

<code>.x</code>	The vector of data to be passed to the function. Must be numeric, and all values must be $0 \leq x \leq 1$
<code>.size</code>	Number of trials, zero or more.
<code>.auto_gen_empirical</code>	This is a boolean value of <code>TRUE/FALSE</code> with default set to <code>TRUE</code> . This will automatically create the <code>tidy_empirical()</code> output for the <code>.x</code> parameter and use the <code>tidy_combine_distributions()</code> . The user can then plot out the data using <code>\$combined_data_tbl</code> from the function output.

**Details**

This function will attempt to estimate the binomial  $p_{\text{hat}}$  and size parameters given some vector of values.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Binomial: [tidy\\_binomial\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_negative\\_binomial\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

tb <- rbinom(50, 1, .1)
output <- util_binomial_param_estimate(tb)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

`util_binomial_stats_tbl`*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_binomial_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a tidy\_ distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Binomial: [tidy\\_binomial\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_negative\\_binomial\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

## Examples

```
library(dplyr)

tidy_binomial() |>
  util_binomial_stats_tbl() |>
  glimpse()
```

---

util\_burr\_param\_estimate

*Estimate Burr Parameters*

---

## Description

This function will attempt to estimate the Burr prob parameter given some vector of values `.x`. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated Burr data.

## Usage

```
util_burr_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function. Must be non-negative integers.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will see if the given vector `.x` is a numeric vector. It will attempt to estimate the prob parameter of a Burr distribution.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: `util_bernoulli_param_estimate()`, `util_beta_param_estimate()`, `util_binomial_param_estimate()`, `util_cauchy_param_estimate()`, `util_chisquare_param_estimate()`, `util_exponential_param_estimate()`, `util_f_param_estimate()`, `util_gamma_param_estimate()`, `util_generalized_beta_param_estimate()`, `util_generalized_pareto_param_estimate()`, `util_geometric_param_estimate()`, `util_hypergeometric_param_estimate()`, `util_inverse_burr_param_estimate()`, `util_inverse_pareto_param_estimate()`, `util_inverse_weibull_param_estimate()`, `util_logistic_param_estimate()`, `util_lognormal_param_estimate()`, `util_negative_binomial_param_estimate()`, `util_normal_param_estimate()`, `util_paralogistic_param_estimate()`, `util_pareto1_param_estimate()`, `util_pareto_param_estimate()`, `util_poisson_param_estimate()`, `util_t_param_estimate()`, `util_triangular_param_estimate()`, `util_uniform_param_estimate()`, `util_weibull_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_geometric_param_estimate()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_poisson_param_estimate()`

Other Burr: `tidy_burr()`, `tidy_inverse_burr()`, `util_burr_stats_tbl()`

**Examples**

```
library(dplyr)
library(ggplot2)

tb <- tidy_burr(.shape1 = 1, .shape2 = 2, .rate = .3) |> pull(y)
output <- util_burr_param_estimate(tb)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util\_burr\_stats\_tbl    *Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_burr_stats_tbl(.data)
```

**Arguments**

`.data`                    The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Burr: [tidy\\_burr\(\)](#), [tidy\\_inverse\\_burr\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_burr() |>
  util_burr_stats_tbl() |>
  glimpse()
```

---

util\_cauchy\_aic

*Calculate Akaike Information Criterion (AIC) for Cauchy Distribution*

---

**Description**

This function estimates the parameters of a Cauchy distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_cauchy_aic(.x)
```

**Arguments**

.x                    A numeric vector containing the data to be fitted to a Cauchy distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a Cauchy distribution fitted to the provided data.

This function fits a Cauchy distribution to the provided data using maximum likelihood estimation. It first estimates the initial parameters of the Cauchy distribution using the method of moments. Then, it optimizes the negative log-likelihood function using the provided data and the initial parameter estimates. Finally, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates for the initial location and scale parameters of the Cauchy distribution.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted Cauchy distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rcauchy(30)
util_cauchy_aic(x)
```



---

`util_cauchy_param_estimate`*Estimate Cauchy Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated cauchy data.

## Usage

```
util_cauchy_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the cauchy location and scale parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#),

```
util_uniform_param_estimate(), util_weibull_param_estimate(), util_zero_truncated_binomial_param_estimate(),  
util_zero_truncated_geometric_param_estimate(), util_zero_truncated_negative_binomial_param_estimate(),  
util_zero_truncated_poisson_param_estimate()
```

Other Cauchy: `tidy_cauchy()`, `util_cauchy_stats_tbl()`

## Examples

```
library(dplyr)  
library(ggplot2)  
  
x <- tidy_cauchy(.location = 0, .scale = 1)$y  
output <- util_cauchy_param_estimate(x)  
  
output$parameter_tbl  
  
output$combined_data_tbl |>  
  tidy_combinedautoplot()
```

---

util\_cauchy\_stats\_tbl *Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_cauchy_stats_tbl(.data)
```

## Arguments

`.data` The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Cauchy: `tidy_cauchy()`, `util_cauchy_param_estimate()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_cauchy() |>
  util_cauchy_stats_tbl() |>
  glimpse()
```

---

```
util_chisquare_param_estimate
      Estimate Chisquare Parameters
```

---

**Description**

This function will attempt to estimate the Chisquare prob parameter given some vector of values `.x`. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated Chisquare data.

**Usage**

```
util_chisquare_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function. Must be non-negative integers.

`.auto_gen_empirical` This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will see if the given vector `.x` is a numeric vector. It will attempt to estimate the prob parameter of a Chisquare distribution. The function first performs `tidyeval` on the input data to ensure it's a numeric vector. It then checks if there are at least two data points, as this is a requirement for parameter estimation.

The estimation of the chi-square distribution parameters is performed using maximum likelihood estimation (MLE) implemented with the `bbmle` package. The negative log-likelihood function is minimized to obtain the estimates for the degrees of freedom (`doff`) and the non-centrality parameter (`ncp`). Initial values for the optimization are set based on the sample variance and mean, but these can be adjusted if necessary.

If the estimation fails or encounters an error, the function returns NA for both `doff` and `ncp`.

Finally, the function returns a tibble containing the following information:

**dist\_type** The type of distribution, which is "Chisquare" in this case.

**samp\_size** The sample size, i.e., the number of data points in the input vector.

**min** The minimum value of the data points.

**max** The maximum value of the data points.

**mean** The mean of the data points.

**degrees\_of\_freedom** The estimated degrees of freedom (`doff`) for the chi-square distribution.

**ncp** The estimated non-centrality parameter (`ncp`) for the chi-square distribution.

Additionally, if the argument `.auto_gen_empirical` is set to TRUE (which is the default behavior), the function also returns a combined tibble containing both empirical and chi-square distribution data, obtained by calling `tidy_empirical` and `tidy_chisquare`, respectively.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Chisquare: [tidy\\_chisquare\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

tc <- tidy_chisquare(.n = 500, .df = 6, .ncp = 1) |> pull(y)
output <- util_chisquare_param_estimate(tc)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util\_chisquare\_stats\_tbl

*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_chisquare_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Chisquare: `tidy_chisquare()`, `util_chisquare_param_estimate()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_chisquare() |>
  util_chisquare_stats_tbl() |>
  glimpse()
```

---

<code>util_chisq_aic</code>	<i>Calculate Akaike Information Criterion (AIC) for Chi-Square Distribution</i>
-----------------------------	---

---

**Description**

This function estimates the parameters of a chi-square distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_chisq_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a chi-square distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a chi-square distribution fitted to the provided data.

**Value**

The AIC value calculated based on the fitted chi-square distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rchisq(30, df = 3)
util_chisq_aic(x)
```

---

`util_exponential_aic` *Calculate Akaike Information Criterion (AIC) for Exponential Distribution*

---

**Description**

This function estimates the rate parameter of an exponential distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_exponential_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to an exponential distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for an exponential distribution fitted to the provided data.

This function fits an exponential distribution to the provided data using maximum likelihood estimation. It estimates the rate parameter of the exponential distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the reciprocal of the mean of the data as the initial estimate for the rate parameter.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

### Value

The AIC value calculated based on the fitted exponential distribution to the provided data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

### Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rexp(30)
util_exponential_aic(x)
```

---

util\_exponential\_param\_estimate

*Estimate Exponential Parameters*

---

### Description

This function will attempt to estimate the exponential rate parameter given some vector of values. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated exponential data.



**Usage**

```
util_exponential_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function. Must be numeric.

`.auto_gen_empirical`  
This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will see if the given vector `.x` is a numeric vector.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Exponential: [tidy\\_exponential\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

te <- tidy_exponential(.rate = .1) |> pull(y)
output <- util_exponential_param_estimate(te)

output$parameter_tbl
```

```
output$combined_data_tbl |>
  tidy_combined_autoplot()
```

---

```
util_exponential_stats_tbl
  Distribution Statistics
```

---

### Description

Returns distribution statistics in a tibble.

### Usage

```
util_exponential_stats_tbl(.data)
```

### Arguments

`.data`            The data being passed from a `tidy_` distribution function.

### Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

### Value

A tibble

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Exponential: [tidy\\_exponential\(\)](#), [tidy\\_inverse\\_exponential\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_exponential() |>
  util_exponential_stats_tbl() |>
  glimpse()
```

---

`util_f_aic`*Calculate Akaike Information Criterion (AIC) for F Distribution*

---

**Description**

This function estimates the parameters of a F distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_f_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to an F distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for an F distribution fitted to the provided data.

This function fits an F distribution to the input data using maximum likelihood estimation and then computes the Akaike Information Criterion (AIC) based on the fitted distribution.

**Value**

The AIC value calculated based on the fitted F distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

[rf](#) for generating F-distributed data, [optim](#) for optimization.

