

# Package ‘ptsuite’

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**Title** Tail Index Estimation for Power Law Distributions

**Version** 1.0.0

**Description** Various estimation methods for the shape parameter of Pareto distributed data. This package contains functions for various estimation methods such as maximum likelihood (Newman, 2005)<[doi:10.1016/j.cities.2012.03.001](https://doi.org/10.1016/j.cities.2012.03.001)>, Hill's estimator (Hill, 1975)<[doi:10.1214/aos/1176343247](https://doi.org/10.1214/aos/1176343247)>, least squares (Zaher et al., 2014)<[doi:10.9734/BJMCS/2014/10890](https://doi.org/10.9734/BJMCS/2014/10890)>, method of moments (Rytgaard, 1990)<[doi:10.2143/AST.20.2.2005443](https://doi.org/10.2143/AST.20.2.2005443)>, percentiles (Bhatti et al., 2018)<[doi:10.1371/journal.pone.0196456](https://doi.org/10.1371/journal.pone.0196456)>, and weighted least squares (Nair et al., 2019) to estimate the shape parameter of Pareto distributed data. It also provides both a heuristic method (Hubert et al., 2013)<[doi:10.1016/j.csda.2012.07.011](https://doi.org/10.1016/j.csda.2012.07.011)> and a goodness of fit test (Gulati and Shapiro, 2008)<[doi:10.1007/978-0-8176-4619-6](https://doi.org/10.1007/978-0-8176-4619-6)> for testing for Pareto data as well as a method for generating Pareto distributed data.

**Depends** R (>= 3.5.0)

**License** GPL-3

**LazyData** true

**LinkingTo** Rcpp

**Imports** Rcpp

**RoxygenNote** 6.1.1

**Suggests** plotly

**NeedsCompilation** yes

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ptsuite-package	<i>ptsuite: Package for Pareto Parameter Estimation</i>
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### Description

The ptsuite package provides functions to estimate parameters of Type I Pareto data. The details of the methods used and the equations implemented are given in the vignette. The package contains the following functions:

### Estimator Functions

- Maximum Likelihood Estimator {[alpha\\_mle](#)}
- Weighted Least Squares Estimator {[alpha\\_wls](#)}
- Hill's Estimator {[alpha\\_hills](#)}
- Method of Moments Estimator {[alpha\\_moment](#)}
- Method of Percentiles Estimator {[alpha\\_percentile](#)}
- Method of Modified Percentiles Estimator {[alpha\\_modified\\_percentile](#)}
- Method of Geometric Percentiles {[alpha\\_geometric\\_percentile](#)}
- Least Squares Estimator {[alpha\\_ls](#)}

### Other Functions

- Generate Pareto Data {[generate\\_pareto](#)}
- Estimates from all estimators {[generate\\_all\\_estimates](#)}
- Q-Q Plot to test for Pareto Distribution {[pareto\\_qq\\_test](#)}
- Pareto Test {[pareto\\_test](#)}

## References

- Newman MEJ (2005). "Power Laws, Pareto Distributions And Zipf's Law." Contemporary Physics, 46, 323-351.
- Nair J, Wierman A, Zwart B (2019). "The Fundamentals Of Heavy Tails: Properties, Emergence, And Identification." <http://users.cms.caltech.edu/~adamw/heavytails.html>.
- Pokorna M (2016). Estimation and Application of the Tail Index. Bachelor's thesis, Charles University in Prague, Faculty of Social Sciences, Institute of Economic Studies.
- Hill B (1975). "A Simple General Approach To Inference About The Tail Of A Distribution." The Annals of Statistics, 3(5), 1163-1174.
- Rytgaard M (1990). "Estimation In The Pareto Distribution." ASTIN Bulletin: The Journal Of The IAA, 20(2), 201-216.
- Brazauskas V, Serfling R (2000). "Robust and Efficient Estimation Of The Tail Index Of A Single-Parameter Pareto Distribution." North American Actuarial Journal, 4, 12-27.
- Bhatti SH, Hussain S, Ahmad T, Aslam M, Aftab M, Raza MA (2018). "Efficient estimation of Pareto model: Some modified percentile estimators." PLoS ONE, 13(5), 1-15.
- Zaher HM, El-Sheik AA, El-Magd NATA (2014). "Estimation of Pareto Parameters Using a Fuzzy Least-Squares Method and Other Known Techniques with a Comparison." British Journal of Mathematics & Computer Science, 4(14), 2067-2088.
- Gulati S, Shapiro S (2008). "Goodness-of-Fit Tests for Pareto Distribution." In F Vonta (ed.), Statistical Models and Methods for Biomedical and Technical Systems, chapter 19, pp. 259-274. Birkhauser Basel. ISBN 978-0-8176-4619-6. doi:10.1007/978-0-8176-4619-6.

