

# Modeling extreme values with a GEV mixture probability distributions

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```
# library(xfun)

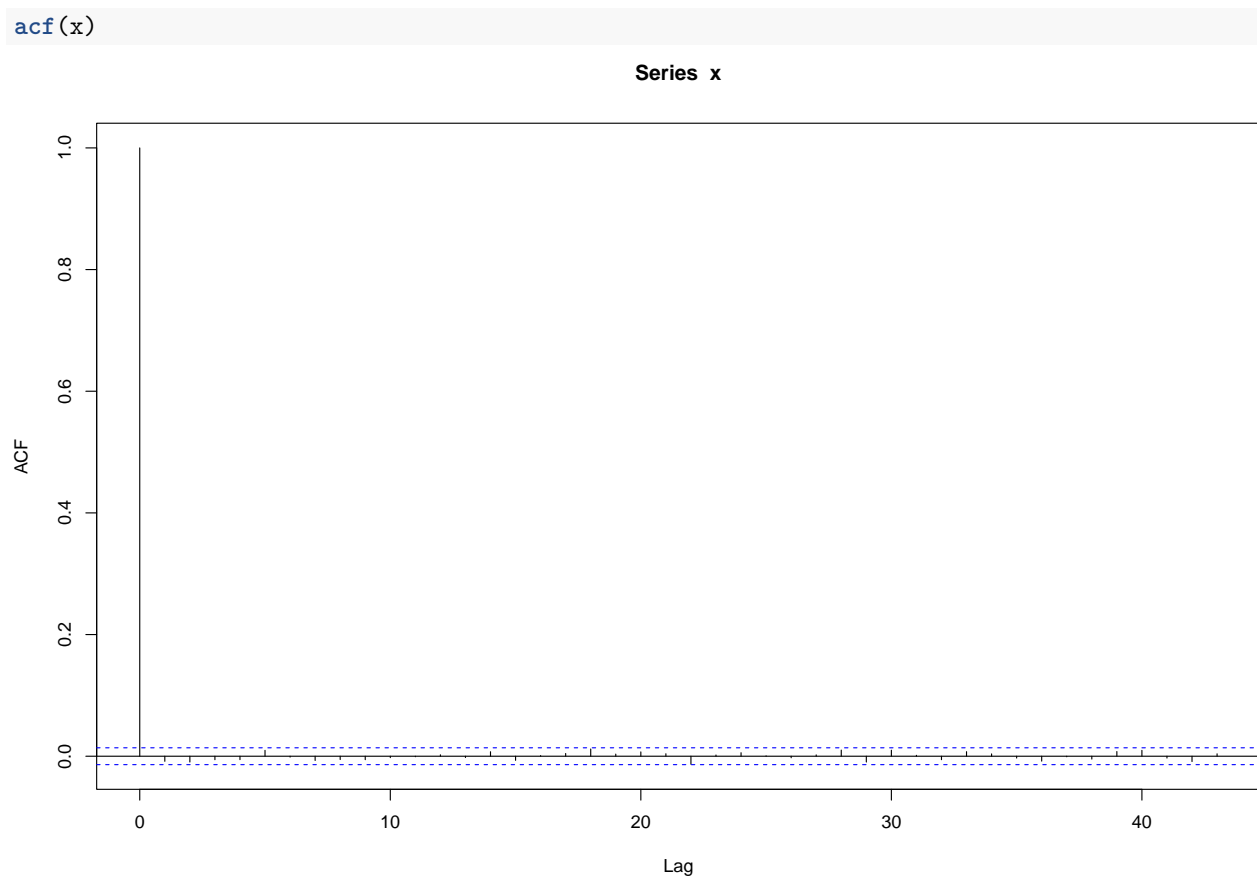
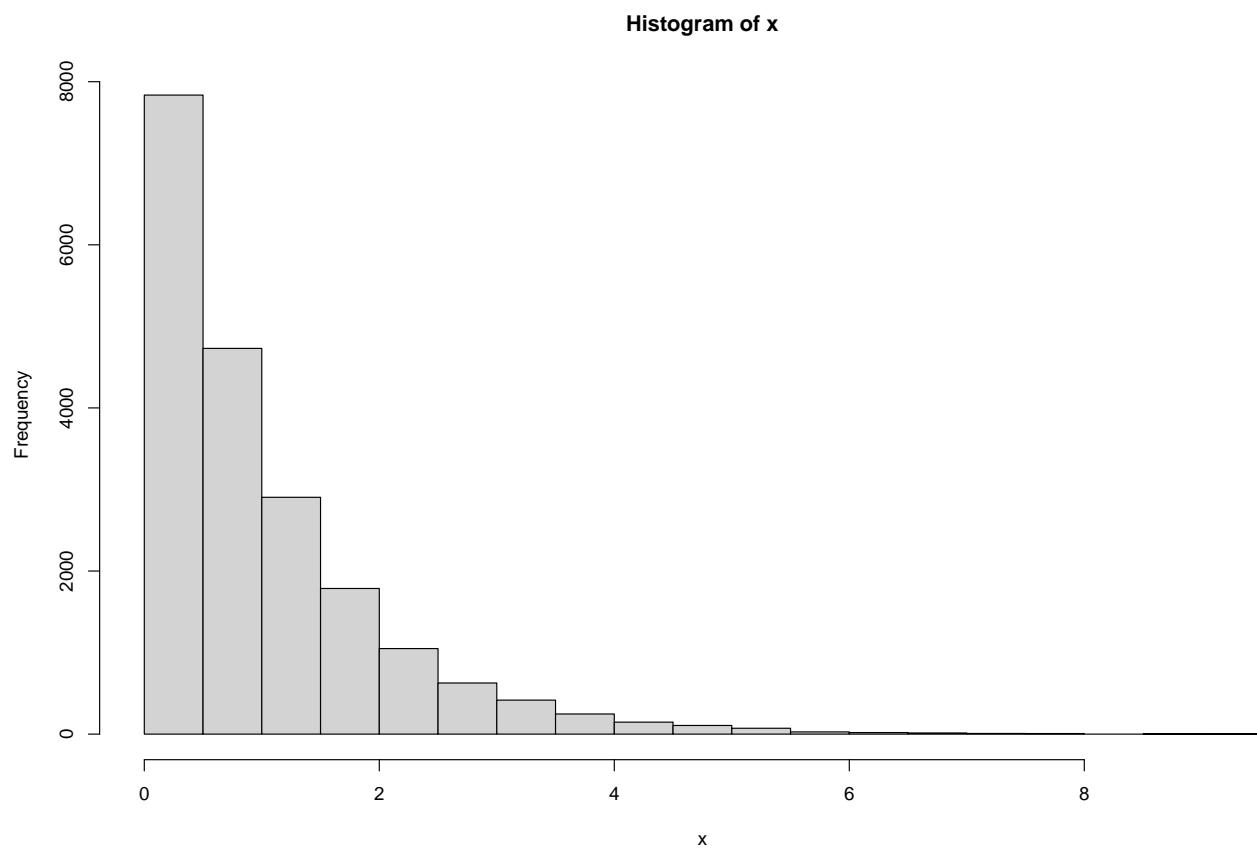
path <- ".."

