Modeling extreme values with a GEV mixture probability distributions

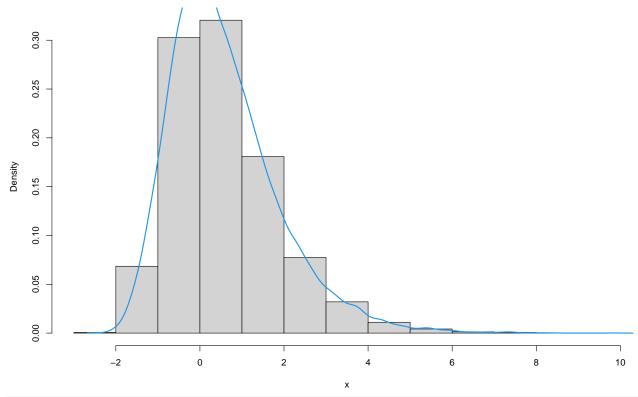
Standard Gumbel distribution

Pascal Alain Dkengne Sielenou

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```
# 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
hist(x, prob = TRUE)
lines(density(x),
     lwd = 2,
     col = 4)
```

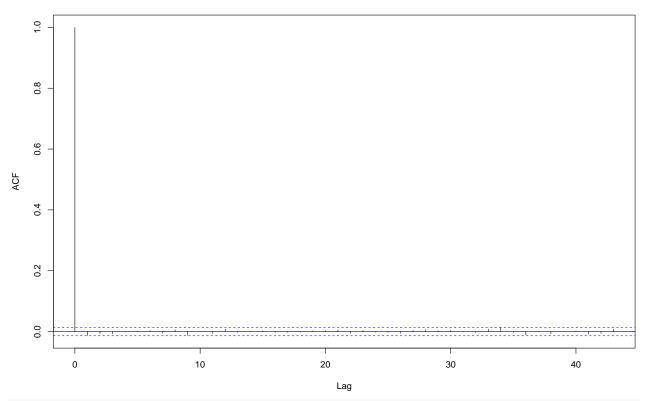




Autocorrelation function of all data

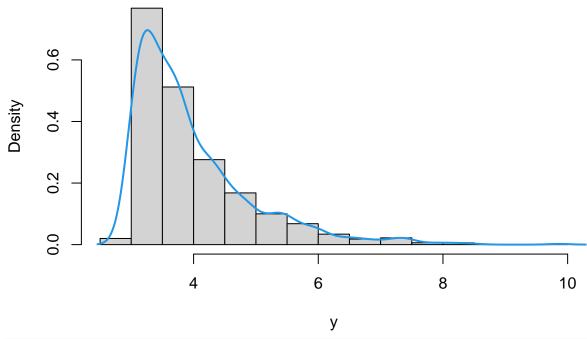
acf(x)

Series x



Histogram of the largest data

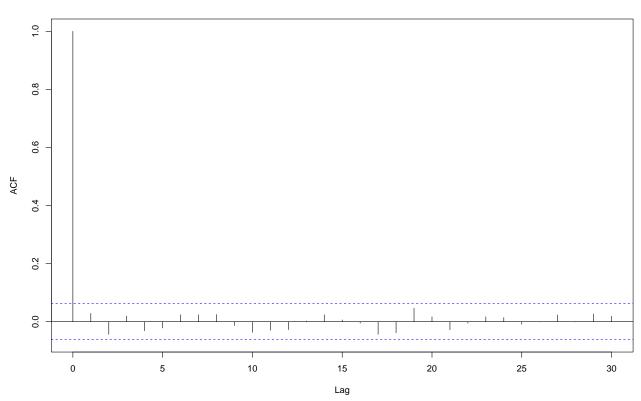
Histogram of y



Autocorrelation function of the largest data

acf(y)

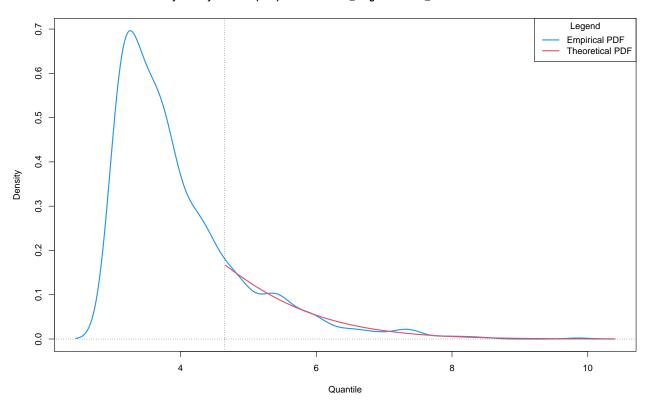
Series y



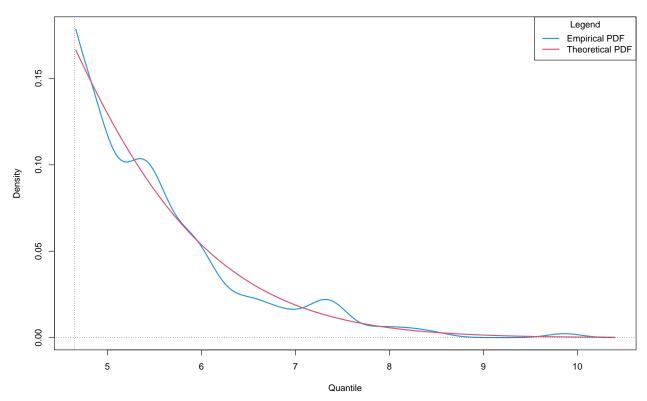
```
# Estimation of gev mixture models
gev_mixture_model <- estimate_gev_mixture_model_parameters(x = x,</pre>
                                                            block_sizes = NULL,
                                                            minimum_nblocks = 50,
                                                            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$extremal_indexes
##
             13
                                       15
                                                                               18
## 0.8920296849 0.9412167819 0.9109104376 0.9663117674 1.0000000000 1.0000000000
##
             19
## 0.9153737685 1.0000000000
gev_mixture_model$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
## 0.1340587202 0.1178963789 0.1227637377 0.1282733987 0.1365001850 0.1057314779
             19
## 0.1494401683 0.1053359333
```

```
gev_mixture_model$automatic_weights_pw_scale
