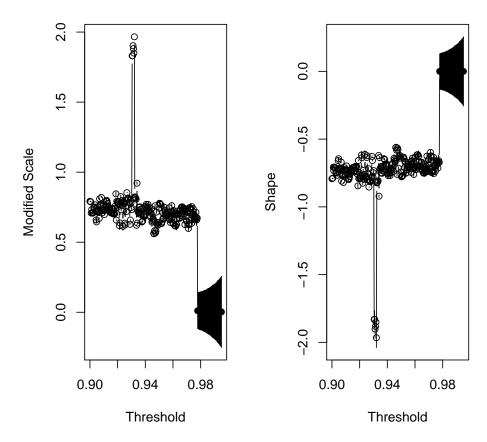
1.6 Legalese

This program is free software; you can redistribute it and/or modify it under the terms of the GNU

5. Two raFdom variables (with unit Fréchet margins) are perfectly depeFdeFt if ${\cal A}$

3.2 Threshold Selection



Mean Residual Life Plot

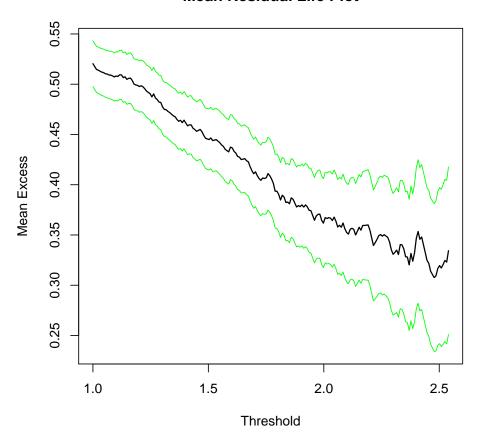


Figure 2: The threshold selection using the mrlplot function

The quantity $E[X - \mu_1/X > \mu_1]$ is linear in μ_1 . Or, $E[X - \mu_1/X > \mu_1]$ is simply the mean of excesses above the threshold μ_1

3.2.3 L-Moments plot: *Imomplot*

L-moments are summary statistics for probability distributions and data samples. They are analogous to ordinary moments – they provide measures of location, (i)1(on)eeersion, skewness, kurtosis, and other aeeects of the shaee of probability distributions or data samples – but are computed from linear combinations of the ordered data values (hence the prefix L).

For the GPD, the following relation holds:

L-Moments Plot

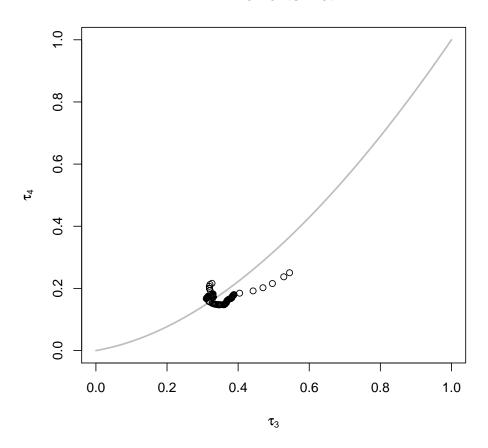


Figure 3: fig: The threshold selection using the Imomplot function

3.3 Fitting the GPD

3.3.1 The univariate case

The main function to fit the GPD is called **fitgpd**. This is a generic function which can fit the GPD according several estimators. There are currently 7 estimators available: method of moments moments, maximum likelihood ml e

Standard Error Type: observed

Standard Errors scale 0.2083

Asymptotic Variance Covariance scale scale 0.04337

Optimization Information Convergence: successful Function Evaluations: 6 Gradient Evaluations: 1 Threshold Call: c(1, 2) Number Above: 500 Proportion Above: 1

Estimates

scal e shape 0. 29500 -0. 02758

Standard Error Type: observed

Standard Errors scale shape 0.01912 0.04692

Asymptotic Variance Covariance scale shape scale 0.0003654 -0.0006651 shape -0.0006651 0.0022010

Optimization Information Convergence: successful Function Evaluations: 45 Gradient Evaluations: 11 scal e1 shape1 scal e2 shape2 al pha

Pickands' Dependence Function

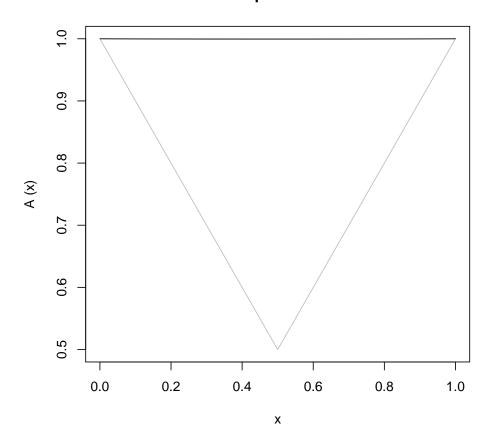


Figure 5: The Pickands' dependence function

Optimization Information Convergence: successful Function Evaluations: 6 Gradient Evaluations: 1

```
> x < - rgpd(200, 1, 2, 0.25)
> mle <- fitgpd(x, 1, method = "mle")
> mom <- fitgpd(x, 1, method = "moments")
> pwmb < - fitgpd(x, 1, method = "pwmb")
> pwmu < - fitgpd(x, 1, method = "pwmu")
> gpd. fiscale(mle, conf = 0.9)
conf.inf.scale conf.sup.scale
      1.400080
                     2. 103698
> gpd.fiscale(mom, conf = 0.9)
conf.inf.scale conf.sup.scale
           NaN
                          NaN
> gpd.fiscale(pwmu, conf = 0.9)
conf.inf.scale conf.sup.scale
      1. 399238
                     2.099229
```