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#),

```
util_paralogistic_aic(), util_pareto1_aic(), util_pareto_aic(), util_poisson_aic(),  
util_t_aic(), util_triangular_aic(), util_uniform_aic(), util_weibull_aic(), util_zero_truncated_binomial_aic(),  
util_zero_truncated_geometric_aic(), util_zero_truncated_negative_binomial_aic(),  
util_zero_truncated_poisson_aic()
```

## Examples

```
# Generate F-distributed data  
set.seed(123)  
x <- rf(100, df1 = 5, df2 = 10, ncp = 1)  
  
# Calculate AIC for the generated data  
util_f_aic(x)
```

---

util\_f\_param\_estimate *Estimate F Distribution Parameters*

---

## Description

Estimate F Distribution Parameters

## Usage

```
util_f_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function, where the data comes from the `rf()` function.

`.auto_gen_empirical` This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the F distribution parameters given some vector of values produced by `rf()`. The estimation method is from the NIST Engineering Statistics Handbook.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: `util_bernoulli_param_estimate()`, `util_beta_param_estimate()`, `util_binomial_param_estimate()`, `util_burr_param_estimate()`, `util_cauchy_param_estimate()`, `util_chisquare_param_estimate()`, `util_exponential_param_estimate()`, `util_gamma_param_estimate()`, `util_generalized_beta_param_estimate()`, `util_generalized_pareto_param_estimate()`, `util_geometric_param_estimate()`, `util_hypergeometric_param_estimate()`, `util_inverse_burr_param_estimate()`, `util_inverse_pareto_param_estimate()`, `util_inverse_weibull_param_estimate()`, `util_logistic_param_estimate()`, `util_lognormal_param_estimate()`, `util_negative_binomial_param_estimate()`, `util_normal_param_estimate()`, `util_paralogistic_param_estimate()`, `util_pareto1_param_estimate()`, `util_pareto_param_estimate()`, `util_poisson_param_estimate()`, `util_t_param_estimate()`, `util_triangular_param_estimate()`, `util_uniform_param_estimate()`, `util_weibull_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_geometric_param_estimate()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_poisson_param_estimate()`

Other F Distribution: `tidy_f()`, `util_f_stats_tbl()`

**Examples**

```
library(dplyr)
library(ggplot2)

set.seed(123)
x <- rf(100, df1 = 5, df2 = 10, ncp = 1)
output <- util_f_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util_f_stats_tbl	<i>Distribution Statistics</i>
------------------	--------------------------------

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_f_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other F Distribution: `tidy_f()`, `util_f_param_estimate()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_f() |>
  util_f_stats_tbl() |>
  glimpse()
```

---

util\_gamma\_aic

*Calculate Akaike Information Criterion (AIC) for Gamma Distribution*

---

**Description**

This function estimates the shape and scale parameters of a gamma distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_gamma_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a gamma distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a gamma distribution fitted to the provided data.

This function fits a gamma distribution to the provided data using maximum likelihood estimation. It estimates the shape and scale parameters of the gamma distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape and scale parameters of the gamma distribution.

Optimization method: The function uses the optim function for optimization. You might explore different optimization methods within optim for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted gamma distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rgamma(30, shape = 1)
util_gamma_aic(x)
```

---

`util_gamma_param_estimate`*Estimate Gamma Parameters*

---

### Description

This function will attempt to estimate the gamma shape and scale parameters given some vector of values. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated gamma data.

### Usage

```
util_gamma_param_estimate(.x, .auto_gen_empirical = TRUE)
```

### Arguments

`.x` The vector of data to be passed to the function. Must be numeric.

`.auto_gen_empirical`

This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

### Details

This function will see if the given vector `.x` is a numeric vector.

### Value

A tibble/list

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#)



```
util_zero_truncated_geometric_param_estimate(), util_zero_truncated_negative_binomial_param_estimate  
util_zero_truncated_poisson_param_estimate()
```

Other Gamma: `tidy_gamma()`, `tidy_inverse_gamma()`, `util_gamma_stats_tbl()`

## Examples

```
library(dplyr)  
library(ggplot2)  
  
tg <- tidy_gamma(.shape = 1, .scale = .3) |> pull(y)  
output <- util_gamma_param_estimate(tg)  
  
output$parameter_tbl  
  
output$combined_data_tbl |>  
  tidy_combinedautoplot()
```

---

util\_gamma\_stats\_tbl *Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_gamma_stats_tbl(.data)
```

## Arguments

`.data`            The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Gamma: `tidy_gamma()`, `tidy_inverse_gamma()`, `util_gamma_param_estimate()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_gamma() |>
  util_gamma_stats_tbl() |>
  glimpse()
```

---

```
util_generalized_beta_aic
```

*Calculate Akaike Information Criterion (AIC) for Generalized Beta Distribution*

---

**Description**

This function estimates the `shape1`, `shape2`, `shape3`, and `rate` parameters of a generalized Beta distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_generalized_beta_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a generalized Beta distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a generalized Beta distribution fitted to the provided data.

This function fits a generalized Beta distribution to the provided data using maximum likelihood estimation. It estimates the shape1, shape2, shape3, and rate parameters of the generalized Beta distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses reasonable initial estimates for the shape1, shape2, shape3, and rate parameters of the generalized Beta distribution.

Optimization method: The function uses the optim function for optimization. You might explore different optimization methods within optim for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted generalized Beta distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- tidy_generalized_beta(100, .shape1 = 2, .shape2 = 3,
                           .shape3 = 4, .rate = 5)[["y"]]
util_generalized_beta_aic(x)
```

---

 util\_generalized\_beta\_param\_estimate

*Estimate Generalized Beta Parameters*


---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated generalized Beta data.

## Usage

```
util_generalized_beta_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the generalized Beta `shape1`, `shape2`, `shape3`, and `rate` parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#),

```
util_weibull_param_estimate(), util_zero_truncated_binomial_param_estimate(), util_zero_truncated_geom  
util_zero_truncated_negative_binomial_param_estimate(), util_zero_truncated_poisson_param_estimate()  
Other Generalized Beta: util_generalized_beta_stats_tbl()
```

## Examples

```
library(dplyr)  
library(ggplot2)  
  
set.seed(123)  
x <- tidy_generalized_beta(100, .shape1 = 2, .shape2 = 3,  
  .shape3 = 4, .rate = 5)[["y"]]  
output <- util_generalized_beta_param_estimate(x)  
  
output$parameter_tbl  
  
output$combined_data_tbl %>%  
  tidy_combinedautoplot()
```

---

util\_generalized\_beta\_stats\_tbl  
*Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_generalized_beta_stats_tbl(.data)
```

## Arguments

`.data`            The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and return the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Generalized Beta: `util_generalized_beta_param_estimate()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

set.seed(123)
tidy_generalized_beta() |>
  util_generalized_beta_stats_tbl() |>
  glimpse()
```

---

```
util_generalized_pareto_aic
```

*Calculate Akaike Information Criterion (AIC) for Generalized Pareto Distribution*

---

**Description**

This function estimates the shape1, shape2, and rate parameters of a generalized Pareto distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_generalized_pareto_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a generalized Pareto distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a generalized Pareto distribution fitted to the provided data.

This function fits a generalized Pareto distribution to the provided data using maximum likelihood estimation. It estimates the shape1, shape2, and rate parameters of the generalized Pareto distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape1, shape2, and rate parameters of the generalized Pareto distribution.

Optimization method: The function uses the optim function for optimization. You might explore different optimization methods within optim for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted generalized Pareto distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- actuar::rgenpareto(100, shape1 = 1, shape2 = 2, scale = 3)
util_generalized_pareto_aic(x)
```

---

 util\_generalized\_pareto\_param\_estimate

*Estimate Generalized Pareto Parameters*


---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated generalized Pareto data.

## Usage

```
util_generalized_pareto_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the generalized Pareto `shape1`, `shape2`, and `rate` parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#),



```
util_weibull_param_estimate(), util_zero_truncated_binomial_param_estimate(), util_zero_truncated_geom  
util_zero_truncated_negative_binomial_param_estimate(), util_zero_truncated_poisson_param_estimate()
```

Other Generalized Pareto: `util_generalized_pareto_stats_tbl()`

## Examples

```
library(dplyr)  
library(ggplot2)  
  
set.seed(123)  
x <- tidy_generalized_pareto(100, .shape1 = 1, .shape2 = 2, .scale = 3)[["y"]]  
output <- util_generalized_pareto_param_estimate(x)  
  
output$parameter_tbl  
  
output$combined_data_tbl %>%  
  tidy_combinedautoplot()
```

---

util\_generalized\_pareto\_stats\_tbl  
*Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_generalized_pareto_stats_tbl(.data)
```

## Arguments

`.data` The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and return the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Generalized Pareto: [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_generalized_pareto() |>
  util_generalized_pareto_stats_tbl() |>
  glimpse()
```

---

util_geometric_aic	<i>Calculate Akaike Information Criterion (AIC) for Geometric Distribution</i>
--------------------	--

---

**Description**

This function estimates the probability parameter of a geometric distribution from the provided data and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_geometric_aic(.x)
```

**Arguments**

.x                    A numeric vector containing the data to be fitted to a geometric distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a geometric distribution fitted to the provided data.

This function fits a geometric distribution to the provided data. It estimates the probability parameter of the geometric distribution from the data. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimate as a starting point for the probability parameter of the geometric distribution.

Optimization method: Since the parameter is directly calculated from the data, no optimization is needed.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

### Value

The AIC value calculated based on the fitted geometric distribution to the provided data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

### Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rgeom(100, prob = 0.2)
util_geometric_aic(x)
```

---

util\_geometric\_param\_estimate

*Estimate Geometric Parameters*

---

### Description

This function will attempt to estimate the geometric prob parameter given some vector of values `.x`. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated geometric data.