## 0.1168023571 0.1315041849 0.1253806495 0.1224047623 0.1157881200 0.1427175822
            19
## 0.1016704514 0.1437318925
gev_mixture_model$automatic_weights_pw_loc
            13
                                      15
                                                                            18
##
                         14
                                                   16
                                                               17
## 0.1161439230 0.1315013058 0.1229014897 0.1244577977 0.1169475723 0.1392863559
##
            19
## 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
## 0.14900807916 0.30984800794 0.33393783784 0.16664609783 0.04055997724
                           19
# Model diagnostics
## GEV mixture model with respect to parameters
par(mfrow = c(2, 1))
plot_gev_mixture_model_pdf(gev_mixture_model,
                          type = "automatic_weights",
                          model_wise = FALSE,
                          zoom = FALSE,
                          xlab = "Quantile",
                          ylab = "Density",
                          main = "Probability Density Function (PDF) Plot")
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")
```

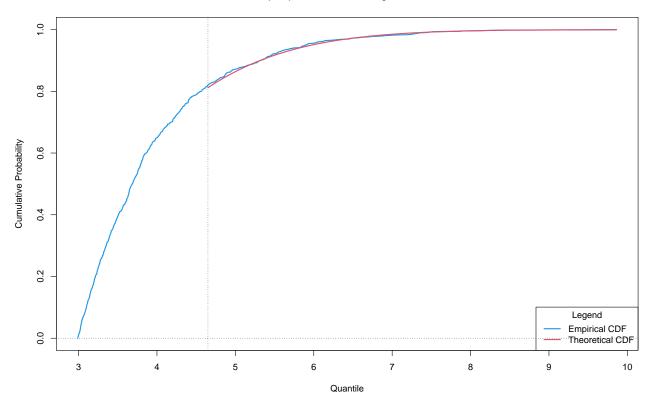
Probability Density Function (PDF) Plot : automatic_weights - model_wise = FALSE : zoom = FALSE



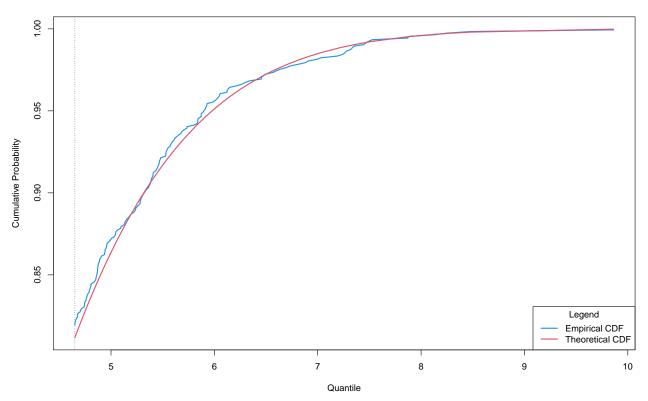
 $\label{eq:probability Density Function (PDF) Plot: automatic_weights - model_wise = FALSE: zoom = TRUE$



 $\label{lem:complex} \textbf{Cumulative Distribution Function (CDF) Plot: automatic_weights - model_wise = FALSE: zoom = FALSE: automatic_weights - model_wise = FALSE: zoom = FALSE: automatic_weights - model_wise = FALSE: zoom = FA$

