### Use of the library fitdistrplus to specify a distribution from non-censored or censored data

Marie Laure Delignette-Muller, Régis Pouillot , Jean-Baptiste Denis and Christophe Dutang December 10, 2009

Here you will find some easy examples of use of the functions of the library fitdistrplus. The aim is to show you by examples how to use these functions to help you to specify a parametric distribution from data corresponding to a random sample drawn from a theoretical distribution that you want to describe. For details, see the documentation of each function, using the R help command (ex.: ?fitdist). Do not forget to load the library using the function library before testing following examples.

#### > library(fitdistrplus)

#### Contents

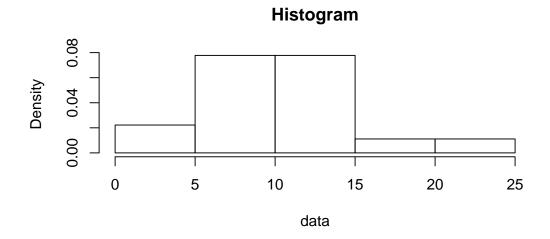
1	Specification of a distribution from non-censored continuous data	2
	1.1 Graphical display of the observed distribution	2
	1.2 Characterization of the observed distribution	2
	1.3 Fitting of a distribution	4
	1.4 Simulation of the uncertainty by boostrap	12
2	Specification of a distribution from non-censored discrete data	13
3	Specification of a distribution from censored data	17
	3.1 Graphical display of the observed distribution	17
	3.2 Fitting of a distribution	20

#### 1 Specification of a distribution from non-censored continuous data

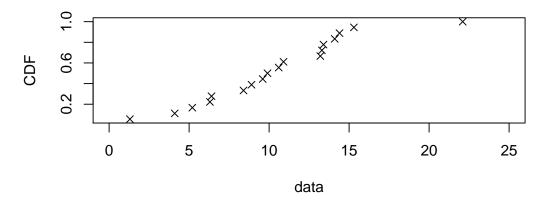
#### 1.1 Graphical display of the observed distribution

First of all, the observed distribution may be plotted using the function plotdist.

```
> x1 <- c(6.4, 13.3, 4.1, 1.3, 14.1, 10.6, 9.9, 9.6, 15.3, 22.1,
+ 13.4, 13.2, 8.4, 6.3, 8.9, 5.2, 10.9, 14.4)
> plotdist(x1)
```



#### **Cumulative distribution**



#### 1.2 Characterization of the observed distribution

Descriptive parameters of the empirical distribution may be computed using the function descdist. This function will also provide by default a skewness-kurtosis plot which may help you to select which distribution(s) to fit among the potential candidates.

#### > descdist(x1)

summary statistics

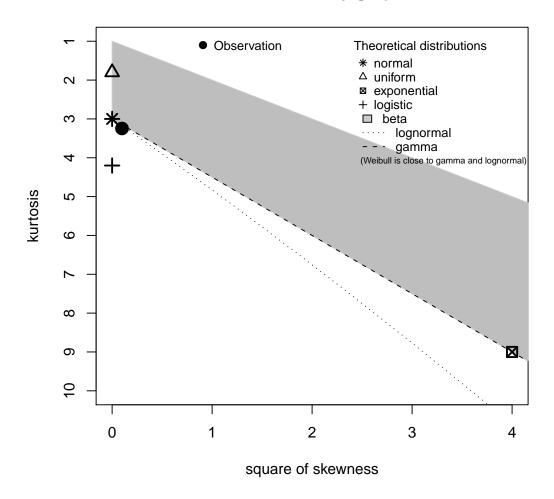
-----

min: 1.3 max: 22.1

median: 10.2 mean: 10.4 sample sd: 4.75

sample skewness: 0.314 sample kurtosis: 3.25

#### **Cullen and Frey graph**



In order to take into account the uncertainty of the estimated values of kurtosis and skewness, the data set may be boostrapped by fixing the argument boot to an integer above 10 in descdist. boot values of skewness and kurtosis corresponding to the boot nonparametric bootstrap samples are then computed and reported in blue color on the skewness-kurtosis plot.