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/estimate_gev_mixture_model_quantile.R"))

n <- 20000

set.seed(1122)
x <- rexp(n = n)

hist(x)
```



```

nlargest <- 1000

#
y <- extract_nlargest_sample(x, n = nlargest)

gev_mixture_model <- estimate_gev_mixture_model_parameters(x,
                                                             nsloc = NULL,
                                                             std.err = FALSE,
                                                             block_sizes = NULL,
                                                             minimum_nblocks = 50,
                                                             threshold = NULL,
                                                             nlargest = nlargest,
                                                             confidence_level = 0.95,
                                                             log_mv = TRUE,
                                                             log_pw = TRUE,
                                                             trace = FALSE)

## Successful convergence.
## Successful convergence.

names(gev_mixture_model)

## [1] "data"
## [2] "data_largest"
## [3] "block_sizes"
## [4] "equivalent_block_sizes"
## [5] "rejected_block_sizes"
## [6] "block_maxima_indexes_object"
## [7] "gev_models_object"
## [8] "extremal_indexes"
## [9] "normalized_gev_parameters_object"
## [10] "weighted_normalized_gev_parameters_object"
## [11] "identic_weights_mw"
## [12] "pessimistic_weights_mw"
## [13] "pessimistic_weights_pw_shape"
## [14] "pessimistic_weights_pw_scale"
## [15] "pessimistic_weights_pw_loc"
## [16] "automatic_weights_mw"
## [17] "automatic_weights_mw_statistics"
## [18] "automatic_weights_pw_shape"
## [19] "automatic_weights_pw_scale"
## [20] "automatic_weights_pw_loc"
## [21] "automatic_weights_pw_statistics"

gev_mixture_model$block_sizes

## [1] 11 12 13 14 15 16 17 18 19 20

gev_mixture_model$normalized_gev_parameters_object

##           loc_star      scale_star      shape_star
## 11 3.29574918040805 0.976445970509164 -0.03667709095381891
## 12 3.52277758042938 0.804214867250803 0.02327082826820088
## 13 3.34734924573158 0.975487676545606 -0.03914751087346882
## 14 3.37760466022731 0.934922331849203 -0.02311039263952277
## 15 3.12226861242757 1.169635359507689 -0.09470560157345080

```

```
## 16 2.93110818059460 1.177531657286269 -0.08031243590085080
## 17 2.65709377754162 1.369077387579498 -0.11932062611815418
## 18 3.75244477098718 0.776090074273978 0.00850247899967913
## 19 3.87855287007820 0.689409149289278 0.04546195765427703
## 20 3.57806473766632 0.970271024691920 -0.06303282380142441
```

```
gev_mixture_model$weighted_normalized_gev_parameters_object
```

```
##               loc_star      scale_star      shape_star
## identic_weights    3.34630136160918 0.984308549878341 -0.0379071216938534
## pessimistic_weights 3.45993502872538 1.024148504273390 -0.0353681535590280
## automatic_weights   3.60700208661585 0.738869635008201 0.0454619576542770
```

```
gev_mixture_model$automatic_weights_mw_statistics
```

```
## $function_value
## [1] 0.00255261684711061
##
## $gradient_value
## [1] 9.90838708142538e-06
##
## $function_reduction
## [1] 0.0157651151485194
##
## $number_iterations
## [1] 1789
##
## $convergence
## [1] 0
##
## $message
## [1] "Successful convergence"
```

```
gev_mixture_model$automatic_weights_pw_statistics
```

```
## $function_value
## [1] 0.00170124879625866
##
## $gradient_value
## [1] 6.95493901186608e-06
##
## $function_reduction
## [1] 0.0190034725999947
##
## $number_iterations
## [1] 3617
##
## $convergence
## [1] 0
##
## $message
## [1] "Successful convergence"
```

```
gev_mixture_model$automatic_weights_mw
```

```
##               11               12               13               14
## 0.0000000000000000 0.844460821843412 0.0000000000000000 0.0000000000000000
```