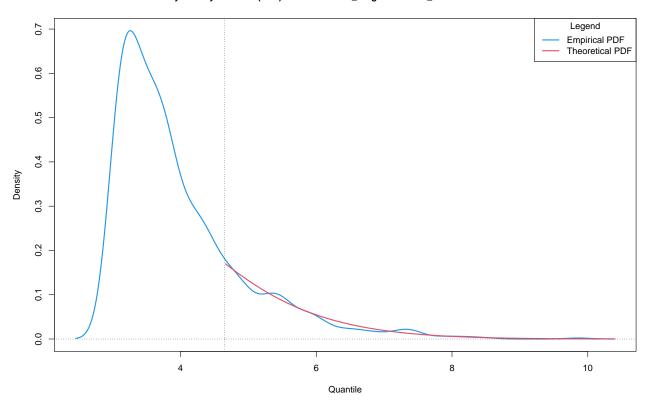


 $\label{lem:cumulative} \textbf{Cumulative Distribution Function (CDF) Plot: automatic_weights - model_wise = FALSE: zoom = TRUE \\$

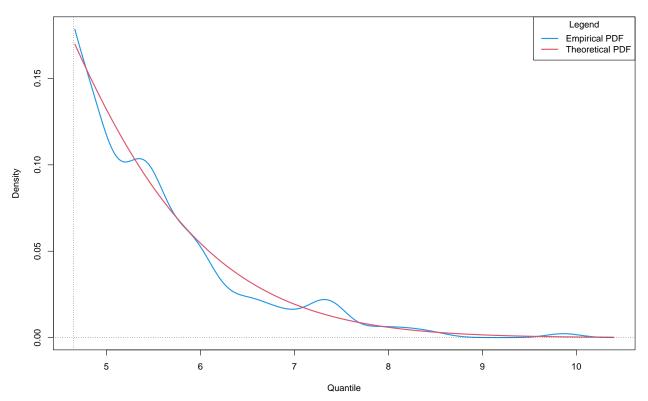


$\hbox{\it \#\# GEV mixture model with respect to distribution functions}$

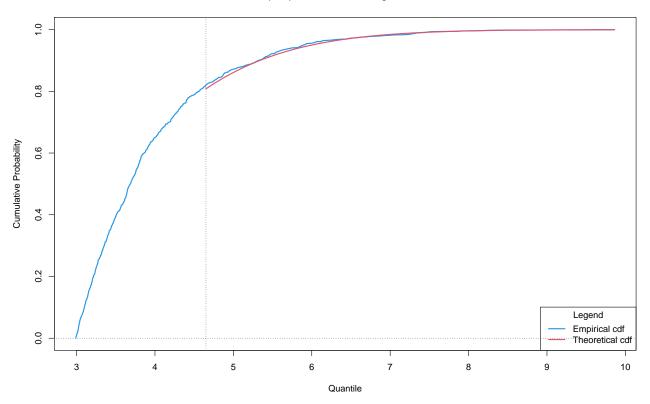
 $\label{probability Density Function (PDF) Plot: automatic_weights - model_wise = TRUE: zoom = FALSE$



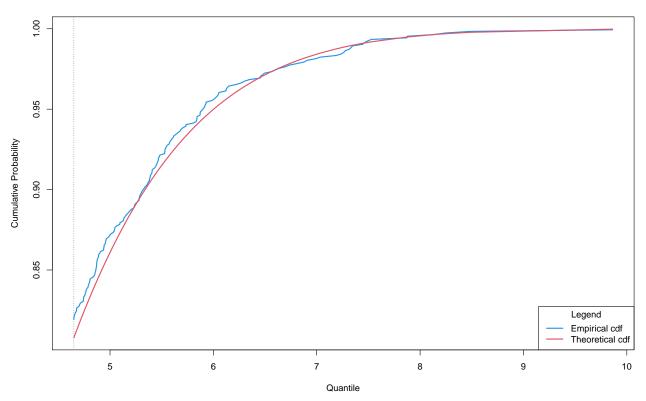
Probability Density Function (PDF) Plot : automatic_weights - model_wise = TRUE : zoom = TRUE



 $\label{lem:cumulative Distribution Function (CDF) Plot: automatic_weights - model_wise = TRUE: zoom = FALSE$



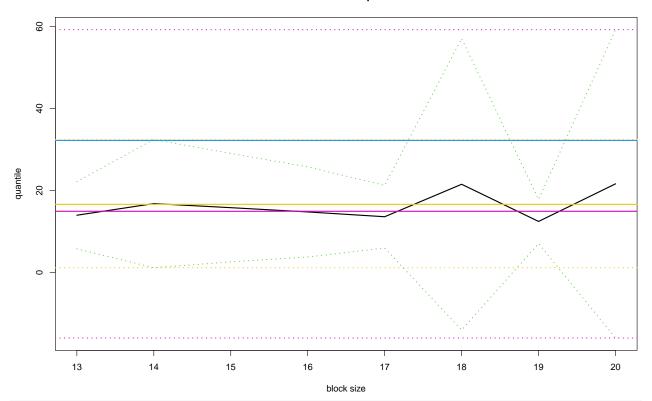
Cumulative Distribution Function (CDF) Plot : automatic_weights - model_wise = TRUE : zoom = TRUE



```
# Estimation of an extreme quantile
estimator_types <- c("automatic_weights_mw",</pre>
                      "pessimistic_weights_mw",
                      "identic_weights_mw",
                      "automatic_weights_pw",
                      "pessimistic weights pw",
                      "identic_weights_pw",
                      "empirical",
                      "confidence_interval_mw",
                      "confidence_interval_pw")
alpha <- 10^{-14}
## Quantile from the true distribution
true_rl <- calculate_gev_inverse_cdf(p = 1 - alpha,</pre>
                                      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[2]
##
        estimate
## 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[2]
##
        estimate
## 1 16.62950934
## Quantiles from equivalent estimated GEV models
est_rl_pw <- estimate_gev_mixture_model_quantile(gev_mixture_model,</pre>