#### > descdist(x1, boot = 1000)

#### summary statistics

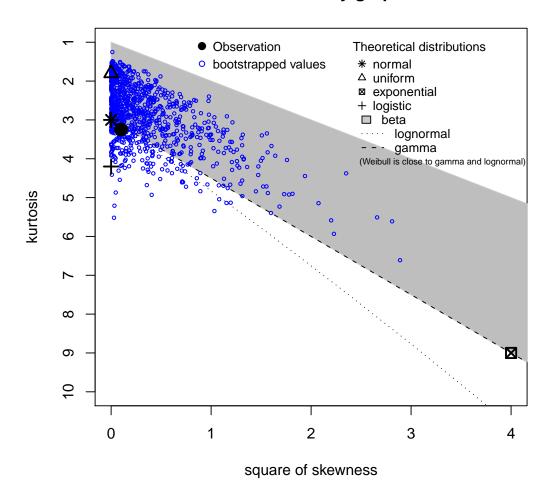
-----

min: 1.3 max: 22.1

median: 10.2 mean: 10.4 sample sd: 4.75

sample skewness: 0.314
sample kurtosis: 3.25

#### **Cullen and Frey graph**



#### 1.3 Fitting of a distribution

One or more parametric distributions may then be fitted to the data set, one at a time, using the fonction fitdist. This function uses the maximum likelihood method if the argument method="mle" (or if it is omitted) or the matching moments estimation if the argument method="mme". When fitting continuous distributions, Kolmogorov-Smirnov and Anderson-Darling statistics are computed and corresponding tests are performed when possible. Even if less appropriate for continuous distributions, the Chi-squared statistic is also computed when possible. For this calculation, cells are defined by the argument chisqbreaks or automatically defined from the data set and from the argument meancount (the approximate mean count per cell) which is fixed to  $(4n)^{2/5}$  if omitted (with n the length of the data set). For more details, see the help of the function fitdist. Four goodness of fit plots are also provided.

Below is the result of a fit of a gamma distribution by maximum likelihood.

```
> f1g <- fitdist(x1, "gamma")</pre>
> plot(f1g)
> summary(f1g)
FITTING OF THE DISTRIBUTION ' gamma ' BY MAXIMUM LIKELIHOOD
PARAMETERS
      estimate Std. Error
         3.575
shape
                     1.140
         0.343
                     0.118
rate
                         AIC:
Loglikelihood:
               -54.4
                               113
                                     BIC:
                                          115
Correlation matrix:
      shape rate
shape 1.000 0.931
rate 0.931 1.000
GOODNESS-OF-FIT STATISTICS
```

\_\_\_\_\_ Chi-squared\_\_\_\_\_

Chi-squared statistic: 7.93

Degree of freedom of the Chi-squared distribution: 3

Chi-squared p-value: 0.0475 !!! the p-value may be wrong

with some theoretical counts < 5 !!!

!!! For continuous distributions, Kolmogorov-Smirnov and Anderson-Darling statistics should be prefered !!!

\_\_\_\_\_Kolmogorov-Smirnov\_\_\_\_\_

Kolmogorov-Smirnov statistic: 0.138 Kolmogorov-Smirnov test: not calculated

\_\_\_\_\_ Anderson-Darling\_\_\_\_\_

Empirical and theoretical distr.

Anderson-Darling statistic: 0.457 Anderson-Darling test: not rejected

## Density 0.08

10

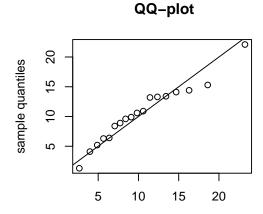
15

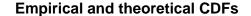
20

25

0

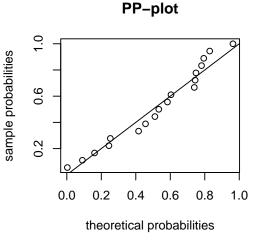
5





data

## 0.5 0 5 10 15 20 25 data



theoretical quantiles

Below is the result of another fit of the same distribution by matching moments.

- > f1gbis <- fitdist(x1, "gamma", method = "mme")</pre>
- > summary(f1gbis)

CDF

FITTING OF THE DISTRIBUTION ' gamma ' BY MATCHING MOMENTS PARAMETERS  $% \left( 1\right) =\left( 1\right) \left( 1\right$ 

estimate shape 4.810 rate 0.462

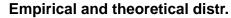
5

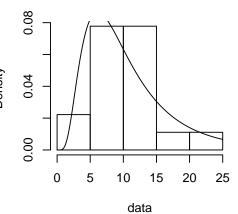
```
_____ Chi-squared_____
Chi-squared statistic: 7.27
Degree of freedom of the Chi-squared distribution: 3
Chi-squared p-value: 0.0637
!!! the p-value may be wrong
                                                with some theoretical counts < 5 !!!
!!! For continuous distributions, Kolmogorov-Smirnov and
      Anderson-Darling statistics should be prefered !!!
_____ Kolmogorov-Smirnov_____
Kolmogorov-Smirnov statistic: 0.144
Kolmogorov-Smirnov test: not calculated
_____ Anderson-Darling_____
Anderson-Darling statistic: 0.471
Anderson-Darling test: not rejected
  As can be seen in this returned summary, the automatic definition of the cells required to calculate the Chi-squared
statistic does not give theoretical counts large enough to validate the use of the test in this example. It is often the
case for small data sets. The observed and theoretical counts may be printed as below:
> f1g$chisqtable
       obscounts theocounts
       3.000 2.895
<= 5.2
<= 8.4 3.000 4.596
<= 9.9 3.000 2.108
<= 13.2 3.000
                    3.706
<= 14.1 3.000
> 14.1 3.000
                  0.758
3.936
  Below is the fit of a lognormal distribution.
> f11 <- fitdist(x1, "lnorm")
> plot(f11)
> summary(f11)
FITTING OF THE DISTRIBUTION ' lnorm ' BY MAXIMUM LIKELIHOOD
PARAMETERS
      estimate Std. Error
meanlog 2.197 0.147
sdlog 0.622 0.104
Loglikelihood: -56.5 AIC: 117 BIC: 119
Correlation matrix:
       meanlog sdlog
meanlog
          1 0
             0
sdlog
GOODNESS-OF-FIT STATISTICS
_____ Chi-squared_____
Chi-squared statistic: 11.1
Degree of freedom of the Chi-squared distribution: 3
Chi-squared p-value: 0.0110
!!! the p-value may be wrong
                                                with some theoretical counts < 5 !!!
!!! For continuous distributions, Kolmogorov-Smirnov and
      Anderson-Darling statistics should be prefered !!!
_____Kolmogorov-Smirnov_____
```