**Usage**

```
util_geometric_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function. Must be non-negative integers.

`.auto_gen_empirical` This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will see if the given vector `.x` is a numeric vector. It will attempt to estimate the prob parameter of a geometric distribution.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_beta\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Geometric: [tidy\\_geometric\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

tg <- tidy_geometric(.prob = .1) |> pull(y)
output <- util_geometric_param_estimate(tg)

output$parameter_tbl
```

```
output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util\_geometric\_stats\_tbl  
*Distribution Statistics*

---

### Description

Returns distribution statistics in a tibble.

### Usage

```
util_geometric_stats_tbl(.data)
```

### Arguments

`.data`            The data being passed from a `tidy_` distribution function.

### Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

### Value

A tibble

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Geometric: [tidy\\_geometric\(\)](#), [tidy\\_zero\\_truncated\\_geometric\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

## Examples

```
library(dplyr)

tidy_geometric() |>
  util_geometric_stats_tbl() |>
  glimpse()
```

---

util\_hypergeometric\_aic

*Calculate Akaike Information Criterion (AIC) for Hypergeometric Distribution*

---

## Description

This function estimates the parameters  $m$ ,  $n$ , and  $k$  of a hypergeometric distribution from the provided data and then calculates the AIC value based on the fitted distribution.

## Usage

```
util_hypergeometric_aic(.x)
```

## Arguments

`.x` A numeric vector containing the data to be fitted to a hypergeometric distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a hypergeometric distribution fitted to the provided data.

This function fits a hypergeometric distribution to the provided data. It estimates the parameters  $m$ ,  $n$ , and  $k$  of the hypergeometric distribution from the data. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function does not estimate parameters; they are directly calculated from the data.

Optimization method: Since the parameters are directly calculated from the data, no optimization is needed.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted hypergeometric distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rhyper(100, m = 10, n = 10, k = 5)
util_hypergeometric_aic(x)
```

---

util\_hypergeometric\_param\_estimate

*Estimate Hypergeometric Parameters*

---

**Description**

This function will attempt to estimate the geometric prob parameter given some vector of values `.x`. Estimate `m`, the number of white balls in the urn, or `m+n`, the total number of balls in the urn, for a hypergeometric distribution.

**Usage**

```
util_hypergeometric_param_estimate(
  .x,
  .m = NULL,
  .total = NULL,
  .k,
  .auto_gen_empirical = TRUE
)
```

**Arguments**

<code>.x</code>	A non-negative integer indicating the number of white balls out of a sample of size <code>.k</code> drawn without replacement from the urn. You cannot have missing, undefined or infinite values.
<code>.m</code>	Non-negative integer indicating the number of white balls in the urn. You must supply <code>.m</code> or <code>.total</code> , but not both. You cannot have missing values.
<code>.total</code>	A positive integer indicating the total number of balls in the urn (i.e., $m+n$ ). You must supply <code>.m</code> or <code>.total</code> , but not both. You cannot have missing values.
<code>.k</code>	A positive integer indicating the number of balls drawn without replacement from the urn. You cannot have missing values.
<code>.auto_gen_empirical</code>	This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the <code>tidy_empirical()</code> output for the <code>.x</code> parameter and use the <code>tidy_combine_distributions()</code> . The user can then plot out the data using <code>\$combined_data_tbl</code> from the function output.

**Details**

This function will see if the given vector `.x` is a numeric integer. It will attempt to estimate the prob parameter of a geometric distribution. Missing (NA), undefined (NaN), and infinite (Inf, -Inf) values are not allowed. Let `.x` be an observation from a hypergeometric distribution with parameters `.m` = M, `.n` = N, and `.k` = K. In R nomenclature, `.x` represents the number of white balls drawn out of a sample of `.k` balls drawn without replacement from an urn containing `.m` white balls and `.n` black balls. The total number of balls in the urn is thus `.m` + `.n`. Denote the total number of balls by  $T = .m + .n$

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Hypergeometric: [tidy\\_hypergeometric\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#)



**Examples**

```
library(dplyr)
library(ggplot2)

th <- rhyper(10, 20, 30, 5)
output <- util_hypergeometric_param_estimate(th, .total = 50, .k = 5)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util\_hypergeometric\_stats\_tbl  
*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_hypergeometric_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Hypergeometric: `tidy_hypergeometric()`, `util_hypergeometric_param_estimate()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_hypergeometric() |>
  util_hypergeometric_stats_tbl() |>
  glimpse()
```

---

`util_inverse_burr_aic` *Calculate Akaike Information Criterion (AIC) for Inverse Burr Distribution*

---

**Description**

This function estimates the `shape1`, `shape2`, and `rate` parameters of an inverse Burr distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_inverse_burr_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to an inverse Burr distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for an inverse Burr distribution fitted to the provided data.

This function fits an inverse Burr distribution to the provided data using maximum likelihood estimation. It estimates the `shape1`, `shape2`, and `rate` parameters of the inverse Burr distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape1, shape2, and rate parameters of the inverse Burr distribution.

Optimization method: The function uses the optim function for optimization. You might explore different optimization methods within optim for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

### Value

The AIC value calculated based on the fitted inverse Burr distribution to the provided data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

### Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- tidy_inverse_burr(100, .shape1 = 2, .shape2 = 3, .scale = 1)[["y"]]
util_inverse_burr_aic(x)
```

---

util\_inverse\_burr\_param\_estimate

*Estimate Inverse Burr Parameters*

---

### Description

This function will attempt to estimate the inverse Burr shape1, shape2, and rate parameters given some vector of values `.x`. The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated inverse Burr data.

**Usage**

```
util_inverse_burr_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function. Must be non-negative integers.

`.auto_gen_empirical` This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will see if the given vector `.x` is a numeric vector. It will attempt to estimate the `shape1`, `shape2`, and `rate` parameters of an inverse Burr distribution.

**Value**

A tibble/list

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Inverse Burr: [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

set.seed(123)
tb <- tidy_burr(.shape1 = 1, .shape2 = 2, .rate = .3) |> pull(y)
output <- util_inverse_burr_param_estimate(tb)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

`util_inverse_burr_stats_tbl`*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_inverse_burr_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a tidy\_ distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Inverse Burr: [util\\_inverse\\_burr\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

## Examples

```
library(dplyr)

set.seed(123)
tidy_inverse_burr() |>
  util_inverse_burr_stats_tbl() |>
  glimpse()
```

---

util\_inverse\_pareto\_aic

*Calculate Akaike Information Criterion (AIC) for Inverse Pareto Distribution*

---

## Description

This function estimates the shape and scale parameters of an inverse Pareto distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

## Usage

```
util_inverse_pareto_aic(.x)
```

## Arguments

`.x` A numeric vector containing the data to be fitted to an inverse Pareto distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for an inverse Pareto distribution fitted to the provided data.

This function fits an inverse Pareto distribution to the provided data using maximum likelihood estimation. It estimates the shape and scale parameters of the inverse Pareto distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape and scale parameters of the inverse Pareto distribution.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted inverse Pareto distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- tidy_inverse_pareto(.n = 100, .shape = 2, .scale = 1)[["y"]]
util_inverse_pareto_aic(x)
```

---

util\_inverse\_pareto\_param\_estimate

*Estimate Inverse Pareto Parameters*

---

**Description**

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated inverse Pareto data.

**Usage**

```
util_inverse_pareto_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will attempt to estimate the inverse Pareto shape and scale parameters given some vector of values.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Inverse Pareto: [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

set.seed(123)
x <- tidy_inverse_pareto(.n = 100, .shape = 2, .scale = 1)[["y"]]
output <- util_inverse_pareto_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl %>%
  tidy_combined_autoplot()
```



**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_inverse_pareto_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a tidy\_ distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Inverse Pareto: [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_inverse_pareto() |>
  util_inverse_pareto_stats_tbl() |>
  glimpse()
```

---

`util_inverse_weibull_aic`

*Calculate Akaike Information Criterion (AIC) for Inverse Weibull Distribution*

---

### Description

This function estimates the shape and scale parameters of an inverse Weibull distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

### Usage

```
util_inverse_weibull_aic(.x)
```

### Arguments

`.x` A numeric vector containing the data to be fitted to an inverse Weibull distribution.

### Details

This function calculates the Akaike Information Criterion (AIC) for an inverse Weibull distribution fitted to the provided data.

This function fits an inverse Weibull distribution to the provided data using maximum likelihood estimation. It estimates the shape and scale parameters of the inverse Weibull distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape and scale parameters of the inverse Weibull distribution.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

### Value

The AIC value calculated based on the fitted inverse Weibull distribution to the provided data.

### Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- tidy_inverse_weibull(.n = 100, .shape = 2, .scale = 1)[["y"]]
util_inverse_weibull_aic(x)
```

---

util\_inverse\_weibull\_param\_estimate

*Estimate Inverse Weibull Parameters*

---

**Description**

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated inverse Weibull data.

**Usage**

```
util_inverse_weibull_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will attempt to estimate the inverse Weibull shape and rate parameters given some vector of values.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: `util_bernoulli_param_estimate()`, `util_beta_param_estimate()`, `util_binomial_param_estimate()`, `util_burr_param_estimate()`, `util_cauchy_param_estimate()`, `util_chisquare_param_estimate()`, `util_exponential_param_estimate()`, `util_f_param_estimate()`, `util_gamma_param_estimate()`, `util_generalized_beta_param_estimate()`, `util_generalized_pareto_param_estimate()`, `util_geometric_param_estimate()`, `util_hypergeometric_param_estimate()`, `util_inverse_burr_param_estimate()`, `util_inverse_pareto_param_estimate()`, `util_logistic_param_estimate()`, `util_lognormal_param_estimate()`, `util_negative_binomial_param_estimate()`, `util_normal_param_estimate()`, `util_paralogistic_param_estimate()`, `util_pareto1_param_estimate()`, `util_pareto_param_estimate()`, `util_poisson_param_estimate()`, `util_t_param_estimate()`, `util_triangular_param_estimate()`, `util_uniform_param_estimate()`, `util_weibull_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_geometric_param_estimate()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_poisson_param_estimate()`

Other Inverse Weibull: `util_inverse_weibull_stats_tbl()`

**Examples**

```
library(dplyr)
library(ggplot2)

set.seed(123)
x <- tidy_inverse_weibull(100, .shape = 2, .scale = 1)[["y"]]
output <- util_inverse_weibull_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl %>%
  tidy_combined_autoplot()
```

---

util\_inverse\_weibull\_stats\_tbl  
*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_inverse_weibull_stats_tbl(.data)
```

**Arguments**

`.data` The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Inverse Weibull: [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

set.seed(123)
tidy_inverse_weibull() |>
  util_inverse_weibull_stats_tbl() |>
  glimpse()
```

---

<code>util_logistic_aic</code>	<i>Calculate Akaike Information Criterion (AIC) for Logistic Distribution</i>
--------------------------------	---

---

**Description**

This function estimates the location and scale parameters of a logistic distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_logistic_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a logistic distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a logistic distribution fitted to the provided data.

