

# Modeling extreme values with a GEV mixture probability distributions

Standard Gumbel distribution

Pascal Alain Dkengne Sielenou

2023-10-28

```
# Load useful functions
```

```
path <- ".."
```

```
xfun::in_dir(dir = path, expr = source("./src/generate_gev_sample.R"))
xfun::in_dir(dir = path, expr = source("./src/calculate_gev_inverse_cdf.R"))
xfun::in_dir(dir = path, expr = source("./src/estimate_gev_mixture_model_parameters.R"))
xfun::in_dir(dir = path, expr = source("./src/plot_gev_mixture_model_pdf.R"))
xfun::in_dir(dir = path, expr = source("./src/plot_gev_mixture_model_cdf.R"))
xfun::in_dir(dir = path, expr = source("./src/estimate_gev_mixture_model_quantile.R"))
```

```
# Generate a random sample
```

```
n <- 20000
```

```
loc <- 0
scale <- 1
shape <- 0
```

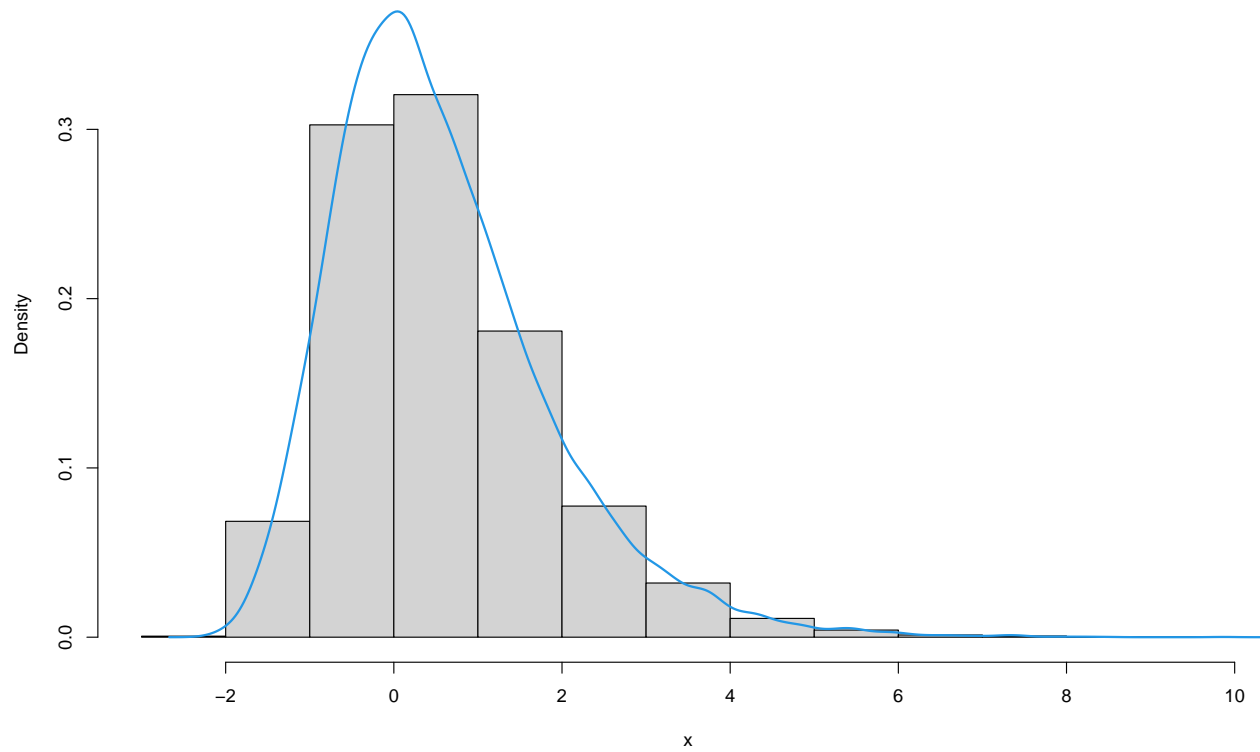
```
set.seed(1122)
```

```
x <- generate_gev_sample(n = n, loc = loc, scale = scale, shape = shape)
```

```
# Histogram of all data
```

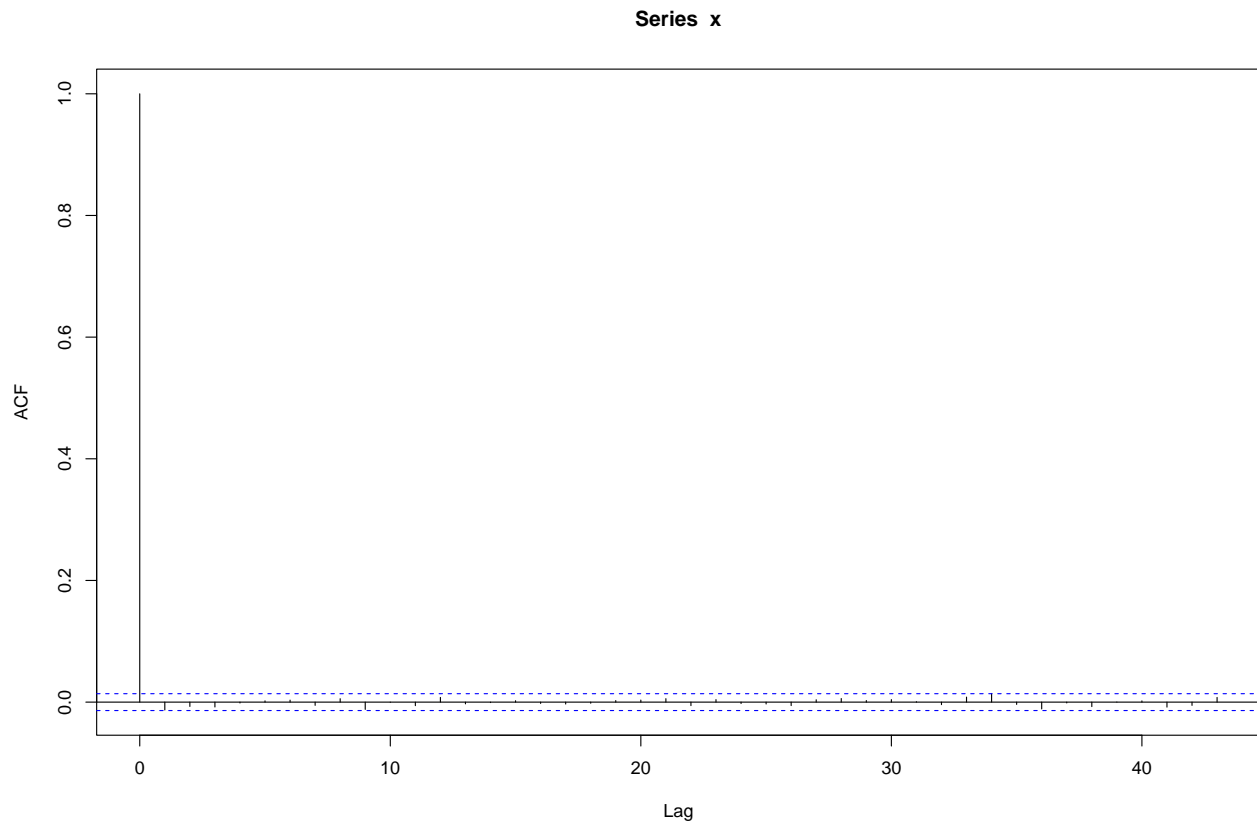
```
dens_x <- density(x)
hist(x, prob = TRUE, ylim = range(dens_x$y))
lines(dens_x, lwd = 2, col = 4)
```

Histogram of x



```
# Autocorrelation function of all data
```

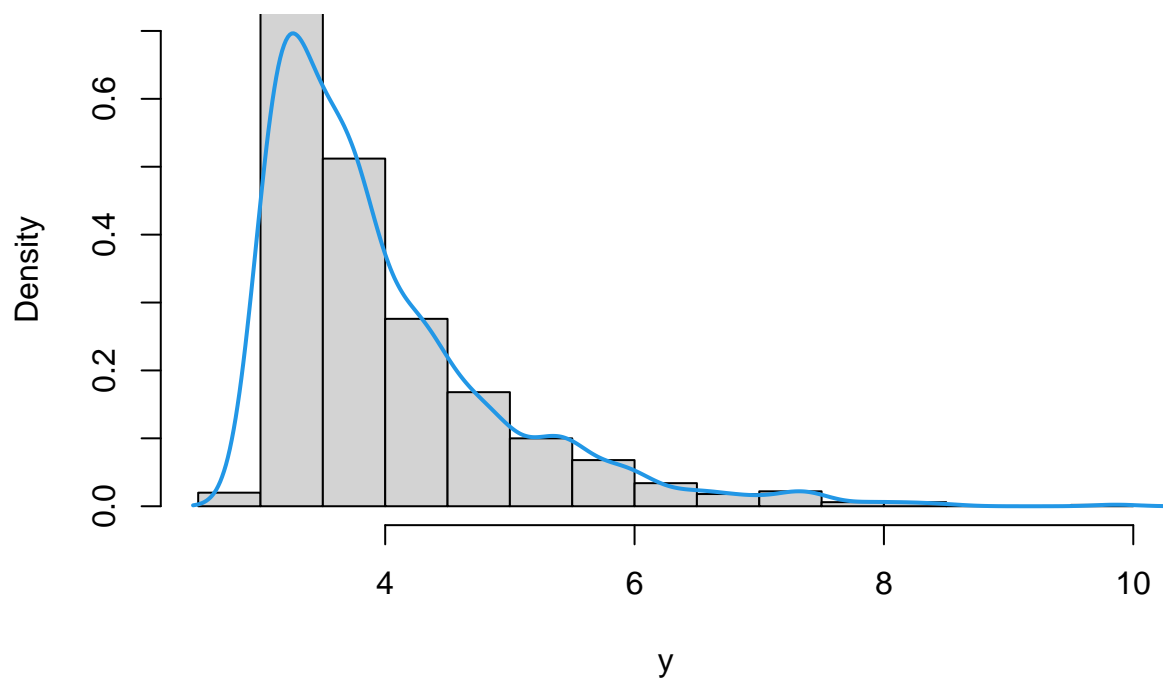
```
acf(x)
```



```