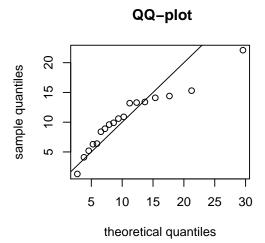
Kolmogorov-Smirnov statistic: 0.178

\_\_\_\_\_ Anderson-Darling\_\_\_\_\_

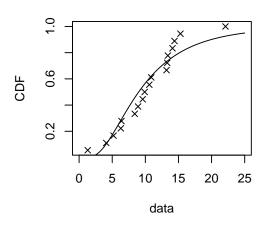
Anderson-Darling statistic: 0.793 Anderson-Darling test: rejected

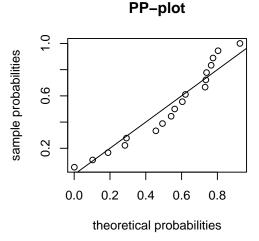






#### **Empirical and theoretical CDFs**





Below is the fit of a normal distribution.

- > f1n <- fitdist(x1, "norm")
- > plot(f1n)
- > summary(f1n)

FITTING OF THE DISTRIBUTION ' norm ' BY MAXIMUM LIKELIHOOD

 ${\tt PARAMETERS}$ 

estimate Std. Error

mean 10.41 1.119

sd 4.75 0.791

Loglikelihood: -53.6 AIC: 111 BIC: 113

Correlation matrix:

mean sd mean 1 0 sd 0 1

\_\_\_\_\_

GOODNESS-OF-FIT STATISTICS

\_\_\_\_\_ Chi-squared\_\_\_\_\_

Chi-squared statistic: 4.83

Degree of freedom of the Chi-squared distribution: 3

Chi-squared p-value: 0.185

#### !!! For continuous distributions, Kolmogorov-Smirnov and Anderson-Darling statistics should be prefered !!!

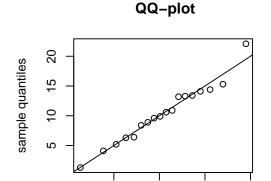
Kolmogorov-Smirnov statistic: 0.110
Kolmogorov-Smirnov test: not calculated

\_\_\_\_\_ Anderson-Darling\_\_\_\_\_

Anderson-Darling statistic: 0.226 Anderson-Darling test: not rejected

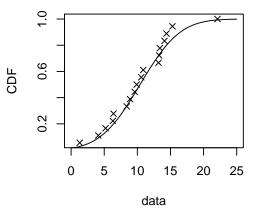
#### **Empirical and theoretical distr.**

# Density 0.0 0 5 10 15 20 25 data



5

#### **Empirical and theoretical CDFs**



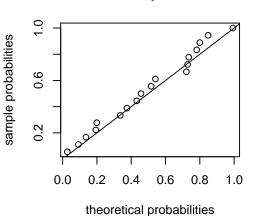
#### PP-plot

10

theoretical quantiles

20

15



Below is the fit of a Weibull distribution.

- > f1w <- fitdist(x1, "weibull")</pre>
- > plot(f1w)
- > summary(f1w)

FITTING OF THE DISTRIBUTION ' weibull ' BY MAXIMUM LIKELIHOOD PARAMETERS

estimate Std. Error pe 2.29 0.426

shape 2.29 0.426 scale 11.70 1.264

Loglikelihood: -53.5 AIC: 111 BIC: 113

Correlation matrix:

shape scale

shape 1.0 0.3 scale 0.3 1.0

#### \_\_\_\_

#### GOODNESS-OF-FIT STATISTICS

\_\_\_\_\_ Chi-squared\_\_\_\_\_

Chi-squared statistic: 5.87

Degree of freedom of the Chi-squared distribution: 3

Chi-squared p-value: 0.118 !!! the p-value may be wrong

with some theoretical counts < 5 !!!

!!! For continuous distributions, Kolmogorov-Smirnov and Anderson-Darling statistics should be prefered !!!

\_\_\_\_\_ Kolmogorov-Smirnov\_\_\_\_\_

Kolmogorov-Smirnov statistic: 0.121 Kolmogorov-Smirnov test: not calculated

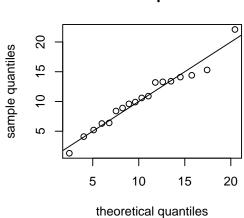
\_\_\_\_\_ Anderson-Darling\_\_\_\_\_

Anderson-Darling statistic: 0.282 Anderson-Darling test: not rejected

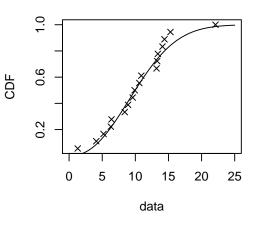
#### Empirical and theoretical distr.