```

##              15              16              17              18
## 0.000000000000000 0.155539178156588 0.000000000000000 0.000000000000000
##              19              20
## 0.000000000000000 0.000000000000000

gev_mixture_model$pessimistic_weights_pw_shape

##              11              12              13              14
## 0.0999961068269341 0.1061739909730453 0.0997493793384735 0.1013619680046930
##              15              16              17              18
## 0.0943586306897419 0.0957265709739753 0.0920643434883038 0.1046174980960356
##              19              20
## 0.1085564486893274 0.0973950629194701

gev_mixture_model$pessimistic_weights_pw_scale

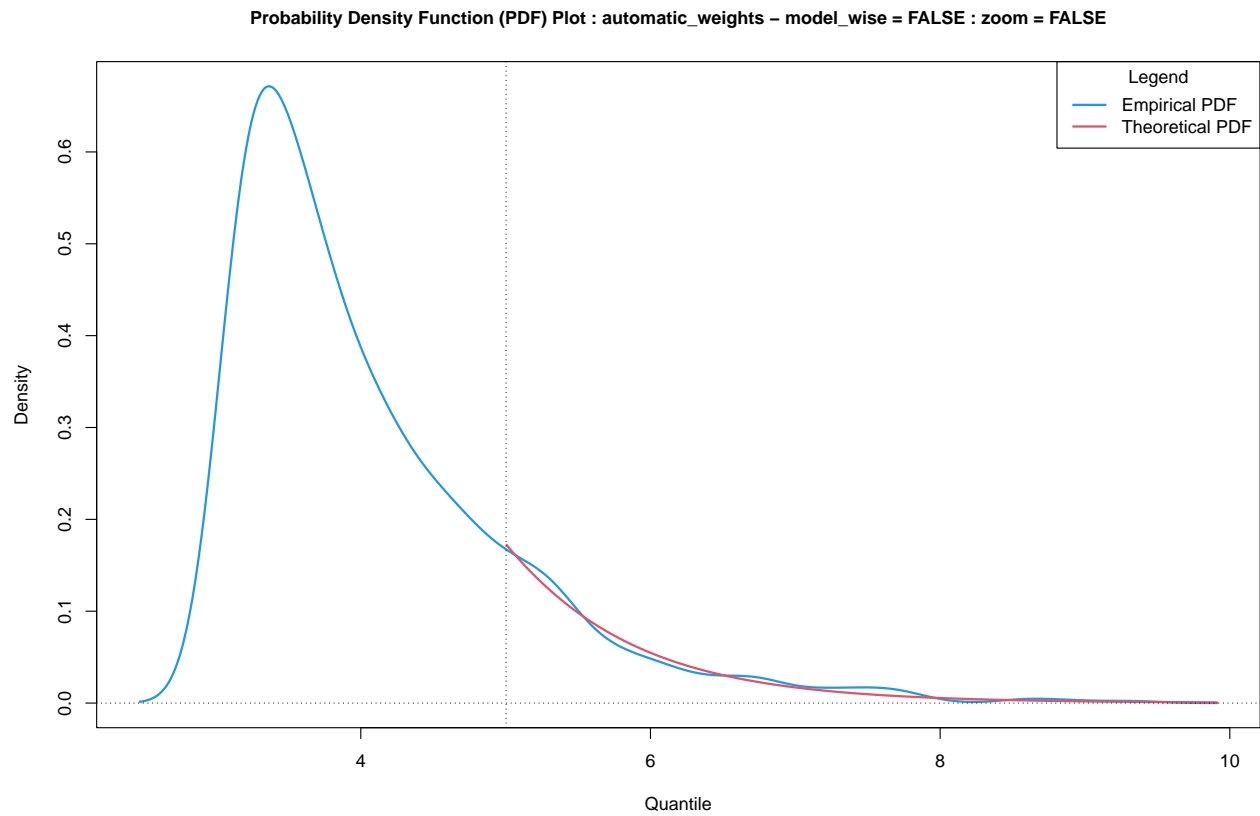
##              11              12              13              14
## 0.0972797133258867 0.0818885656908950 0.0971865354168743 0.0933230222641345
##              15              16              17              18
## 0.1180112383653980 0.1189467790375891 0.1440588488104659 0.0796175523205252
##              19              20
## 0.0730068775772010 0.0966808671910303