# Histogram of the largest data
```

```
nlargest <- 1000  
y <- extract_nlargest_sample(x, n = nlargest)  
dens_y <- density(y)  
hist(y, prob = TRUE, ylim = range(dens_y$y))  
lines(density(y), lwd = 2, col = 4)
```

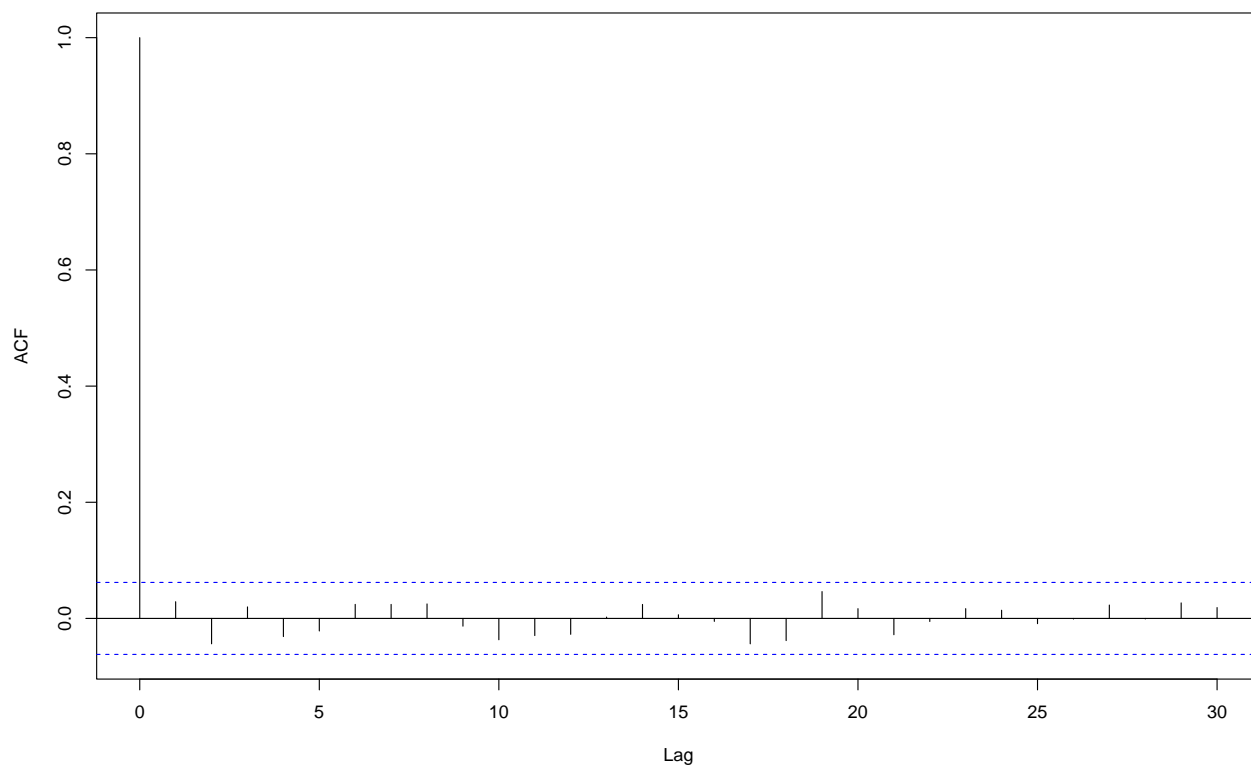
## Histogram of y



```
# Autocorrelation function of the largest data
```

```
acf(y)
```

## Series y



```
# Estimation of gev mixture models
```

```
gev_mixture_model <- estimate_gev_mixture_model_parameters(x = x,  
                                                           block_sizes = NULL,  
                                                           minimum_nblocks = 50,  
                                                           threshold = NULL,  
                                                           nlargest = nlargest,  
                                                           confidence_level = 0.95,  
                                                           use_extremal_index = TRUE,  
                                                           use_lower_threshold = FALSE,  
                                                           maximum_iterations = 1500,  
                                                           log_mv = TRUE,  
