

MAXIMUM SPACING ESTIMATION
A NEW METHOD IN FITDISTRPLUS

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The **fitdistrplus** project

- started in 2009: stable version 1.0-9 on CRAN (first release),
- extensively enhanced between 2009-2018: 17 versions on CRAN,
- published 2015: publication in JSS,
- currently, 2019: last stable version 1.0-14.

Presented at

- useR 2009 in Rennes, useR 2011 in Warwick,
- Rencontres R 2013 in Lyon, Rencontres R 2018 in Rennes.

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fitdistrplus extends the `fitdistr()` function (of the MASS package)

- Censored data may contain left censored, right censored and interval censored values, with several lower and upper bounds.
- the package also provides moment matching (MME), quantile matching (QME) and maximum goodness-of-fit estimation (MGE) methods.

Today, we present the implementation of a new estimation method: maximum spacing estimation (MSE)

- This method was introduced by Cheng and Amin (1983) and Ranneby (1984) independently.
- Currently, the **BMT** package provides MSE for the Bezier-Montenegro-Torres distribution.
- the **MPS** package provides MSE only for a selected number of distribution.

MAXIMUM SPACING ESTIMATION (MSE)

Consider a sample of observations (x_1, \dots, x_n) (assuming real-valued observations).

- We can compute order statistics as $x_{(1)} < \dots < x_{(j)} < \dots < x_{(n)}$.

- Spacings on the distribution function $F(\cdot; \theta)$ are defined as

$$D_i(\theta) = F(x_{(i)}; \theta) - F(x_{(i-1)}; \theta), \quad i = 1, \dots, n + 1$$

where $x_{(0)} = -\infty$ and $x_{(n+1)} = +\infty$.

- we know $D_i(\theta) > 0$ for any continuous cdf.

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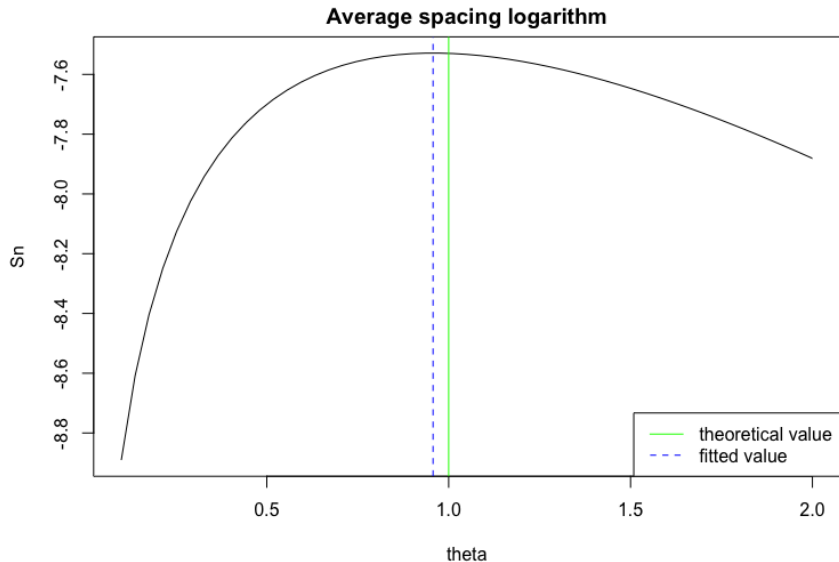
MSE consists in maximizing the average of the spacing logarithm

$$S_n(\theta) = \frac{1}{n+1} \sum_{i=1}^{n+1} \log D_i(\theta).$$

Under certain regularity conditions, MSE has asymptotically a normal distribution as MLE, see Ekstrom (2015).

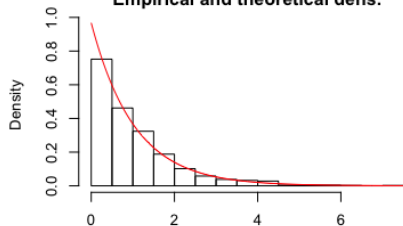
EXAMPLE 1 – STD. EXPONENTIAL DISTRIBUTION – AVG. SPACING

Simulated dataset from $\mathcal{E}(1)$ and $n = 1000$.



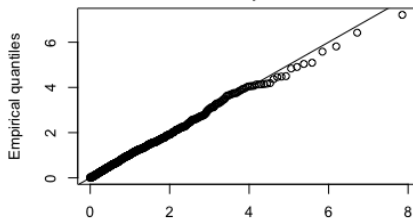
EXAMPLE 1 – STD. EXPONENTIAL DISTRIBUTION – GOF. PLOTS

Empirical and theoretical dens.