This function fits a logistic distribution to the provided data using maximum likelihood estimation. It estimates the location and scale parameters of the logistic distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the location and scale parameters of the logistic distribution.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

**Value**

The AIC value calculated based on the fitted logistic distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rlogis(30)
util_logistic_aic(x)
```

---

`util_logistic_param_estimate`*Estimate Logistic Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated logistic data.

Three different methods of shape parameters are supplied:

- MLE
- MME
- MMUE

## Usage

```
util_logistic_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical`

This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the logistic location and scale parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: `util_bernoulli_param_estimate()`, `util_beta_param_estimate()`, `util_binomial_param_estimate()`, `util_burr_param_estimate()`, `util_cauchy_param_estimate()`, `util_chisquare_param_estimate()`, `util_exponential_param_estimate()`, `util_f_param_estimate()`, `util_gamma_param_estimate()`, `util_generalized_beta_param_estimate()`, `util_generalized_pareto_param_estimate()`, `util_geometric_param_estimate()`, `util_hypergeometric_param_estimate()`, `util_inverse_burr_param_estimate()`, `util_inverse_pareto_param_estimate()`, `util_inverse_weibull_param_estimate()`, `util_lognormal_param_estimate()`, `util_negative_binomial_param_estimate()`, `util_normal_param_estimate()`, `util_paralogistic_param_estimate()`, `util_pareto1_param_estimate()`, `util_pareto_param_estimate()`, `util_poisson_param_estimate()`, `util_t_param_estimate()`, `util_triangular_param_estimate()`, `util_uniform_param_estimate()`, `util_weibull_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_geometric_param_estimate()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_poisson_param_estimate()`

Other Logistic: `tidy_logistic()`, `tidy_paralogistic()`, `util_logistic_stats_tbl()`

**Examples**

```
library(dplyr)
library(ggplot2)

x <- mtcars$mpg
output <- util_logistic_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

t <- rlogis(50, 2.5, 1.4)
util_logistic_param_estimate(t)$parameter_tbl
```

---

```
util_logistic_stats_tbl
```

*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_logistic_stats_tbl(.data)
```

**Arguments**

`.data` The data being passed from a `tidy_` distribution function.



**Details**

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Logistic: [tidy\\_logistic\(\)](#), [tidy\\_paralogistic\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_logistic() |>
  util_logistic_stats_tbl() |>
  glimpse()
```

---

util_lognormal_aic	<i>Calculate Akaike Information Criterion (AIC) for Log-Normal Distribution</i>
--------------------	---

---

**Description**

This function estimates the meanlog and sdlog parameters of a log-normal distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_lognormal_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a log-normal distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a log-normal distribution fitted to the provided data.

This function fits a log-normal distribution to the provided data using maximum likelihood estimation. It estimates the `meanlog` and `sdlog` parameters of the log-normal distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the `meanlog` and `sdlog` parameters of the log-normal distribution.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

**Value**

The AIC value calculated based on the fitted log-normal distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rlnorm(100, meanlog = 0, sdlog = 1)
util_lognormal_aic(x)
```

---

util\_lognormal\_param\_estimate  
*Estimate Lognormal Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated lognormal data.

Three different methods of shape parameters are supplied:

- `mme`, see [EnvStats::elnorm\(\)](#)
- `mle`, see [EnvStats::elnorm\(\)](#)

## Usage

```
util_lognormal_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the lognormal meanlog and log sd parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#)

[util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Lognormal: [tidy\\_lognormal\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#)

## Examples

```

library(dplyr)
library(ggplot2)

x <- mtcars$mpg
output <- util_lognormal_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

tb <- tidy_lognormal(.meanlog = 2, .sdlog = 1) |> pull(y)
util_lognormal_param_estimate(tb)$parameter_tbl

```

---

```

util_lognormal_stats_tbl
      Distribution Statistics

```

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_lognormal_stats_tbl(.data)
```

## Arguments

`.data`            The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Lognormal: [tidy\\_lognormal\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_lognormal() |>
  util_lognormal_stats_tbl() |>
  glimpse()
```

---

util\_negative\_binomial\_aic

*Calculate Akaike Information Criterion (AIC) for Negative Binomial Distribution*

---

**Description**

This function estimates the parameters size ( $r$ ) and probability ( $prob$ ) of a negative binomial distribution from the provided data and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_negative_binomial_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a negative binomial distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a negative binomial distribution fitted to the provided data.

This function fits a negative binomial distribution to the provided data. It estimates the parameters size ( $r$ ) and probability ( $prob$ ) of the negative binomial distribution from the data. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimate as a starting point for the size ( $r$ ) parameter of the negative binomial distribution, and the probability ( $prob$ ) is estimated based on the mean and variance of the data.

Optimization method: Since the parameters are directly calculated from the data, no optimization is needed.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted negative binomial distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
data <- rnbinom(n = 100, size = 5, mu = 10)
util_negative_binomial_aic(data)
```

---

`util_negative_binomial_param_estimate`*Estimate Negative Binomial Parameters*

---

### Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated negative binomial data.

Three different methods of shape parameters are supplied:

- MLE/MME
- MMUE
- MLE via `optim` function.

### Usage

```
util_negative_binomial_param_estimate(  
  .x,  
  .size = 1,  
  .auto_gen_empirical = TRUE  
)
```

### Arguments

`.x` The vector of data to be passed to the function.

`.size` The size parameter, the default is 1.

`.auto_gen_empirical`

This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

### Details

This function will attempt to estimate the negative binomial size and prob parameters given some vector of values.

### Value

A tibble/list

### Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Binomial: [tidy\\_binomial\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_negative\\_binomial\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

x <- as.integer(mtcars$mpg)
output <- util_negative_binomial_param_estimate(x, .size = 1)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

t <- rnbinom(50, 1, .1)
util_negative_binomial_param_estimate(t, .size = 1)$parameter_tbl
```

---

```
util_negative_binomial_stats_tbl
      Distribution Statistics
```

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_negative_binomial_stats_tbl(.data)
```

**Arguments**

`.data` The data being passed from a `tidy_` distribution function.



**Details**

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Negative Binomial: [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_negative_binomial() |>
  util_negative_binomial_stats_tbl() |>
  glimpse()
```

---

util\_normal\_aic

*Calculate Akaike Information Criterion (AIC) for Normal Distribution*

---

**Description**

This function estimates the parameters of a normal distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_normal_aic(.x)
```

## Arguments

`.x` A numeric vector containing the data to be fitted to a normal distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a normal distribution fitted to the provided data.

## Value

The AIC value calculated based on the fitted normal distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
data <- rnorm(30)
util_normal_aic(data)
```

---

util\_normal\_param\_estimate

*Estimate Normal Gaussian Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated normal data.

Three different methods of shape parameters are supplied:

- MLE/MME
- MVUE

**Usage**

```
util_normal_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical`  
This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will attempt to estimate the normal gaussian mean and standard deviation parameters given some vector of values.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Gaussian: [tidy\\_inverse\\_normal\(\)](#), [tidy\\_normal\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

x <- mtcars$mpg
output <- util_normal_param_estimate(x)

output$parameter_tbl
```

```
output$combined_data_tbl |>
  tidy_combinedautoplot()

t <- rnorm(50, 0, 1)
util_normal_param_estimate(t)$parameter_tbl
```

---

util\_normal\_stats\_tbl *Distribution Statistics*

---

### Description

Returns distribution statistics in a tibble.

### Usage

```
util_normal_stats_tbl(.data)
```

### Arguments

`.data`            The data being passed from a `tidy_` distribution function.

### Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

### Value

A tibble

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Gaussian: [tidy\\_inverse\\_normal\(\)](#), [tidy\\_normal\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_normal() |>
  util_normal_stats_tbl() |>
  glimpse()
```

---

util\_paralogistic\_aic *Calculate Akaike Information Criterion (AIC) for Paralogistic Distribution*

---

**Description**

This function estimates the shape and rate parameters of a paralogistic distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_paralogistic_aic(.x)
```

**Arguments**

.x                    A numeric vector containing the data to be fitted to a paralogistic distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a paralogistic distribution fitted to the provided data.

This function fits a paralogistic distribution to the provided data using maximum likelihood estimation. It estimates the shape and rate parameters of the paralogistic distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape and rate parameters of the paralogistic distribution.

Optimization method: The function uses the optim function for optimization. You might explore different optimization methods within optim for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

**Value**

The AIC value calculated based on the fitted paralogistic distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

Other Paralogistic: `util_paralogistic_param_estimate()`, `util_paralogistic_stats_tbl()`

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- tidy_paralogistic(30, .shape = 2, .rate = 1)[["y"]]
util_paralogistic_aic(x)
```

---

util\_paralogistic\_param\_estimate

*Estimate Paralogistic Parameters*

---

**Description**

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated paralogistic data.

The method of parameter estimation is:

- MLE

**Usage**

```
util_paralogistic_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical`

This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will attempt to estimate the paralogistic shape and rate parameters given some vector of values.

**Value**

A tibble/list

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Paralogistic: [util\\_paralogistic\\_aic\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

x <- mtcars$mpg
output <- util_paralogistic_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

t <- tidy_paralogistic(50, 2.5, 1.4)[["y"]]
util_paralogistic_param_estimate(t)$parameter_tbl
```

---

util\_paralogistic\_stats\_tbl

*Distribution Statistics for Paralogistic Distribution*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_paralogistic_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

**Value**

A tibble

**See Also**

Other Paralogistic: [util\\_paralogistic\\_aic\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

set.seed(123)
tidy_paralogistic(.n = 50, .shape = 5, .rate = 6) |>
  util_paralogistic_stats_tbl() |>
  glimpse()
```

---

util\_pareto1\_aic

*Calculate Akaike Information Criterion (AIC) for Pareto Distribution*

---

**Description**

This function estimates the shape and scale parameters of a Pareto distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.