                                                           log_pw = TRUE,  
                                                           trace = FALSE,  
                                                           method = "MLE")
```

```
## Successful convergence.
```

```
## Successful convergence.
```

```
gev_mixture_model$extremal_indexes
```

```
##           13           14           15           16           17           18  
## 0.8920296849 0.9412167819 0.9109104376 0.9663117674 1.0000000000 1.0000000000  
##           19           20  
## 0.9153737685 1.0000000000
```

```
gev_mixture_model$normalized_gev_parameters_object
```

```
##      loc_star  scale_star  shape_star  
## 13 2.738730574 1.2972204786 -0.11094214484  
## 14 3.173500173 1.0537637674 -0.06604485185  
## 15 2.962237100 1.1338401059 -0.07956577765  
## 16 2.943030605 1.1978828474 -0.09487103167  
## 17 2.640938537 1.3340390618 -0.11772435182  
## 18 3.339054187 0.9284680982 -0.02948815858  
## 19 2.200619920 1.5943004373 -0.15367097316  
## 20 3.387462307 0.9167513605 -0.02821859821
```

```
gev_mixture_model$full_normalized_gev_parameters_object
```

```
##      loc_star  scale_star  shape_star  
## 13 2.589572170 1.3137684319 -0.11094214484  
## 14 3.109533392 1.0579884440 -0.06604485185  
## 15 2.856043969 1.1422894450 -0.07956577765  
## 16 2.901913848 1.2017836365 -0.09487103167  
## 17 2.640938537 1.3340390618 -0.11772435182  
## 18 3.339054187 0.9284680982 -0.02948815858  
## 19 2.058685279 1.6161116717 -0.15367097316  
## 20 3.387462307 0.9167513605 -0.02821859821
```

```
gev_mixture_model$automatic_weights_pw_shape
```

```
##           13           14           15           16           17           18  
## 0.1340587202 0.1178963789 0.1227637377 0.1282733987 0.1365001850 0.1057314779  
##           19           20  
## 0.1494401683 0.1053359333
```

```

gev_mixture_model$automatic_weights_pw_scale

##           13           14           15           16           17           18
## 0.1168023571 0.1315041849 0.1253806495 0.1224047623 0.1157881200 0.1427175822
##           19           20
## 0.1016704514 0.1437318925

gev_mixture_model$automatic_weights_pw_loc

##           13           14           15           16           17           18
## 0.1161439230 0.1315013058 0.1229014897 0.1244577977 0.1169475723 0.1392863559
##           19           20
## 0.1078334539 0.1409281017

gev_mixture_model$weighted_normalized_gev_parameters_object[3, ]