# Density 0.0 0 5 10 15 20 25 data

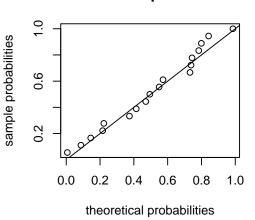
#### QQ-plot



#### **Empirical and theoretical CDFs**



#### PP-plot



The values of the Anderson-Darling statistic (or another result of the fit: see the help of fitdist for details) for the different fittings may be extracted and compared to help the selection of a distribution:

- > anderson <- list(lnorm = f11\$ad, gamma = f1g\$ad, norm = f1n\$ad,
- + weibull = f1w\$ad)
- > anderson

\$lnorm

[1] 0.793

#### \$gamma

[1] 0.457

#### \$norm

[1] 0.226

#### \$weibull

[1] 0.282

For some distributions (see the help of fitdist for details), it is necessary to specify initial values for the distribution parameters in the argument start when using the maximum likelihood method. start must be a named list of parameters initial values. The names of the parameters in start must correspond exactly to their definition in R or to their definition in a previous R code. The function plotdist may help to find correct initial values for the distribution parameters in non trivial cases, by an manual iterative use if necessary.

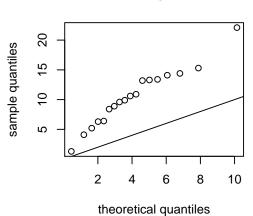
For example, below is the definition of the Gumbel distribution (also named extreme value distribution) and a first plot of the data set with the Gumbel distribution with arbitrary values for parameters.

```
> dgumbel \leftarrow function(x, a, b) 1/b * exp((a - x)/b) * exp(-exp((a - x)/b))
> pgumbel \leftarrow function(q, a, b) exp(-exp((a - q)/b))
> qgumbel \leftarrow function(p, a, b) a - b * log(-log(p))
> plotdist(x1, "gumbel", para = list(a = 3, b = 2))
```

#### Empirical and theoretical distr.

#### 

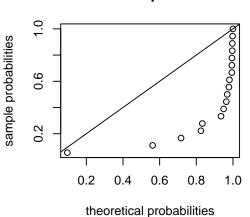
#### QQ-plot



#### **Empirical and theoretical CDFs**

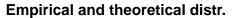
### 0: 9: 0 5 10 15 20 25 data

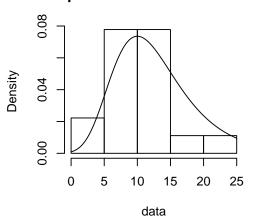
#### PP-plot



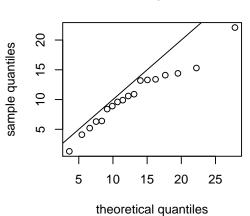
The same data set may be plotted with a Gumbel distribution with modified values for parameters.

> plotdist(x1, "gumbel", para = list(a = 10, b = 5))

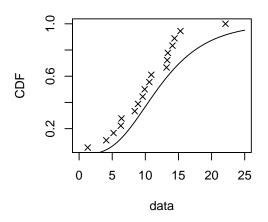




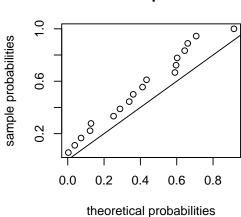
#### QQ-plot



#### **Empirical and theoretical CDFs**



#### PP-plot



And a Gumbel distribution may be fitted to data with these values for initial parameter values.

```
> fgu \leftarrow fitdist(x1, "gumbel", start = list(a = 10, b = 5))
```

> plot(fgu)

> summary(fgu)

FITTING OF THE DISTRIBUTION ' gumbel ' BY MAXIMUM LIKELIHOOD

PARAMETERS

estimate Std. Error a 8.09 1.092 b 4.38 0.766

Loglikelihood: -54.1 AIC: 112 BIC: 114

Correlation matrix:

a 1.000 0.330 b 0.330 1.000

-----

GOODNESS-OF-FIT STATISTICS

\_\_\_\_\_Chi-squared\_\_\_\_\_

Chi-squared statistic: 7.56

Degree of freedom of the Chi-squared distribution: 3

Chi-squared p-value: 0.056 !!! the p-value may be wrong

with some theoretical counts < 5 !!!

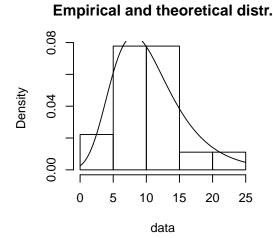
!!! For continuous distributions, Kolmogorov-Smirnov and Anderson-Darling statistics should be prefered !!!