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alpha\_geometric\_percentile

*Estimating the Shape Parameter by Geometric Method of Percentiles*

---

## Description

This function uses the Geometric Method of Percentiles to estimate the shape parameter of a given set of data. (Bhatti et al. 2018)

## Usage

```
alpha_geometric_percentile(dat)
```

## Arguments

dat                    vector of observations

## Value

A list of the following form:

**shape** Estimate of the shape parameter of the data

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)

## References

Bhatti SH, Hussain S, Ahmad T, Aslam M, Aftab M, Raza MA (2018). "Efficient estimation of Pareto model: Some modified percentile estimators." PLoS ONE, 13(5), 1-15.

## Examples

```
x <- generate_pareto(10000, 5, 2)
alpha_geometric_percentile(x)
```

---

alpha\_hills

*Estimating the Shape Parameter by Hill's Estimator*

---

## Description

This function uses the Hill's Estimator to estimate the shape parameter of a given set of data. (Nair et al. 2019; Pokorna 2016; Hill 1975) It is especially useful when the data is known not to follow an exact Pareto distribution but the tail of the data does. Thus, the specification of  $k$ , the  $k$ th largest observation, allows to specify the point from where Pareto-like behavior may be seen. It is also possible to specify the value at which the tail begins. When  $k=n$ , the Hill's Estimator returns the same estimate as `alpha_mle` with a warning notifying the user.

## Usage

```
alpha_hills(dat, k, value = FALSE)
```

## Arguments

<code>dat</code>	vector of observations
<code>k</code>	number of observations / value equal to or greater than to consider for tail
<code>value</code>	(TRUE/FALSE) indicating if the value which is specified in "k" (TRUE)

## Value

A list of the following form:

**shape** Estimate of the shape parameter of the data

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)

## References

Nair J, Wierman A, Zwart B (2019). "The Fundamentals Of Heavy Tails: Properties, Emergence, And Identification." <http://users.cms.caltech.edu/~adamw/heavytails.html>.

Pokorna M (2016). Estimation and Application of the Tail Index. Bachelor's thesis, Charles University in Prague, Faculty of Social Sciences, Institute of Economic Studies.

Hill B (1975). "A Simple General Approach To Inference About The Tail Of A Distribution." The Annals of Statistics, 3(5), 1163-1174.

**Examples**

```
x <- generate_pareto(10000, 5, 2)
alpha_hills(x, 400)
```

---

alpha\_ls

*Estimating the Shape Parameter by Method of Least Squares*

---

**Description**

This function uses the Method of Least Squares to estimate the shape parameter of a given set of data. (Zaher et al. 2014; Nair et al. 2019)

**Usage**

```
alpha_ls(dat)
```

**Arguments**

dat                    vector of observations

**Value**

A list of the following form:

**shape** Estimate of the shape parameter of the data

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)

**References**

Zaher HM, El-Sheik AA, El-Magd NATA (2014). "Estimation of Pareto Parameters Using a Fuzzy Least-Squares Method and Other Known Techniques with a Comparison." *British Journal of Mathematics & Computer Science*, 4(14), 2067-2088.

Nair J, Wierman A, Zwart B (2019). "The Fundamentals Of Heavy Tails: Properties, Emergence, And Identification." <http://users.cms.caltech.edu/~adamw/heavytails.html>.