##           loc_star  scale_star  shape_star
## automatic_weights 2.95437791 1.159715179 -0.08972070362

gev_mixture_model$automatic_weights_mw

##           13           14           15           16           17
## 0.14900807916 0.30984800794 0.33393783784 0.16664609783 0.04055997724
##           18           19           20
## 0.00000000000 0.00000000000 0.00000000000

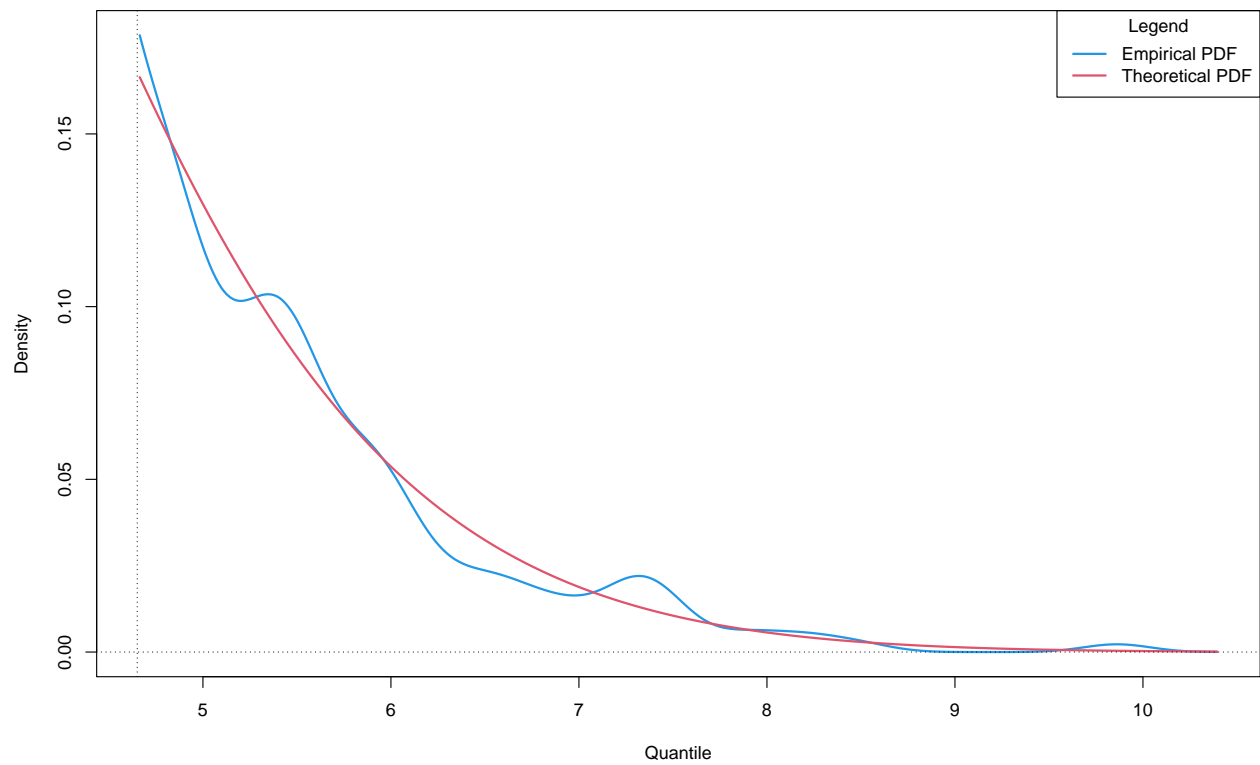
# Model diagnostics

## GEV mixture model with respect to parameters

plot_gev_mixture_model_pdf(gev_mixture_model,
  type = "automatic_weights",
  model_wise = FALSE,
  zoom = TRUE,
  xlab = "Quantile",
  ylab = "Density",
  main = "Probability Density Function (PDF) Plot")

```

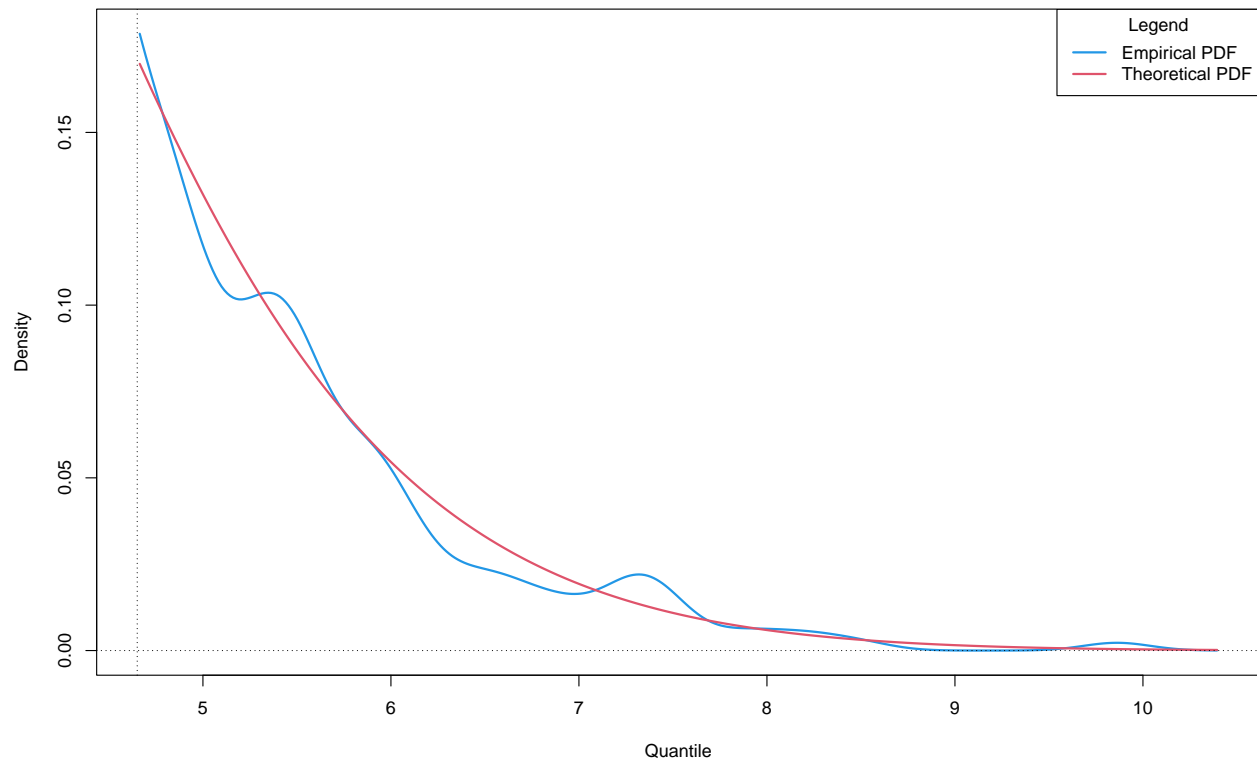
Probability Density Function (PDF) Plot : automatic\_weights – model\_wise = FALSE : zoom = TRUE



```
## GEV mixture model with respect to distribution functions
```

```
plot_gev_mixture_model_pdf(gev_mixture_model,  
                             type = "automatic_weights",  
                             model_wise = TRUE,  
                             zoom = TRUE,  
                             xlab = "Quantile",  
                             ylab = "Density",  
                             main = "Probability Density Function (PDF) Plot")
```