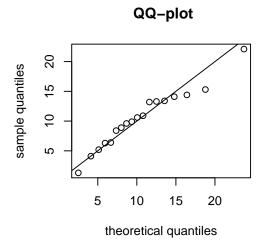
\_\_\_\_\_Kolmogorov-Smirnov\_\_\_\_\_

Kolmogorov-Smirnov statistic: 0.121 Kolmogorov-Smirnov test: not calculated

\_\_\_\_\_ Anderson-Darling\_\_\_\_\_

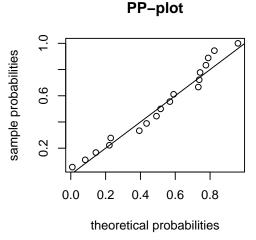
Anderson-Darling statistic: 0.34 Anderson-Darling test: not calculated





#### 

**Empirical and theoretical CDFs** 



#### 1.4 Simulation of the uncertainty by boostrap

The uncertainty in the parameters of the fitted distribution may be simulated by parametric or nonparametric boostrap using the function boodist. This function returns the boostrapped values of parameters which may be plotted to visualize the bootstrap region. It also calculates the 95 percent confidence intervals for each parameter from the 2.5 and 97.5 percentiles of the boostrap values of each parameter (see the help of the function bootdist for details).

Below is an example of the use of this function with the previous fit of the gamma distribution.

- > b1g <- bootdist(f1g)</pre>
- > plot(b1g)
- > summary(b1g)

Parametric bootstrap medians and 95% percentile CI

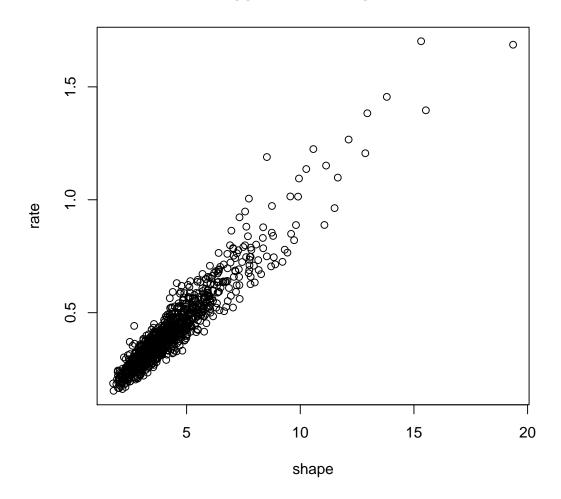
Median 2.5% 97.5%

shape 3.941 2.223 8.437

rate 0.387 0.204 0.839

Maximum likelihood method converged for 1001 among 1001 iterations

#### **Boostrapped values of parameters**

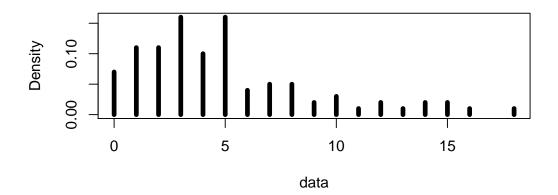


#### 2 Specification of a distribution from non-censored discrete data

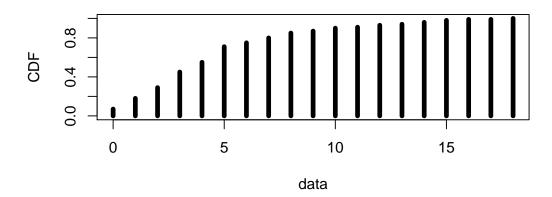
A discrete data set may be considered as a continuous one for example for a large data set from a binomial distribution converging to a normal one. A discrete plot of the distribution may also be provided, fixing the argument discrete of the function plotdist to TRUE.

```
> x2 <- rnbinom(n = 100, size = 2, prob = 0.3)
> plotdist(x2, discrete = TRUE)
```

#### **Empirical distribution**



#### **Empirical CDFs**



As for continuous distributions, descriptive parameters of the empirical distribution may be computed using the function descdist which also provides a skewness-kurtosis plot which may help you to choose which distribution(s) to fit.

#### > descdist(x2, discrete = T)

#### summary statistics

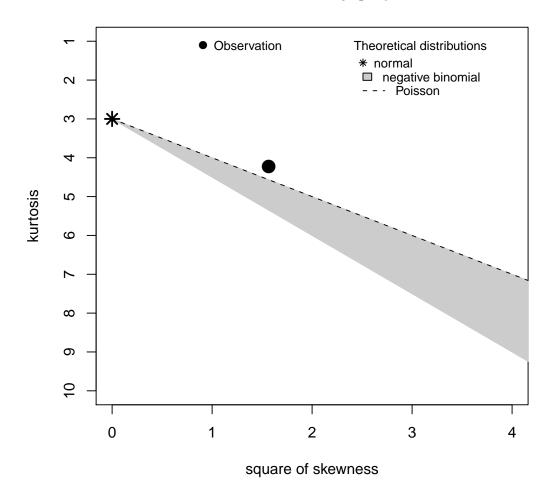
-----

min: 0 max: 18

median: 4
mean: 4.88
sample sd: 3.94
sample skewness: 1.25

sample kurtosis: 4.22

#### **Cullen and Frey graph**



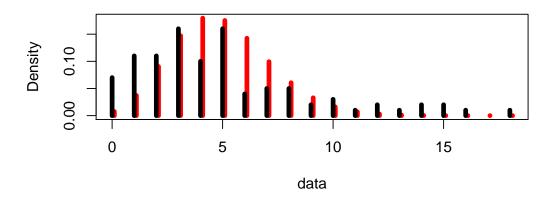
As for continuous distributions, one or more parametric distributions may then be fitted to the data set by maximum likelihood or matching moments.