**Usage**

```
util_pareto1_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a Pareto distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a Pareto distribution fitted to the provided data.

This function fits a Pareto distribution to the provided data using maximum likelihood estimation. It estimates the shape and scale parameters of the Pareto distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape and scale parameters of the Pareto distribution.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

**Value**

The AIC value calculated based on the fitted Pareto distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

Other Pareto: [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- tidy_pareto1()$y
util_pareto1_aic(x)
```

---

util\_pareto1\_param\_estimate

*Estimate Pareto Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated Pareto data.

Two different methods of shape parameters are supplied:

- LSE
- MLE

## Usage

```
util_pareto1_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical`

This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the Pareto shape and scale parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Pareto: [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

x <- mtcars[["mpg"]]
output <- util_pareto1_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

set.seed(123)
t <- tidy_pareto1(.n = 100, .shape = 1.5, .min = 1)[["y"]]
util_pareto1_param_estimate(t)$parameter_tbl
```

---

```
util_pareto1_stats_tbl
```

*Distribution Statistics for Pareto1 Distribution*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_pareto1_stats_tbl(.data)
```

**Arguments**

`.data` The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**See Also**

Other Pareto: `tidy_generalized_pareto()`, `tidy_inverse_pareto()`, `tidy_pareto()`, `tidy_pareto1()`, `util_pareto1_aic()`, `util_pareto1_param_estimate()`, `util_pareto_param_estimate()`, `util_pareto_stats_tbl()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_pareto1() |>
  util_pareto1_stats_tbl() |>
  glimpse()
```

---

util\_pareto\_aic

*Calculate Akaike Information Criterion (AIC) for Pareto Distribution*

---

**Description**

This function estimates the shape and scale parameters of a Pareto distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_pareto_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a Pareto distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a Pareto distribution fitted to the provided data.

This function fits a Pareto distribution to the provided data using maximum likelihood estimation. It estimates the shape and scale parameters of the Pareto distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape and scale parameters of the Pareto distribution.

Optimization method: The function uses the optim function for optimization. You might explore different optimization methods within optim for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted Pareto distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- TidyDensity::tidy_pareto()$y
util_pareto_aic(x)
```

---

`util_pareto_param_estimate`*Estimate Pareto Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated pareto data.

Two different methods of shape parameters are supplied:

- LSE
- MLE

## Usage

```
util_pareto_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the pareto shape and scale parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#)

```
util_inverse_pareto_param_estimate(), util_inverse_weibull_param_estimate(), util_logistic_param_esti
util_lognormal_param_estimate(), util_negative_binomial_param_estimate(), util_normal_param_estimate(
util_paralogistic_param_estimate(), util_pareto1_param_estimate(), util_poisson_param_estimate(),
util_t_param_estimate(), util_triangular_param_estimate(), util_uniform_param_estimate(),
util_weibull_param_estimate(), util_zero_truncated_binomial_param_estimate(), util_zero_truncated_geor
util_zero_truncated_negative_binomial_param_estimate(), util_zero_truncated_poisson_param_estimate()
```

```
Other Pareto: tidy_generalized_pareto(), tidy_inverse_pareto(), tidy_pareto(), tidy_pareto1(),
util_pareto1_aic(), util_pareto1_param_estimate(), util_pareto1_stats_tbl(), util_pareto_stats_tbl()
```

## Examples

```
library(dplyr)
library(ggplot2)

x <- mtcars$mpg
output <- util_pareto_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

t <- tidy_pareto(50, 1, 1) |> pull(y)
util_pareto_param_estimate(t)$parameter_tbl
```

---

util\_pareto\_stats\_tbl *Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_pareto_stats_tbl(.data)
```

## Arguments

`.data` The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Pareto: [tidy\\_generalized\\_pareto\(\)](#), [tidy\\_inverse\\_pareto\(\)](#), [tidy\\_pareto\(\)](#), [tidy\\_pareto1\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_pareto() |>
  util_pareto_stats_tbl() |>
  glimpse()
```

---

util_poisson_aic	<i>Calculate Akaike Information Criterion (AIC) for Poisson Distribution</i>
------------------	--

---

**Description**

This function estimates the lambda parameter of a Poisson distribution from the provided data and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_poisson_aic(.x)
```

**Arguments**

.x                    A numeric vector containing the data to be fitted to a Poisson distribution.



## Details

This function calculates the Akaike Information Criterion (AIC) for a Poisson distribution fitted to the provided data.

This function fits a Poisson distribution to the provided data. It estimates the lambda parameter of the Poisson distribution from the data. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimate as a starting point for the lambda parameter of the Poisson distribution.

Optimization method: Since the parameter is directly calculated from the data, no optimization is needed.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted Poisson distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rpois(100, lambda = 2)
util_poisson_aic(x)
```

---

 util\_poisson\_param\_estimate

*Estimate Poisson Parameters*


---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated poisson data.

## Usage

```
util_poisson_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical`

This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the pareto lambda parameter given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Poisson: `tidy_poisson()`, `tidy_zero_truncated_poisson()`, `util_poisson_stats_tbl()`, `util_zero_truncated_poisson_param_estimate()`, `util_zero_truncated_poisson_stats_tbl()`

### Examples

```
library(dplyr)
library(ggplot2)

x <- as.integer(mtcars$mpg)
output <- util_poisson_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combined_autoplot()

t <- rpois(50, 5)
util_poisson_param_estimate(t)$parameter_tbl
```

---

util\_poisson\_stats\_tbl

*Distribution Statistics*

---

### Description

Returns distribution statistics in a tibble.

### Usage

```
util_poisson_stats_tbl(.data)
```

### Arguments

`.data`            The data being passed from a `tidy_` distribution function.

### Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

### Value

A tibble

### Author(s)

Steven P. Sanderson II, MPH

**See Also**

Other Poisson: `tidy_poisson()`, `tidy_zero_truncated_poisson()`, `util_poisson_param_estimate()`, `util_zero_truncated_poisson_param_estimate()`, `util_zero_truncated_poisson_stats_tbl()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

tidy_poisson() |>
  util_poisson_stats_tbl() |>
  glimpse()
```

---

<code>util_triangular_aic</code>	<i>Calculate Akaike Information Criterion (AIC) for Triangular Distribution</i>
----------------------------------	---

---

**Description**

This function estimates the parameters of a triangular distribution (min, max, and mode) from the provided data and calculates the AIC value based on the fitted distribution.

**Usage**

```
util_triangular_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a triangular distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a triangular distribution fitted to the provided data.

The function operates in several steps:

1. **Parameter Estimation:** The function extracts the minimum, maximum, and mode values from the data via the `TidyDensity::util_triangular_param_estimate` function. It returns these initial parameters as the starting point for optimization.
2. **Negative Log-Likelihood Calculation:** A custom function calculates the negative log-likelihood using the `EnvStats::dtri` function to obtain density values for each data point. The densities are logged manually to simulate the behavior of a log parameter.
3. **Parameter Validation:** During optimization, the function checks that the constraints  $\text{min} \leq \text{mode} \leq \text{max}$  are met, and returns an infinite loss if not.
4. **Optimization:** The optimization process utilizes the "SANN" (Simulated Annealing) method to minimize the negative log-likelihood and find optimal parameter values.
5. **AIC Calculation:** The Akaike Information Criterion (AIC) is calculated using the optimized negative log-likelihood and the total number of parameters (3).

### Value

The AIC value calculated based on the fitted triangular distribution to the provided data.

### Author(s)

Steven P. Sanderson II, MPH

### See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

### Examples

```
# Example: Calculate AIC for a sample dataset
set.seed(123)
data <- tidy_triangular(.min = 0, .max = 1, .mode = 1/2)$y
util_triangular_aic(data)
```

---

 util\_triangular\_param\_estimate

*Estimate Triangular Parameters*


---

## Description

This function will attempt to estimate the triangular min, mode, and max parameters given some vector of values.

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated beta data.

## Usage

```
util_triangular_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function. Must be numeric, and all values must be  $0 \leq x \leq 1$

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the triangular min, mode, and max parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#)

```
util_lognormal_param_estimate(), util_negative_binomial_param_estimate(), util_normal_param_estimate(),
util_paralogistic_param_estimate(), util_pareto1_param_estimate(), util_pareto_param_estimate(),
util_poisson_param_estimate(), util_t_param_estimate(), util_uniform_param_estimate(),
util_weibull_param_estimate(), util_zero_truncated_binomial_param_estimate(), util_zero_truncated_geom(),
util_zero_truncated_negative_binomial_param_estimate(), util_zero_truncated_poisson_param_estimate()
```

Other Triangular: `tidy_triangular()`, `util_triangular_stats_tbl()`

## Examples

```
library(dplyr)
library(ggplot2)

x <- mtcars$mpg
output <- util_triangular_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

params <- tidy_triangular()$y |>
  util_triangular_param_estimate()
params$parameter_tbl
```

---

```
util_triangular_stats_tbl
```

*Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_triangular_stats_tbl(.data)
```

## Arguments

`.data`            The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Triangular: [tidy\\_triangular\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_triangular() |>
  util_triangular_stats_tbl() |>
  glimpse()
```

---

 util\_t\_aic

---

*Calculate Akaike Information Criterion (AIC) for t Distribution*


---

**Description**

This function estimates the parameters of a t distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_t_aic(.x)
```

**Arguments**

.x                    A numeric vector containing the data to be fitted to a t distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a t distribution fitted to the provided data.

This function fits a t distribution to the input data using maximum likelihood estimation and then computes the Akaike Information Criterion (AIC) based on the fitted distribution.



**Value**

The AIC value calculated based on the fitted t distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

[rt](#) for generating t-distributed data, [optim](#) for optimization.

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

**Examples**

```
# Generate t-distributed data
set.seed(123)
x <- rt(100, df = 5, ncp = 0.5)

# Calculate AIC for the generated data
util_t_aic(x)
```

---

util\_t\_param\_estimate *Estimate t Distribution Parameters*

---

**Description**

Estimate t Distribution Parameters

**Usage**

```
util_t_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function, where the data comes from the `rt()` function.