Probability Density Function (PDF) Plot : automatic\_weights – model\_wise = TRUE : zoom = TRUE



```
# Estimation of an extreme quantile
```

```
estimator_types <- c("automatic_weights_mw",
                     "pessimistic_weights_mw",
                     "identic_weights_mw",
                     "automatic_weights_pw",
                     "pessimistic_weights_pw",
                     "identic_weights_pw",
                     "model_wise",
                     "parameter_wise",
                     "empirical")
```

```
alpha <- 10(-14)
```

```
## Quantile from the true distribution
```

```
true_rl <- calculate_gev_inverse_cdf(p = 1 - alpha,
                                   loc = loc,
                                   scale = scale,
                                   shape = shape)
```

```
true_rl
```

```
## [1] 32.2369909
```

```
## Quantile from GEV mixture model with respect to parameters
```

```
rl_pw <- estimate_gev_mixture_model_quantile(gev_mixture_model,
                                             alpha = alpha,
                                             confidence_level = 0.95,
```



```

do.ci = TRUE,
estimator_type = estimator_types[4])

rl_pw

## [1] 14.94246142
## Quantile from GEV mixture model with respect to distribution functions

rl_mw <- estimate_gev_mixture_model_quantile(gev_mixture_model,
alpha = alpha,
confidence_level = 0.95,
do.ci = TRUE,
estimator_type = estimator_types[1])

rl_mw

## [1] 16.61948946
## Quantiles from equivalent estimated distributions in GEV mixture model with respect to parameters

est_rl_pw <- suppressWarnings(estimate_gev_mixture_model_quantile(gev_mixture_model,
alpha = alpha,
confidence_level = 0.95,
do.ci = TRUE,
estimator_type = estimator_types[8]))

est_rl_pw

##           lower    quantile    upper
## 13  2.3468996728 13.96958275 25.59226582
## 14 -3.7474302756 16.80635608 37.36014243
## 15 -2.1419903358 15.81091531 33.76382096
## 16 -0.5952710422 14.77892345 30.15311794
## 17  2.3700405431 13.61030101 24.85056147
## 18 -23.8227373416 21.53142934 66.88559603
## 19  3.6014287009 12.45779173 21.31415476
## 20 -26.7220637170 21.63967153 70.00140678
## Comparison of estimated quantiles

est_rl_pw_range <- range(as.matrix(est_rl_pw))

## Quantiles from equivalent estimated GEV distributions in GEV mixture model respect to distribution f

est_rl_mw <- suppressWarnings(estimate_gev_mixture_model_quantile(gev_mixture_model,
alpha = alpha,
confidence_level = 0.95,
do.ci = TRUE,
estimator_type = estimator_types[7]))

est_rl_mw