**Examples**

```
x <- generate_pareto(10000, 5, 2)
alpha_ls(x)
```

---

alpha_mle	<i>Estimating the Shape Parameter by Method of Maximum Likelihood (MLE)</i>
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---

### Description

This function can be used to estimate the shape parameter using the Maximum Likelihood Estimator method (Newman 2005). It can be used to obtain biased and unbiased estimates of the shape and scale parameters as well as the confidence interval for the shape parameter for the biased estimates.

### Usage

```
alpha_mle(dat, biased = TRUE, significance = NULL)
```

### Arguments

dat	vector of observations
biased	TRUE/FALSE to indicate biased or unbiased estimates
significance	level of significance

### Value

A list of the following form:

**shape** Estimate of the shape parameter of the data

**lower\_bound** Upper error bound of the estimate of shape

**upper\_bound** Lower error bound of the estimate of shape

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)

### References

Newman MEJ (2005). "Power Laws, Pareto Distributions And Zipf's Law." Contemporary Physics, 46, 323-351.

### Examples

```
x <- generate_pareto(10000, 5, 2)
alpha_mle(x, TRUE, 0.05)
```

```
x <- generate_pareto(10000, 5, 2)
alpha_mle(x, FALSE)
```

---

`alpha_modified_percentile`*Estimating the Shape Parameter by Method of Modified Percentiles*

---

**Description**

This function uses the Method of Modified Percentiles to estimate the shape parameter of a given set of data. (Bhatti et al. 2018)

**Usage**

```
alpha_modified_percentile(dat)
```

**Arguments**

`dat`                    vector of observations

**Value**

A list of the following form:

**shape** Estimate of the shape parameter of the data

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)

**References**

Bhatti SH, Hussain S, Ahmad T, Aslam M, Aftab M, Raza MA (2018). "Efficient estimation of Pareto model: Some modified percentile estimators." PLoS ONE, 13(5), 1-15.

**Examples**

```
x <- generate_pareto(10000, 5, 2)
alpha_modified_percentile(x)
```

---

`alpha_moment`*Estimating the Shape Parameter by Method of Moments*

---

**Description**

This function uses the Method of Moments to estimate the shape parameter of a given set of data. (Rytgaard 1990) The method of moments is only accurate if  $\alpha$  (shape parameter) is greater than or equal to 1 (Brazauskas and Serfling 2000). This function issues a warning if it detects that  $\alpha$  may be less than 1.

**Usage**

```
alpha_moment(dat)
```

**Arguments**

dat                    vector of observations

**Value**

A list of the following form:

**shape** Estimate of the shape parameter of the data

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)

**References**

Rytgaard M (1990). "Estimation In The Pareto Distribution." ASTIN Bulletin: The Journal Of The IAA, 20(2), 201-216.

Brazauskas V, Serfling R (2000). "Robust and Efficient Estimation Of The Tail Index Of A Single-Parameter Pareto Distribution." North American Actuarial Journal, 4, 12-27.

**Examples**

```
x <- generate_pareto(10000, 5, 2)
alpha_moment(x)
```

---

alpha\_percentile                    *Estimating the Shape Parameter by Method of Percentiles*

---

**Description**

This function uses the Method of Percentiles to estimate the shape parameter of a given set of data. (Bhatti et al. 2018)

**Usage**

```
alpha_percentile(dat)
```

**Arguments**

dat                    vector of observations

**Value**

A list of the following form:

**shape** Estimate of the shape parameter of the data

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)



## References

Bhatti SH, Hussain S, Ahmad T, Aslam M, Aftab M, Raza MA (2018). "Efficient estimation of Pareto model: Some modified percentile estimators." PLoS ONE, 13(5), 1-15.

## Examples

```
x <- generate_pareto(10000, 5, 2)
alpha_percentile(x)
```

---

alpha_wls	<i>Estimating the Shape Parameter by Weighted Least Squares Method (WLS)</i>
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## Description

This function uses the Weighted Least Squares Method (WLS) to estimate the shape parameter of a given set of data. (Nair et al. 2019)

## Usage

```
alpha_wls(dat)
```

## Arguments

dat                    vector of observations

## Value

A list of the following form:

**shape** Estimate of the shape parameter of the data

**scale** Estimate of the scale parameter of the data (which is taken to be the minimum of the data)

## References

Nair J, Wierman A, Zwart B (2019). "The Fundamentals Of Heavy Tails: Properties, Emergence, And Identification." <http://users.cms.caltech.edu/~adamw/heavytails.html>.