> gpd. fi scale(pwmb, conf = 0.9)

2. 111391

conf.inf.scale conf.sup.scale

1. 410214

```
If there is some troubles try to put vert.lines = FALSE or change
  the range...
conf.inf conf.sup
1.431818 2.122727
```

If there is some troubles try to put vert.lines = FALSE or change
the range...
conf.inf conf.sup
0.2454545 0.5424242

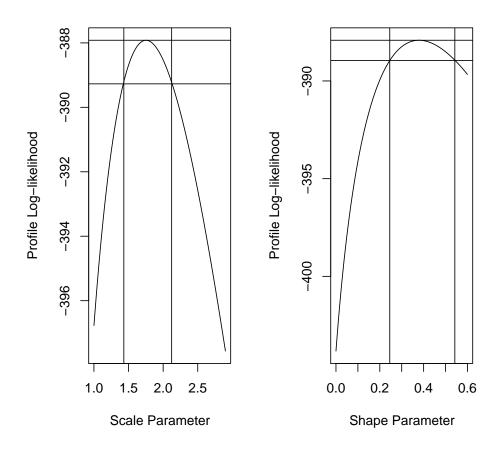
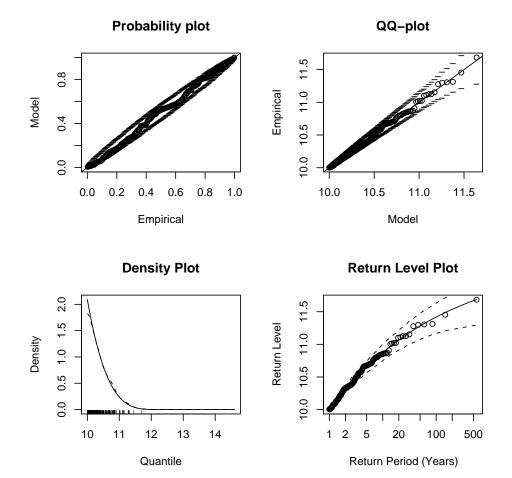
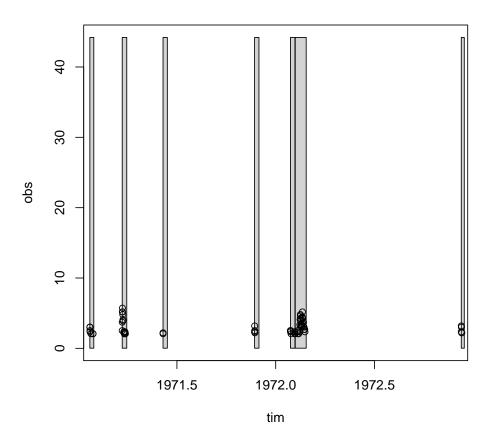
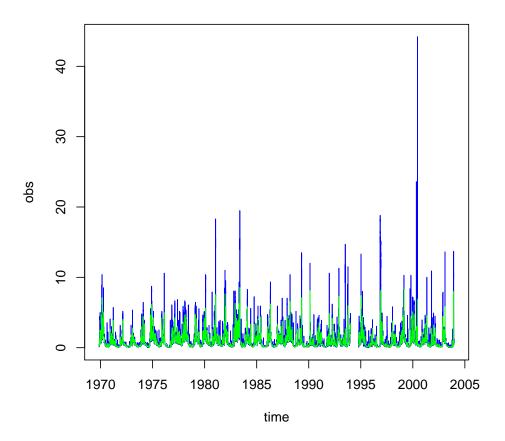


Figure 6: The profile log-likelihood confidence intervals

conf.inf conf.sup







time obs Min. :1970 Min. : 0.022 1st Qu.:1981 1st Qu.: 0.236

The result of function **fitgpd** gives the name of the estimator, if a varying threshold was used, the threshold value, the number and the proportion of observations above the threshold, parameter estimates, standard error estimates and type, the asymptotic variance-covariance matrix and convergence diagnostic.

Figure 12 shows graphic diagnostics for the fitted model. It can be seen that the fitted model "mle" seems to be appropriate. Suppose we want to know the return level associated to the 100-year return period.

To take into account uncertainties, Figure 13 depicts the profile confidence interval for the quantile associated to the 100-year return period.

```
> gpd.pfrl(mle, prob, range = c(25, 100), nrang = 200)

If there is some troubles try to put vert.lines = FALSE or change the range...
conf.inf conf.sup
25.56533 90.76633
```

Sometimes it is necessary to know the estimated return period of a specified events. Lets do it with the larger events in "events1".