##           lower    quantile    upper
## 13  2.3468996728 13.96958275 25.59226582
## 14 -3.7474302756 16.80635608 37.36014243
## 15 -2.1419903358 15.81091531 33.76382096

```

```
## 16 -0.5952710422 14.77892345 30.15311794
## 17 2.3700405431 13.61030101 24.85056147

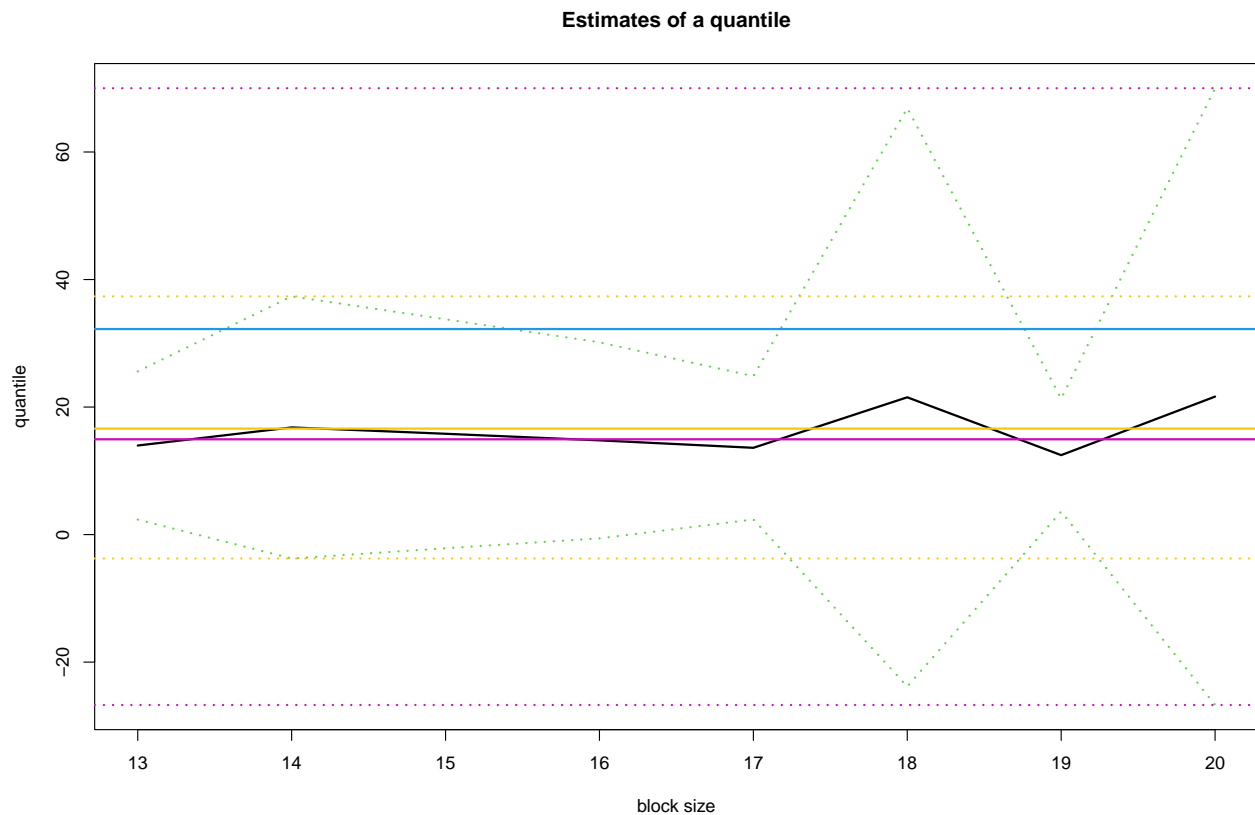
est_rl_mw_range <- range(as.matrix(est_rl_mw))

est_rl_mw_range

## [1] -3.747430276 37.360142435

matplot(x = rownames(est_rl_pw),
        y = est_rl_pw,
        xlab = "block size",
        ylab = "quantile",
        main = "Estimates of a quantile",
        ylim = range(c(est_rl_pw_range, true_rl, rl_pw)),
        cex = 1,
        cex.lab = 1,
        cex.axis = 1,
        type = "l",
        lty = c("dotted", "solid", "dotted"),
        lwd = c(2,2,2),
        col = c(3, 1, 3))

abline(h = true_rl, col = 4, lwd = 2)
abline(h = rl_mw, col = 7, lwd = 2)
abline(h = rl_pw, col = 6, lwd = 2)
abline(h = est_rl_pw_range, col = 6, lty = "dotted", lwd = 2)
abline(h = est_rl_mw_range, col = 7, lty = "dotted", lwd = 2)
```



```
# Legend:  
# blue: Quantile from the true distribution  
# yellow: Quantile from GEV mixture model with respect to distribution functions  
# pink: Quantile from GEV mixture model with respect to parameters
```