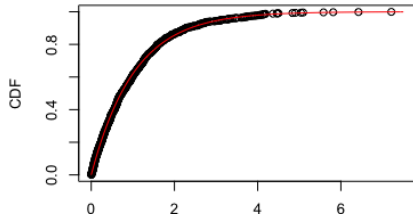
Data

Q-Q plot



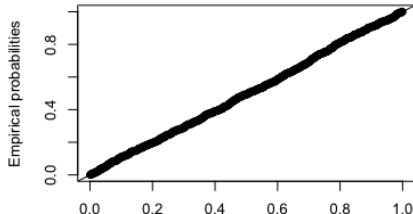
Theoretical quantiles

Empirical and theoretical CDFs



Data

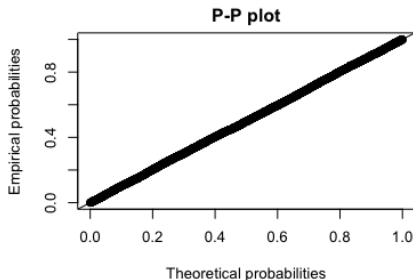
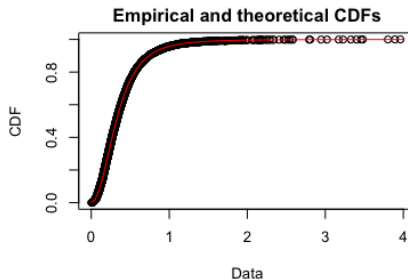
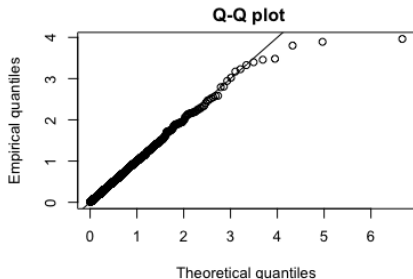
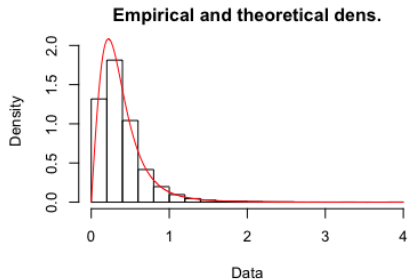
P-P plot



Theoretical probabilities

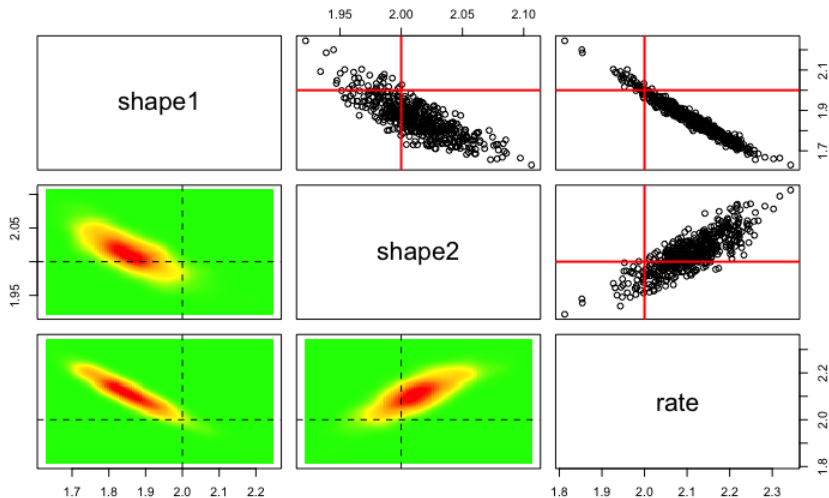
EXAMPLE 2 – BURR DISTRIBUTION – GOF. PLOTS

Simulated dataset from $Burr(2, 2, 2)$ and $n = 10000$.