`.auto_gen_empirical`

This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the t distribution parameters given some vector of values produced by `rt()`. The estimation method uses both method of moments and maximum likelihood estimation.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

## Examples

```
library(dplyr)
library(ggplot2)

set.seed(123)
x <- rt(100, df = 10, ncp = 0.5)
output <- util_t_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util_t_stats_tbl	<i>Distribution Statistics</i>
------------------	--------------------------------

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_t_stats_tbl(.data)
```

## Arguments

`.data` The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other T Distribution: [tidy\\_t\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_t() |>
  util_t_stats_tbl() |>
  glimpse()
```

---

util_uniform_aic	<i>Calculate Akaike Information Criterion (AIC) for Uniform Distribution</i>
------------------	--

---

**Description**

This function estimates the min and max parameters of a uniform distribution from the provided data and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_uniform_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a uniform distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a uniform distribution fitted to the provided data.

This function fits a uniform distribution to the provided data. It estimates the min and max parameters of the uniform distribution from the range of the data. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the minimum and maximum values of the data as starting points for the min and max parameters of the uniform distribution.

Optimization method: Since the parameters are directly calculated from the data, no optimization is needed.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

**Value**

The AIC value calculated based on the fitted uniform distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

## See Also

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

## Examples

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- runif(30)
util_uniform_aic(x)
```

---

util\_uniform\_param\_estimate

*Estimate Uniform Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated uniform data.

## Usage

```
util_uniform_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the uniform min and max parameters given some vector of values.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Uniform: [tidy\\_uniform\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

x <- tidy_uniform(.min = 1, .max = 3)$y
output <- util_uniform_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util\_uniform\_stats\_tbl

*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_uniform_stats_tbl(.data)
```

**Arguments**

`.data` The data being passed from a `tidy_` distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Uniform: [tidy\\_uniform\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_uniform() |>
  util_uniform_stats_tbl() |>
  glimpse()
```

---

util\_weibull\_aic

*Calculate Akaike Information Criterion (AIC) for Weibull Distribution*

---

**Description**

This function estimates the shape and scale parameters of a Weibull distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

**Usage**

```
util_weibull_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a Weibull distribution.

**Details**

This function calculates the Akaike Information Criterion (AIC) for a Weibull distribution fitted to the provided data.

This function fits a Weibull distribution to the provided data using maximum likelihood estimation. It estimates the shape and scale parameters of the Weibull distribution using maximum likelihood estimation. Then, it calculates the AIC value based on the fitted distribution.

Initial parameter estimates: The function uses the method of moments estimates as starting points for the shape and scale parameters of the Weibull distribution.

Optimization method: The function uses the `optim` function for optimization. You might explore different optimization methods within `optim` for potentially better performance.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

**Value**

The AIC value calculated based on the fitted Weibull distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_geometric\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_poisson\\_aic\(\)](#)

**Examples**

```
# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rweibull(100, shape = 2, scale = 1)
util_weibull_aic(x)
```



---

`util_weibull_param_estimate`*Estimate Weibull Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated weibull data.

## Usage

```
util_weibull_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical` This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the weibull shape and scale parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#),

```
util_uniform_param_estimate(), util_zero_truncated_binomial_param_estimate(), util_zero_truncated_geom_param_estimate(),  
util_zero_truncated_negative_binomial_param_estimate(), util_zero_truncated_poisson_param_estimate()
```

Other Weibull: `tidy_inverse_weibull()`, `tidy_weibull()`, `util_weibull_stats_tbl()`

## Examples

```
library(dplyr)  
library(ggplot2)  
  
x <- tidy_weibull(.shape = 1, .scale = 2)$y  
output <- util_weibull_param_estimate(x)  
  
output$parameter_tbl  
  
output$combined_data_tbl %>%  
  tidy_combinedautoplot()
```

---

util\_weibull\_stats\_tbl

*Distribution Statistics*

---

## Description

Returns distribution statistics in a tibble.

## Usage

```
util_weibull_stats_tbl(.data)
```

## Arguments

`.data` The data being passed from a `tidy_` distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of `tidy_` distribution. It is required that data be passed from a `tidy_` distribution function.

## Value

A tibble

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Weibull: `tidy_inverse_weibull()`, `tidy_weibull()`, `util_weibull_param_estimate()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_zero_truncated_binomial_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

## Examples

```
library(dplyr)

tidy_weibull() |>
  util_weibull_stats_tbl() |>
  glimpse()
```

---

```
util_zero_truncated_binomial_aic
```

*Calculate Akaike Information Criterion (AIC) for Zero-Truncated Binomial Distribution*

---

## Description

This function estimates the parameters (size and prob) of a ZTB distribution from the provided data using maximum likelihood estimation (via the `optim()` function), and then calculates the AIC value based on the fitted distribution.

## Usage

```
util_zero_truncated_binomial_aic(.x)
```

## Arguments

`.x` A numeric vector containing the data (non-zero counts) to be fitted to a ZTB distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a zero-truncated binomial (ZTB) distribution fitted to the provided data.

**Initial parameter estimates:** The choice of initial values for size and probab can impact the convergence of the optimization. Consider using prior knowledge or method of moments estimates to obtain reasonable starting values.

**Optimization method:** The default optimization method used is "L-BFGS-B," which allows for box constraints to keep the parameters within valid bounds. You might explore other optimization methods available in `optim()` for potentially better performance or different constraint requirements.

**Data requirements:** The input data `.x` should consist of non-zero counts, as the ZTB distribution does not include zero values. Additionally, the values in `.x` should be less than or equal to the estimated size parameter.

**Goodness-of-fit:** While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen ZTB model using visualization (e.g., probability plots, histograms) and other statistical tests (e.g., chi-square goodness-of-fit test) to ensure it adequately describes the data.

## Value

The AIC value calculated based on the fitted ZTB distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative_binomial_aic()`, `util_zero_truncated_poisson_aic()`

## Examples

```
# Example data
set.seed(123)
x <- tidy_zero_truncated_binomial(30, .size = 10, .prob = 0.4)[["y"]]

# Calculate AIC
util_zero_truncated_binomial_aic(x)
```

---

`util_zero_truncated_binomial_param_estimate`*Estimate Zero Truncated Binomial Parameters*

---

## Description

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to `TRUE` then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated binomial data.

One method of estimating the parameters is done via:

- MLE via `optim` function.

## Usage

```
util_zero_truncated_binomial_param_estimate(.x, .auto_gen_empirical = TRUE)
```

## Arguments

`.x` The vector of data to be passed to the function.

`.auto_gen_empirical`

This is a boolean value of `TRUE/FALSE` with default set to `TRUE`. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

## Details

This function will attempt to estimate the zero truncated binomial size and prob parameters given some vector of values.

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: `util_bernoulli_param_estimate()`, `util_beta_param_estimate()`, `util_binomial_param_estimate()`, `util_burr_param_estimate()`, `util_cauchy_param_estimate()`, `util_chisquare_param_estimate()`, `util_exponential_param_estimate()`, `util_f_param_estimate()`, `util_gamma_param_estimate()`, `util_generalized_beta_param_estimate()`, `util_generalized_pareto_param_estimate()`, `util_geometric_param_estimate()`, `util_hypergeometric_param_estimate()`, `util_inverse_burr_param_estimate()`, `util_inverse_pareto_param_estimate()`, `util_inverse_weibull_param_estimate()`, `util_logistic_param_estimate()`, `util_lognormal_param_estimate()`, `util_negative_binomial_param_estimate()`, `util_normal_param_estimate()`

```

util_paralogistic_param_estimate(), util_pareto1_param_estimate(), util_pareto_param_estimate(),
util_poisson_param_estimate(), util_t_param_estimate(), util_triangular_param_estimate(),
util_uniform_param_estimate(), util_weibull_param_estimate(), util_zero_truncated_geometric_param_est
util_zero_truncated_negative_binomial_param_estimate(), util_zero_truncated_poisson_param_estimate()
Other Binomial: tidy_binomial(), tidy_negative_binomial(), tidy_zero_truncated_binomial(),
tidy_zero_truncated_negative_binomial(), util_binomial_param_estimate(), util_binomial_stats_tbl(),
util_negative_binomial_param_estimate(), util_zero_truncated_binomial_stats_tbl(),
util_zero_truncated_negative_binomial_param_estimate(), util_zero_truncated_negative_binomial_stats_
Other Zero Truncated Distribution: tidy_zero_truncated_binomial(), tidy_zero_truncated_geometric(),
tidy_zero_truncated_poisson()

```

## Examples

```

library(dplyr)
library(ggplot2)

x <- as.integer(mtcars$mpg)
output <- util_zero_truncated_binomial_param_estimate(x)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

set.seed(123)
t <- tidy_zero_truncated_binomial(100, 10, .1)[["y"]]
util_zero_truncated_binomial_param_estimate(t)$parameter_tbl

```

---

```
util_zero_truncated_binomial_stats_tbl
```

*Distribution Statistics for Zero Truncated Binomial Distribution*

---

## Description

Returns distribution statistics in a tibble for Zero Truncated Binomial distribution.

## Usage