Below is the result of the fit of a Poisson distribution with the bootstrap simulations.

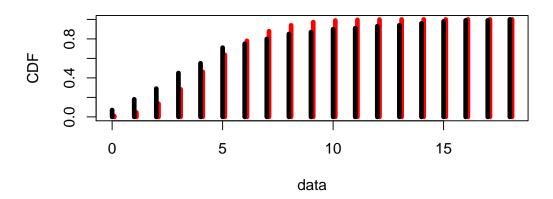
```
> f2p <- fitdist(x2, "pois")</pre>
> plot(f2p)
> summary(f2p)
FITTING OF THE DISTRIBUTION ' pois ' BY MAXIMUM LIKELIHOOD
PARAMETERS
       estimate Std. Error
lambda
           4.88
                     0.221
Loglikelihood: -305
                       AIC:
                             613
                                   BIC: 615
GOODNESS-OF-FIT STATISTICS
   _____ Chi-squared_
Chi-squared statistic: 122
Degree of freedom of the Chi-squared distribution: 6
Chi-squared p-value: 6.73e-24
!!! the p-value may be wrong
                                                  with some theoretical counts < 5 !!!
> b2p <- bootdist(f2p)
> summary(b2p)
Parametric bootstrap medians and 95% percentile CI
Median
         2.5% 97.5%
  4.89
         4.49
                5.31
```

Maximum likelihood method converged for 1001 among 1001 iterations

#### Empirical (black) and theoretical (red) distr.



#### **Empirical (black) and theoretical (red) CDFs**



Below is the result of the fit of a negative binomial distribution with the boostrap simulations.

```
> f2n <- fitdist(x2, "nbinom")</pre>
```

FITTING OF THE DISTRIBUTION ' nbinom ' BY MAXIMUM LIKELIHOOD **PARAMETERS** 

estimate Std. Error 2.31 0.501 size 4.88 0.389

AIC: 524 BIC: 529 Loglikelihood: -260

Correlation matrix:

size size 1.000000 0.000101 0.000101 1.000000

GOODNESS-OF-FIT STATISTICS

\_\_\_\_\_ Chi-squared\_\_\_\_\_

Chi-squared statistic: 8.12

Degree of freedom of the Chi-squared distribution: 5

Chi-squared p-value: 0.150

> b2n <- bootdist(f2n)

> summary(b2n)

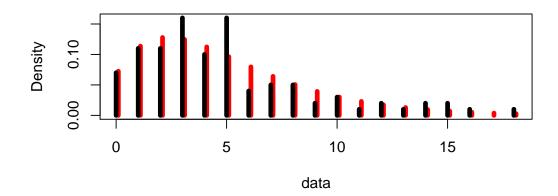
Parametric bootstrap medians and 95% percentile CI Median 2.5% 97.5% size 2.40 1.58 3.83

<sup>&</sup>gt; plot(f2n)

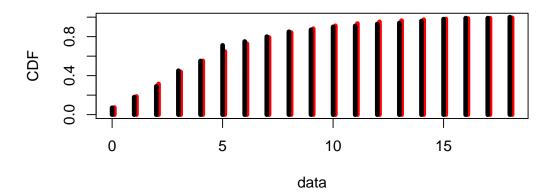
<sup>&</sup>gt; summary(f2n)

Maximum likelihood method converged for 1001 among 1001 iterations

#### Empirical (black) and theoretical (red) distr.



#### Empirical (black) and theoretical (red) CDFs



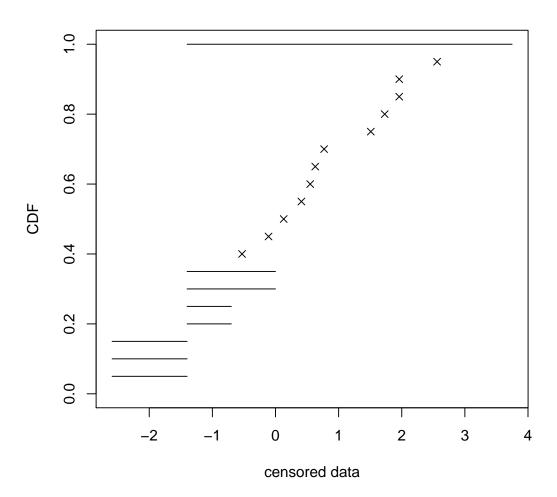
#### 3 Specification of a distribution from censored data

Censored data may contain left censored, right censored and interval censored values, with several lower and upper bounds. Data must be coded into a dataframe with two columns, respectively named left and right, describing each observed value as an interval. The left column contains either NA for left censored observations, the left bound of the interval for interval censored observations, or the observed value for non-censored observations. The right column contains either NA for right censored observations, the right bound of the interval for interval censored observations, or the observed value for non-censored observations.