## Examples

```
x <- generate_pareto(10000, 5, 2)
alpha_percentile(x)
```

---

`generate_all_estimates`*Obtain estimates for Parameters of Pareto Data from all methods*

---

**Description**

This function combines the results of all the methods (included in this package) provided to estimate the shape and scale parameters of the Pareto data and provides the results in a data frame. Hill's Estimator is not used in this comparison as it discards a set of observations. We also note here that when considering the entire data set, Hill's Estimate is equivalent to the MLE.

**Usage**

```
generate_all_estimates(dat)
```

**Arguments**

`dat`                    vector of observations

**Value**

Dataframe with the following columns:

**Method.of.Estimation** Name of the method used for estimation

**Shape.Parameter** Estimates of the shape parameter of the data

**Scale.Parameter** Estimates of the scale parameter of the data

**Examples**

```
x <- generate_pareto(10000, 5, 2)
generate_all_estimates(x)
```

---

`generate_pareto`*Generating data from a Pareto Distribution.*

---

**Description**

This function is able to generate random Pareto distributed data with the specified shape and scale parameters. The function has been written to be similar in type to the popular `runif` and `rexp` type of functions for generating data from a particular distribution.

**Usage**

```
generate_pareto(sample_size, shape, scale)
```

**Arguments**

sample_size	number of observations
shape	shape parameter
scale	scale parameter

**Value**

Vector of Pareto distributed data of sample size `sample_size` with shape parameter `shape` and scale parameter `scale`.

**Examples**

```
generate_pareto(10000, 5, 2)
generate_pareto(100, 15, 6)
```

---

pareto_qq_test	<i>Q-Q Plot to test for Pareto Distribution</i>
----------------	---

---

**Description**

This function can be used as a first step to identify whether the data is Pareto distributed before estimating the tail index. If most of the data points appear to be distributed along a line, it is possible that the data may be Pareto. Conversely, if most of the data are distributed non-linearly, then the data is most probably not Pareto distributed.

**Usage**

```
pareto_qq_test(dat)
```

**Arguments**

dat	Data to be tested for Pareto distribution
-----	---

**Details**

This function plots the quantiles of the standard exponential distribution on the x-axis and the log values of the provided data on the y-axis. If Pareto data was supplied, a log transformation of this data would result in an exponential distribution with mean  $\alpha$ . These data points would then show up on the QQ-plot as a line with slope  $1/\alpha$ .

The function makes use of the `plotly` package if available and installed or if not, defaults to the standard R plot.

**Value**

A Q-Q plot either using `plotly` if package is available or else a standard R plot.

**Examples**

```
x <- generate_pareto(10000, 5, 2)
pareto_qq_test(x)
```

---

pareto\_test

*Goodness of Fit Test for Pareto Distribution*

---

**Description**

The `pareto_test` function can be used to identify whether the data is Pareto distributed (Gulati and Shapiro 2008). The test generates a p-value corresponding to the actual distribution of the data and is tested for significance. In the case of Pareto data, the p-value should be greater than the pre-determined significance level (generally taken as 0.05).

**Usage**

```
pareto_test(dat)
```

**Arguments**

`dat`                    vector of observations

**Value**

A list of the following form:

**p-value** p-value indicating significance of the test

**References**

Gulati S, Shapiro S (2008). "Goodness-of-Fit Tests for Pareto Distribution." In F Vonta (ed.), *Statistical Models and Methods for Biomedical and Technical Systems*, chapter 19, pp. 259-274. Birkhauser Basel. ISBN 978-0-8176-4619-6. doi:10.1007/978-0-8176-4619-6.

**Examples**

```
x <- generate_pareto(10000, 5, 2)
pareto_test(x)
```

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