```
util_zero_truncated_binomial_stats_tbl(.data)
```

## Arguments

`.data`            The data being passed from a tidy\_ distribution function.

## Details

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**See Also**

Other Binomial: `tidy_binomial()`, `tidy_negative_binomial()`, `tidy_zero_truncated_binomial()`, `tidy_zero_truncated_negative_binomial()`, `util_binomial_param_estimate()`, `util_binomial_stats_tbl()`, `util_negative_binomial_param_estimate()`, `util_zero_truncated_binomial_param_estimate()`, `util_zero_truncated_negative_binomial_param_estimate()`, `util_zero_truncated_negative_binomial_stats_tbl()`

Other Distribution Statistics: `util_bernoulli_stats_tbl()`, `util_beta_stats_tbl()`, `util_binomial_stats_tbl()`, `util_burr_stats_tbl()`, `util_cauchy_stats_tbl()`, `util_chisquare_stats_tbl()`, `util_exponential_stats_tbl()`, `util_f_stats_tbl()`, `util_gamma_stats_tbl()`, `util_generalized_beta_stats_tbl()`, `util_generalized_pareto_stats_tbl()`, `util_geometric_stats_tbl()`, `util_hypergeometric_stats_tbl()`, `util_inverse_burr_stats_tbl()`, `util_inverse_pareto_stats_tbl()`, `util_inverse_weibull_stats_tbl()`, `util_logistic_stats_tbl()`, `util_lognormal_stats_tbl()`, `util_negative_binomial_stats_tbl()`, `util_normal_stats_tbl()`, `util_paralogistic_stats_tbl()`, `util_pareto1_stats_tbl()`, `util_pareto_stats_tbl()`, `util_poisson_stats_tbl()`, `util_t_stats_tbl()`, `util_triangular_stats_tbl()`, `util_uniform_stats_tbl()`, `util_weibull_stats_tbl()`, `util_zero_truncated_geometric_stats_tbl()`, `util_zero_truncated_negative_binomial_stats_tbl()`, `util_zero_truncated_poisson_stats_tbl()`

**Examples**

```
library(dplyr)

set.seed(123)
tidy_zero_truncated_binomial(.size = 10, .prob = 0.1) |>
  util_zero_truncated_binomial_stats_tbl() |>
  glimpse()
```

---

```
util_zero_truncated_geometric_aic
```

*Calculate Akaike Information Criterion (AIC) for Zero-Truncated Geometric Distribution*

---

**Description**

This function estimates the probability parameter of a Zero-Truncated Geometric distribution from the provided data and calculates the AIC value based on the fitted distribution.

**Usage**

```
util_zero_truncated_geometric_aic(.x)
```

**Arguments**

`.x` A numeric vector containing the data to be fitted to a Zero-Truncated Geometric distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a Zero-Truncated Geometric distribution fitted to the provided data.

This function fits a Zero-Truncated Geometric distribution to the provided data. It estimates the probability parameter using the method of moments and calculates the AIC value.

Optimization method: Since the parameter is directly calculated from the data, no optimization is needed.

Goodness-of-fit: While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen model using visualization and other statistical tests.

## Value

The AIC value calculated based on the fitted Zero-Truncated Geometric distribution to the provided data.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Utility: [check\\_duplicate\\_rows\(\)](#), [convert\\_to\\_ts\(\)](#), [quantile\\_normalize\(\)](#), [tidy\\_mcmc\\_sampling\(\)](#), [util\\_beta\\_aic\(\)](#), [util\\_binomial\\_aic\(\)](#), [util\\_cauchy\\_aic\(\)](#), [util\\_chisq\\_aic\(\)](#), [util\\_exponential\\_aic\(\)](#), [util\\_f\\_aic\(\)](#), [util\\_gamma\\_aic\(\)](#), [util\\_generalized\\_beta\\_aic\(\)](#), [util\\_generalized\\_pareto\\_aic\(\)](#), [util\\_geometric\\_aic\(\)](#), [util\\_hypergeometric\\_aic\(\)](#), [util\\_inverse\\_burr\\_aic\(\)](#), [util\\_inverse\\_pareto\\_aic\(\)](#), [util\\_inverse\\_weibull\\_aic\(\)](#), [util\\_logistic\\_aic\(\)](#), [util\\_lognormal\\_aic\(\)](#), [util\\_negative\\_binomial\\_aic\(\)](#), [util\\_normal\\_aic\(\)](#), [util\\_paralogistic\\_aic\(\)](#), [util\\_pareto1\\_aic\(\)](#), [util\\_pareto\\_aic\(\)](#), [util\\_poisson\\_aic\(\)](#), [util\\_t\\_aic\(\)](#), [util\\_triangular\\_aic\(\)](#), [util\\_uniform\\_aic\(\)](#), [util\\_weibull\\_aic\(\)](#), [util\\_zero\\_truncated\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_aic\(\)](#), [util\\_zero\\_truncated\\_](#)

## Examples

```
library(actuar)

# Example: Calculate AIC for a sample dataset
set.seed(123)
x <- rztgeom(100, prob = 0.2)
util_zero_truncated_geometric_aic(x)
```

---

util\_zero\_truncated\_geometric\_param\_estimate

*Estimate Zero-Truncated Geometric Parameters*

---



**Description**

This function will estimate the prob parameter for a Zero-Truncated Geometric distribution from a given vector `.x`. The function returns a list with a parameter table, and if `.auto_gen_empirical` is set to TRUE, the empirical data is combined with the estimated distribution data.

**Usage**

```
util_zero_truncated_geometric_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function. Must contain non-negative integers and should have no zeros.

`.auto_gen_empirical` Boolean value (default TRUE) that, when set to TRUE, will generate `tidy_empirical()` output for `.x` and combine it with the estimated distribution data.

**Details**

This function will attempt to estimate the prob parameter of the Zero-Truncated Geometric distribution using given vector `.x` as input data. If the parameter `.auto_gen_empirical` is set to TRUE, the empirical data in `.x` will be run through the `tidy_empirical()` function and combined with the estimated zero-truncated geometric data.

**Value**

A tibble/list

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Zero-Truncated Geometric: [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#)

**Examples**

```
library(actuar)
library(dplyr)
library(ggplot2)
library(actuar)
```

```

set.seed(123)
ztg <- rztgeom(100, prob = 0.2)
output <- util_zero_truncated_geometric_param_estimate(ztg)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combined_autoplot()

```

---

util\_zero\_truncated\_geometric\_stats\_tbl

*Distribution Statistics for Zero-Truncated Geometric*

---

### Description

Returns distribution statistics for Zero-Truncated Geometric distribution in a tibble.

### Usage

```
util_zero_truncated_geometric_stats_tbl(.data)
```

### Arguments

`.data`            The data being passed from a `tidy_ztgeom` distribution function.

### Details

This function takes in a tibble generated by a `tidy_ztgeom` distribution function and returns the relevant statistics for a Zero-Truncated Geometric distribution. It requires data to be passed from a `tidy_ztgeom` distribution function.

### Value

A tibble

### See Also

Other Zero-Truncated Geometric: [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

## Examples

```
library(dplyr)

set.seed(123)
tidy_zero_truncated_geometric(.prob = 0.1) |>
  util_zero_truncated_geometric_stats_tbl() |>
  glimpse()
```

---

```
util_zero_truncated_negative_binomial_aic
```

*Calculate Akaike Information Criterion (AIC) for Zero-Truncated  
Negative Binomial Distribution*

---

## Description

This function estimates the parameters (size and prob) of a ZTNB distribution from the provided data using maximum likelihood estimation (via the `optim()` function), and then calculates the AIC value based on the fitted distribution.

## Usage

```
util_zero_truncated_negative_binomial_aic(.x)
```

## Arguments

`.x` A numeric vector containing the data (non-zero counts) to be fitted to a ZTNB distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a zero-truncated negative binomial (ZTNB) distribution fitted to the provided data.

**Initial parameter estimates:** The choice of initial values for size and prob can impact the convergence of the optimization. Consider using prior knowledge or method of moments estimates to obtain reasonable starting values.

**Optimization method:** The default optimization method used is "Nelder-Mead". You might explore other optimization methods available in `optim()` for potentially better performance or different constraint requirements.

**Data requirements:** The input data `.x` should consist of non-zero counts, as the ZTNB distribution does not include zero values.

**Goodness-of-fit:** While AIC is a useful metric for model comparison, it's recommended to also assess the goodness-of-fit of the chosen ZTNB model using visualization (e.g., probability plots, histograms) and other statistical tests (e.g., chi-square goodness-of-fit test) to ensure it adequately describes the data.

**Value**

The AIC value calculated based on the fitted ZTNB distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_poisson_aic()`

**Examples**

```
library(actuar)

# Example data
set.seed(123)
x <- rztnbinom(30, size = 2, prob = 0.4)

# Calculate AIC
util_zero_truncated_negative_binomial_aic(x)
```

---

```
util_zero_truncated_negative_binomial_param_estimate
  Estimate Zero Truncated Negative Binomial Parameters
```

---

**Description**

The function will return a list output by default, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated negative binomial data.

One method of estimating the parameters is done via:

- MLE via `optim` function.

**Usage**

```
util_zero_truncated_negative_binomial_param_estimate(
  .x,
  .auto_gen_empirical = TRUE
)
```

**Arguments**

- `.x` The vector of data to be passed to the function.
- `.auto_gen_empirical`  
This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.

**Details**

This function will attempt to estimate the zero truncated negative binomial size and prob parameters given some vector of values.