#### 3.1 Graphical display of the observed distribution

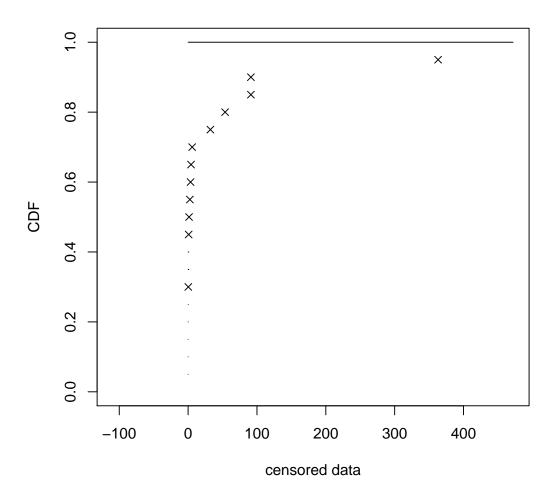
First of all, the observed distribution may be plotted using the function plotdistcens. Data are reported directly as segments for interval, left and right censored data, and as points for non-censored data. For more details, see the help of the function plotdistcens.

```
> d1 < -data.frame(left = c(1.73, 1.51, 0.77, 1.96, 1.96, -1.4, + -1.4, NA, -0.11, 0.55, 0.41, 2.56, NA, -0.53, 0.63, -1.4, + -1.4, -1.4, NA, 0.13), right = c(1.73, 1.51, 0.77, 1.96, + 1.96, 0, -0.7, -1.4, -0.11, 0.55, 0.41, 2.56, -1.4, -0.53, + 0.63, 0, -0.7, NA, -1.4, 0.13)) <math>> plotdistcens(d1)
```



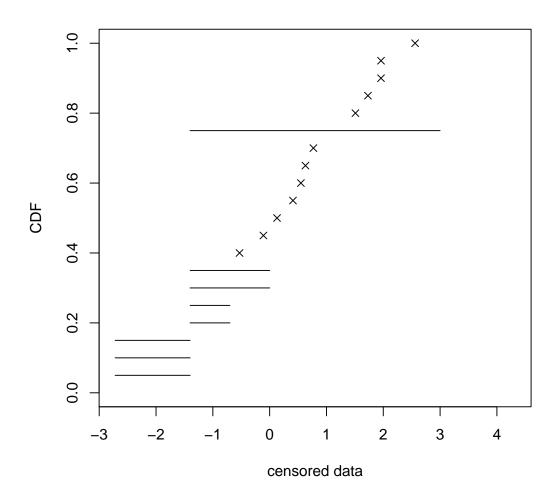
When left or right NA-values correspond to finite value (for example 0 for left NA-values of positive data), the arguments leftNA (or rightNA) must be affected to this finite value to ensure a correct plot of left (or right) censored observations, as in the example below.

```
> d2 <- data.frame(left = 10^{(d1\$left)}, right = 10^{(d1\$right)}) > plotdistcens(d2, leftNA = 0)
```



It is also possible to fix rightNA or leftNA to a realistic extreme value, even if not exactly known, to obtain a reasonable global ranking of observations, as in the example below for the first dataset.

> plotdistcens(d1, rightNA = 3)

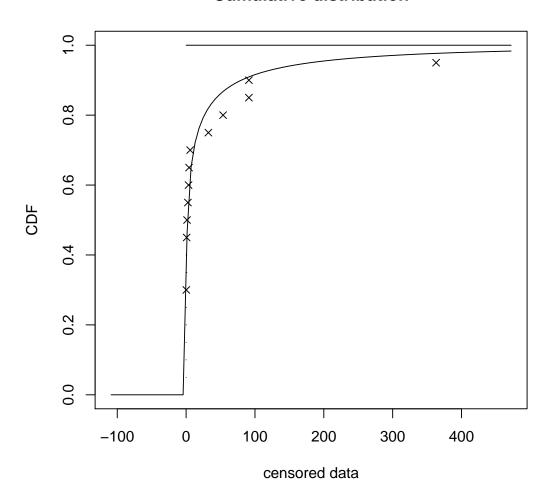


#### 3.2 Fitting of a distribution

One or more parametric distributions may then be fitted to the censored data set, one at a time, using the fonction fitdistcens. This function always uses the maximum likelihood method. For more details, see the help of the function fitdistcens. Only one goodness of fit plot is provided for censored data, in cumulative frequencies. The uncertainty in the parameters of the fitted distribution may be simulated by nonparametric boostrap only, using the function boodistcens.