EXAMPLE 2 – BURR DISTRIBUTION – BOOTSTRAP

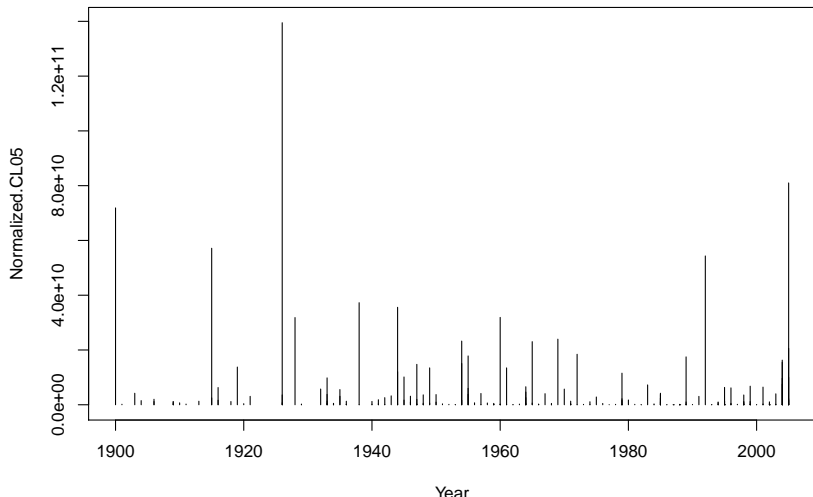
Bootstrapped values of parameters



EXAMPLE 3 – REAL DATASET

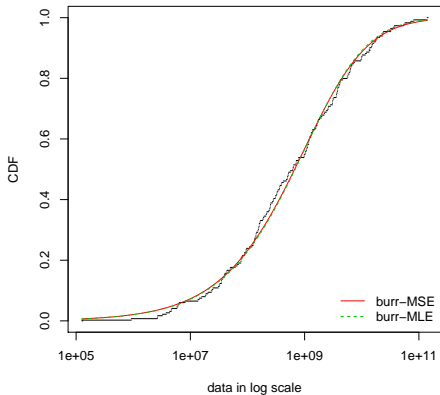
Consider Normalized Hurricane Damages in the United States: 1900-2005 used in Pielke et al. (2008).

Normalized Hurricane Damages in United States

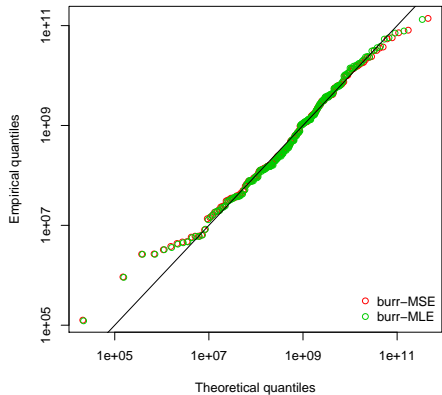


EXAMPLE 3 – REAL DATASET – GRAPHICS

Empirical and theoretical CDFs



Q-Q plot



EXAMPLE 3 – REAL DATASET – GOF. STATISTICS FOR BURR DISTRIBUTION

Coefficients values		
	MSE	MLE
shape1	2.2075	2.4563
shape2	0.5611	0.5604
rate	2.5318e-10	1.9825e-10


Goodness-of-fit statistics		
	MSE	MLE
Kolmogorov-Smirnov statistic	0.04421733	0.04867166
Cramer-von Mises statistic	0.06798511	0.08520099
Anderson-Darling statistic	0.43766768	0.51206510

Goodness-of-fit criteria		
	MSE	MLE
Akaike's Information Criterion	9301.882	9301.723
Bayesian Information Criterion	9311.880	9311.721

We implements (very quickly) in **fitdistrplus**


- a new statistical procedure Maximum Spacing Estimation.
- as well as the generalized Maximum Spacing Estimation considering ϕ divergence function.
- automatically all generic functions (`plot`, `summary`, `coef`,...) are available.

REFERENCES

 Cheng, R.C.H., Amin, N.A.K. (1983),
Estimating parameters in continuous univariate distributions with a shifted origin,
Journal of the Royal Statistical Society Ser. B, 45, 394-403.

 Ranneby B (1984),
The Maximum Spacing Method. An Estimation Method Related to the Maximum Likelihood Method,
Scandinavian Journal of Statistics, 11(2), 93-112

 Ranneby B & Ekstrom M (1997),
Maximum spacing estimates based on different metrics,
Umea universitet, research paper.

 Ekstrom M (2015),
On the consistency of the maximum spacing method,
Journal of Statistical Planning and Inference, 70, 209-224.

 Delignette-Muller ML & Dutang C (2019),
fitdistrplus: An R Package for Fitting Distributions,
<https://CRAN.R-project.org/package=fitdistrplus>

 Teimouri M. (2018),
MPS: Estimating Through the Maximum Product Spacing Approach,
<https://CRAN.R-project.org/package=MPS>

 Torres-Jimenez C.J. (2017),
BMT: The BMT Distribution,
<https://CRAN.R-project.org/package=BMT>