**Value**

A tibble/list

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Binomial: [tidy\\_binomial\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_negative\\_binomial\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

Other Zero Truncated Negative Distribution: [tidy\\_zero\\_truncated\\_negative\\_binomial\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)
library(actuar)

x <- as.integer(mtcars$mpg)
output <- util_zero_truncated_negative_binomial_param_estimate(x)
```

```
output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()

set.seed(123)
t <- rzttnbinom(100, 10, .1)
util_zero_truncated_negative_binomial_param_estimate(t)$parameter_tbl
```

---

util\_zero\_truncated\_negative\_binomial\_stats\_tbl

*Distribution Statistics for Zero-Truncated Negative Binomial*

---

## Description

Computes distribution statistics for a zero-truncated negative binomial distribution.

## Usage

```
util_zero_truncated_negative_binomial_stats_tbl(.data)
```

## Arguments

`.data`            The data from a zero-truncated negative binomial distribution.

## Details

This function computes statistics for a zero-truncated negative binomial distribution.

## Value

A tibble with distribution statistics.

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Binomial: [tidy\\_binomial\(\)](#), [tidy\\_negative\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_binomial\(\)](#), [tidy\\_zero\\_truncated\\_negative\\_binomial\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#)

Other Negative Binomial: [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\(\)](#)

```
util_geometric_stats_tbl(), util_hypergeometric_stats_tbl(), util_inverse_burr_stats_tbl(),  
util_inverse_pareto_stats_tbl(), util_inverse_weibull_stats_tbl(), util_logistic_stats_tbl(),  
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util_weibull_stats_tbl(), util_zero_truncated_binomial_stats_tbl(), util_zero_truncated_geometric_stats_tbl(),  
util_zero_truncated_poisson_stats_tbl()
```

## Examples

```
library(dplyr)  
  
tidy_zero_truncated_negative_binomial(.size = 1, .prob = 0.1) |>  
  util_zero_truncated_negative_binomial_stats_tbl() |>  
  glimpse()
```

---

util\_zero\_truncated\_poisson\_aic

*Calculate Akaike Information Criterion (AIC) for zero-truncated poisson Distribution*

---

## Description

This function estimates the parameters of a zero-truncated poisson distribution from the provided data using maximum likelihood estimation, and then calculates the AIC value based on the fitted distribution.

## Usage

```
util_zero_truncated_poisson_aic(.x)
```

## Arguments

`.x` A numeric vector containing the data to be fitted to a zero-truncated poisson distribution.

## Details

This function calculates the Akaike Information Criterion (AIC) for a zero-truncated poisson distribution fitted to the provided data.

## Value

The AIC value calculated based on the fitted zero-truncated poisson distribution to the provided data.

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Utility: `check_duplicate_rows()`, `convert_to_ts()`, `quantile_normalize()`, `tidy_mcmc_sampling()`, `util_beta_aic()`, `util_binomial_aic()`, `util_cauchy_aic()`, `util_chisq_aic()`, `util_exponential_aic()`, `util_f_aic()`, `util_gamma_aic()`, `util_generalized_beta_aic()`, `util_generalized_pareto_aic()`, `util_geometric_aic()`, `util_hypergeometric_aic()`, `util_inverse_burr_aic()`, `util_inverse_pareto_aic()`, `util_inverse_weibull_aic()`, `util_logistic_aic()`, `util_lognormal_aic()`, `util_negative_binomial_aic()`, `util_normal_aic()`, `util_paralogistic_aic()`, `util_pareto1_aic()`, `util_pareto_aic()`, `util_poisson_aic()`, `util_t_aic()`, `util_triangular_aic()`, `util_uniform_aic()`, `util_weibull_aic()`, `util_zero_truncated_binomial_aic()`, `util_zero_truncated_geometric_aic()`, `util_zero_truncated_negative`

**Examples**

```
library(actuar)

# Example 1: Calculate AIC for a sample dataset
set.seed(123)
x <- rztpois(30, lambda = 3)
util_zero_truncated_poisson_aic(x)
```

---

```
util_zero_truncated_poisson_param_estimate
      Estimate Zero Truncated Poisson Parameters
```

---

**Description**

This function will attempt to estimate the Zero Truncated Poisson lambda parameter given some vector of values `.x`. The function will return a tibble output, and if the parameter `.auto_gen_empirical` is set to TRUE then the empirical data given to the parameter `.x` will be run through the `tidy_empirical()` function and combined with the estimated Zero Truncated Poisson data.

**Usage**

```
util_zero_truncated_poisson_param_estimate(.x, .auto_gen_empirical = TRUE)
```

**Arguments**

`.x` The vector of data to be passed to the function. Must be non-negative integers.

`.auto_gen_empirical`

This is a boolean value of TRUE/FALSE with default set to TRUE. This will automatically create the `tidy_empirical()` output for the `.x` parameter and use the `tidy_combine_distributions()`. The user can then plot out the data using `$combined_data_tbl` from the function output.



## Details

This function estimates the parameter  $\lambda$  of a Zero-Truncated Poisson distribution based on a vector of non-negative integer values  $.x$ . The Zero-Truncated Poisson distribution is a discrete probability distribution that models the number of events occurring in a fixed interval of time, given that at least one event has occurred.

The estimation is performed by minimizing the negative log-likelihood of the observed data  $.x$  under the Zero-Truncated Poisson model. The negative log-likelihood function used for optimization is defined as:

$$-\sum_{i=1}^n \log(P(X_i = x_i | X_i > 0, \lambda))$$

where  $(X_i)$  are the observed values in  $.x$  and  $\lambda$  is the parameter of the Zero-Truncated Poisson distribution.

The optimization process uses the `optim` function to find the value of  $\lambda$  that minimizes this negative log-likelihood. The chosen optimization method is Brent's method (`method = "Brent"`) within a specified interval  $[\emptyset, \max(.x)]$ .

If `.auto_gen_empirical` is set to `TRUE`, the function will generate empirical data statistics using `tidy_empirical()` for the input data  $.x$  and then combine this empirical data with the estimated Zero-Truncated Poisson distribution using `tidy_combine_distributions()`. This combined data can be accessed via the `$combined_data_tbl` element of the function output.

The function returns a tibble containing the estimated parameter  $\lambda$  along with other summary statistics of the input data (sample size, minimum, maximum).

## Value

A tibble/list

## Author(s)

Steven P. Sanderson II, MPH

## See Also

Other Parameter Estimation: [util\\_bernoulli\\_param\\_estimate\(\)](#), [util\\_beta\\_param\\_estimate\(\)](#), [util\\_binomial\\_param\\_estimate\(\)](#), [util\\_burr\\_param\\_estimate\(\)](#), [util\\_cauchy\\_param\\_estimate\(\)](#), [util\\_chisquare\\_param\\_estimate\(\)](#), [util\\_exponential\\_param\\_estimate\(\)](#), [util\\_f\\_param\\_estimate\(\)](#), [util\\_gamma\\_param\\_estimate\(\)](#), [util\\_generalized\\_beta\\_param\\_estimate\(\)](#), [util\\_generalized\\_pareto\\_param\\_estimate\(\)](#), [util\\_geometric\\_param\\_estimate\(\)](#), [util\\_hypergeometric\\_param\\_estimate\(\)](#), [util\\_inverse\\_burr\\_param\\_estimate\(\)](#), [util\\_inverse\\_pareto\\_param\\_estimate\(\)](#), [util\\_inverse\\_weibull\\_param\\_estimate\(\)](#), [util\\_logistic\\_param\\_estimate\(\)](#), [util\\_lognormal\\_param\\_estimate\(\)](#), [util\\_negative\\_binomial\\_param\\_estimate\(\)](#), [util\\_normal\\_param\\_estimate\(\)](#), [util\\_paralogistic\\_param\\_estimate\(\)](#), [util\\_pareto1\\_param\\_estimate\(\)](#), [util\\_pareto\\_param\\_estimate\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_t\\_param\\_estimate\(\)](#), [util\\_triangular\\_param\\_estimate\(\)](#), [util\\_uniform\\_param\\_estimate\(\)](#), [util\\_weibull\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_binomial\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_geometric\\_param\\_estimate\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_param\\_estimate\(\)](#)

Other Poisson: [tidy\\_poisson\(\)](#), [tidy\\_zero\\_truncated\\_poisson\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)
library(ggplot2)

tc <- tidy_zero_truncated_poisson() |> pull(y)
output <- util_zero_truncated_poisson_param_estimate(tc)

output$parameter_tbl

output$combined_data_tbl |>
  tidy_combinedautoplot()
```

---

util\_zero\_truncated\_poisson\_stats\_tbl  
*Distribution Statistics*

---

**Description**

Returns distribution statistics in a tibble.

**Usage**

```
util_zero_truncated_poisson_stats_tbl(.data)
```

**Arguments**

`.data`            The data being passed from a tidy\_ distribution function.

**Details**

This function will take in a tibble and returns the statistics of the given type of tidy\_ distribution. It is required that data be passed from a tidy\_ distribution function.

**Value**

A tibble

**Author(s)**

Steven P. Sanderson II, MPH

**See Also**

Other Poisson: [tidy\\_poisson\(\)](#), [tidy\\_zero\\_truncated\\_poisson\(\)](#), [util\\_poisson\\_param\\_estimate\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_poisson\\_param\\_estimate\(\)](#)

Other Distribution Statistics: [util\\_bernoulli\\_stats\\_tbl\(\)](#), [util\\_beta\\_stats\\_tbl\(\)](#), [util\\_binomial\\_stats\\_tbl\(\)](#), [util\\_burr\\_stats\\_tbl\(\)](#), [util\\_cauchy\\_stats\\_tbl\(\)](#), [util\\_chisquare\\_stats\\_tbl\(\)](#), [util\\_exponential\\_stats\\_tbl\(\)](#), [util\\_f\\_stats\\_tbl\(\)](#), [util\\_gamma\\_stats\\_tbl\(\)](#), [util\\_generalized\\_beta\\_stats\\_tbl\(\)](#), [util\\_generalized\\_pareto\\_stats\\_tbl\(\)](#), [util\\_geometric\\_stats\\_tbl\(\)](#), [util\\_hypergeometric\\_stats\\_tbl\(\)](#), [util\\_inverse\\_burr\\_stats\\_tbl\(\)](#), [util\\_inverse\\_pareto\\_stats\\_tbl\(\)](#), [util\\_inverse\\_weibull\\_stats\\_tbl\(\)](#), [util\\_logistic\\_stats\\_tbl\(\)](#), [util\\_lognormal\\_stats\\_tbl\(\)](#), [util\\_negative\\_binomial\\_stats\\_tbl\(\)](#), [util\\_normal\\_stats\\_tbl\(\)](#), [util\\_paralogistic\\_stats\\_tbl\(\)](#), [util\\_pareto1\\_stats\\_tbl\(\)](#), [util\\_pareto\\_stats\\_tbl\(\)](#), [util\\_poisson\\_stats\\_tbl\(\)](#), [util\\_t\\_stats\\_tbl\(\)](#), [util\\_triangular\\_stats\\_tbl\(\)](#), [util\\_uniform\\_stats\\_tbl\(\)](#), [util\\_weibull\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_binomial\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_geometric\\_stats\\_tbl\(\)](#), [util\\_zero\\_truncated\\_negative\\_binomial\\_stats\\_tbl\(\)](#)

**Examples**

```
library(dplyr)

tidy_zero_truncated_poisson() |>
  util_zero_truncated_poisson_stats_tbl() |>
  glimpse()
```

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