Below is the result of a fit of a Weibull distribution by maximum likelihood and the results of the corresponding boostrap simulations.

```
> f2w <- fitdistcens(d2, "weibull")</pre>
> summary(f2w)
FITTING OF THE DISTRIBUTION ' weibull ' BY MAXIMUM LIKELIHOOD ON CENSORED DATA
PARAMETERS
      estimate Std. Error
         0.324
                   0.0613
shape
scale
         6.124
                   4.5872
Loglikelihood:
                -68.5
                         AIC:
                               141
                                     BIC: 143
Correlation matrix:
      shape scale
shape 1.000 0.326
scale 0.326 1.000
> plot(f2w, leftNA = 0)
```

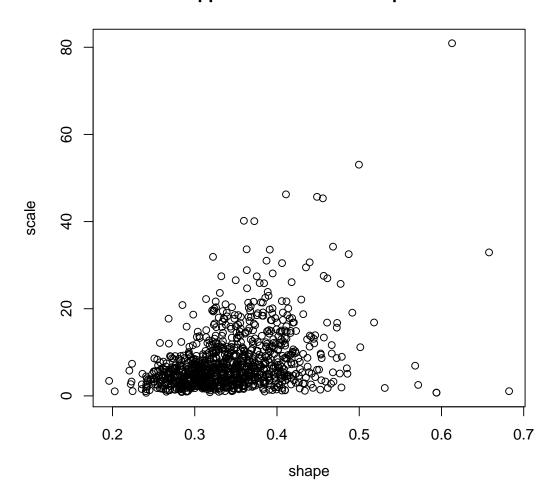


- > b2w <- bootdistcens(f2w)
- > summary(b2w)

Nonparametric bootstrap medians and 95% percentile CI Median 2.5% 97.5% shape 0.336 0.243 0.473 scale 5.603 1.221 23.296

Maximum likelihood method converged for 1001 among 1001 iterations > plot(b2w)

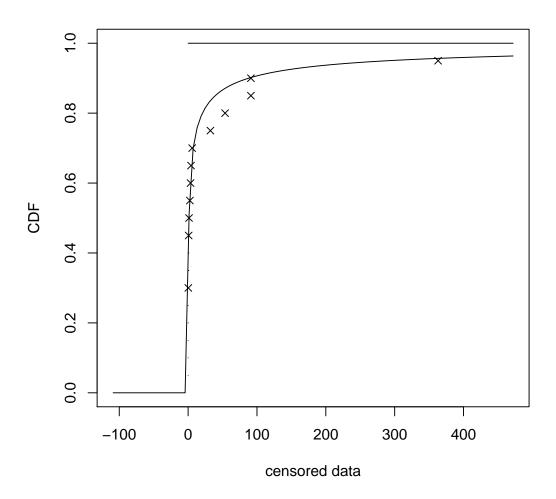
#### Boostrapped values of the two parameters



Goodness of fit statistics are not computed for fit on censored data, so the quality of fit may only be estimated from the loglikelihood and the goodness of fit plot.

Below is the fit of a lognormal distribution to the same censored data set.

```
> f21 <- fitdistcens(d2, "lnorm")</pre>
> summary(f21)
FITTING OF THE DISTRIBUTION ' lnorm ' BY MAXIMUM LIKELIHOOD ON CENSORED DATA
PARAMETERS
        estimate Std. Error
meanlog
            0.27
                      0.764
sdlog
            3.28
                      0.600
Loglikelihood: -68.7
                        AIC:
                              141
                                     BIC: 143
Correlation matrix:
        meanlog
                  sdlog
meanlog 1.0000 -0.0739
sdlog
        -0.0739 1.0000
> plot(f21, leftNA = 0)
```



Below is the fit of an exponential distribution.

```
> f2e <- fitdistcens(d2, "exp")</pre>
```

> summary(f2e)

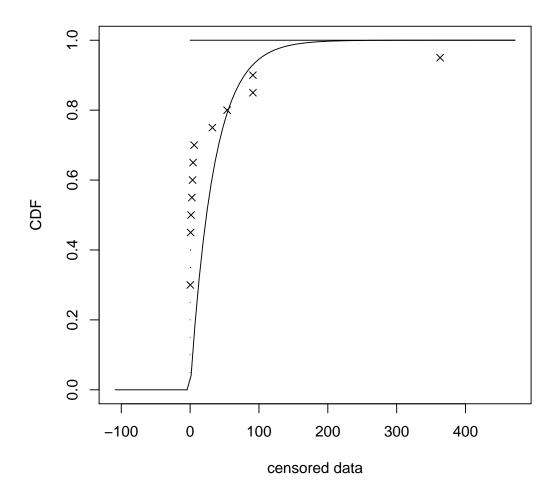
FITTING OF THE DISTRIBUTION '  $\exp$  ' BY MAXIMUM LIKELIHOOD ON CENSORED DATA PARAMETERS

estimate Std. Error

rate 0.0292 0.00668

Loglikelihood: -99.6 AIC: 201 BIC: 202

> plot(f2e, leftNA = 0)



As with fitdist, for some distributions (see the help of fitdistcens for details), it is necessary to specify initial values for the distribution parameters in the argument start. start must be a named list of parameters initial values. The names of the parameters in start must correspond exactly to their definition in R or to their definition in a previous R code. The function plotdistcens may help to find correct initial values for the distribution parameters in non trivial cases, by an manual iterative use if necessary, as explained previously for non-censored continuous data.