                                                   alpha = alpha,
                                                   confidence_level = 0.95,
                                                   do.ci = TRUE,
```

```
estimator_type = estimator_types[9])
est_rl_pw
##
             lower
                      estimate
## 13 5.767240436 13.97538891 22.18353739
## 14 1.170296259 16.81563188 32.4609675
## 15 2.588521355 15.82128038 29.05403941
## 16 3.787271938 14.78148292 25.7756939
## 17 5.959090952 13.61031275 21.26153455
## 18 -14.02932298 21.5313576 57.09203818
## 19 7.071913005 12.45937114 17.84682927
## 20 -15.99101196 21.63956997 59.27015189
## Comparison of estimated quantiles
est_rl_pw_range <- range(as.matrix(est_rl_pw))</pre>
est_rl_mw <- estimate_gev_mixture_model_quantile(gev_mixture_model,</pre>
                                                  alpha = alpha,
                                                  confidence_level = 0.95,
                                                  do.ci = TRUE,
                                                  estimator_type = estimator_types[8])
est_rl_mw_range <- range(as.matrix(est_rl_mw))</pre>
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)),
       cex = 1,
        cex.lab = 1,
        cex.axis = 1,
        type = "1",
       lty = c("dotted", "solid", "dotted"),
       1wd = c(2,2,2),
        col = c(3, 1, 3))
abline(h = true_rl, col = 4, lwd = 2)
abline(h = rl_mw[2], col = 7, lwd = 2)
abline(h = rl_pw[2], 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)
```

Estimates of a quantile



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