gev_mixture_model$pessimistic_weights_pw_loc

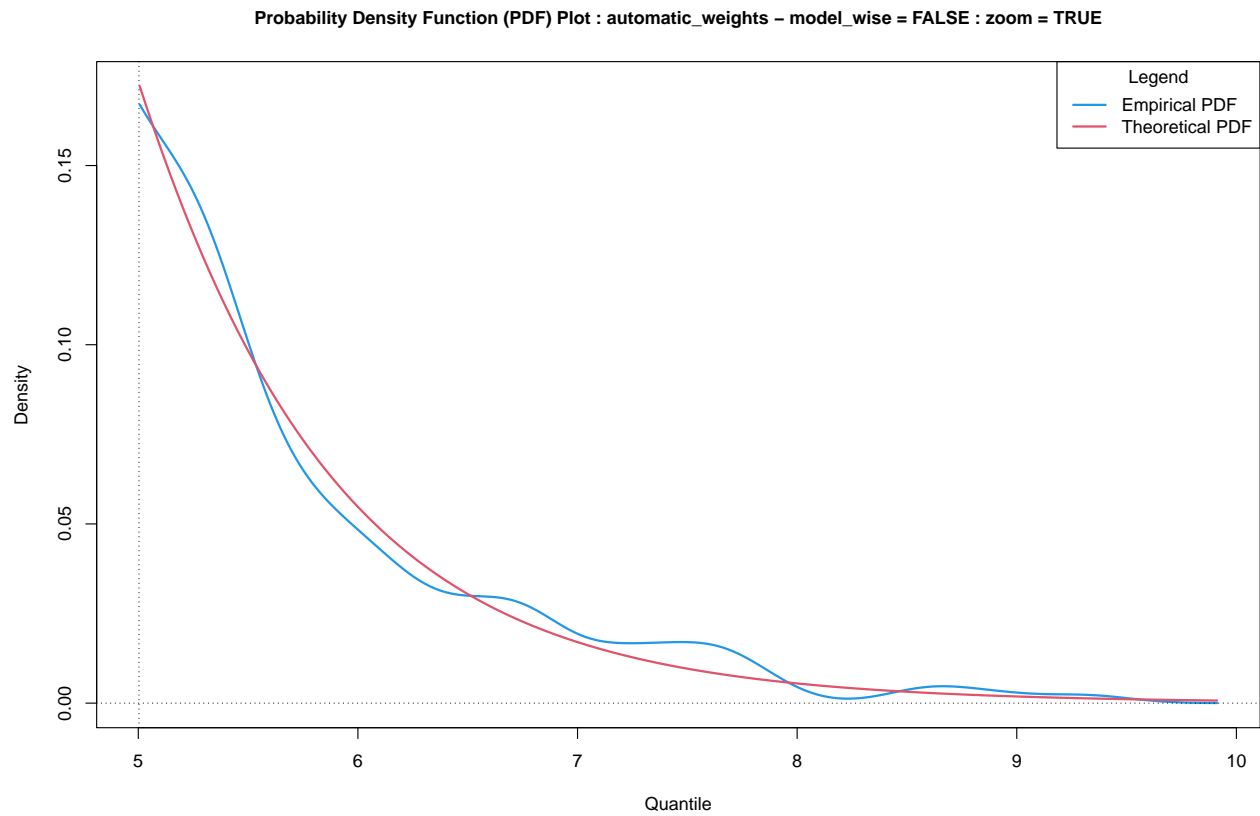
##              11              12              13              14
## 0.0896668094859343 0.1125197862248132 0.0944150746069881 0.0973152942651538
##              15              16              17              18
## 0.0753858895106338 0.0622687493205185 0.0473442454881823 0.1415702802926139
##              19              20
## 0.1605980040547069 0.1189158667504553

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")

```

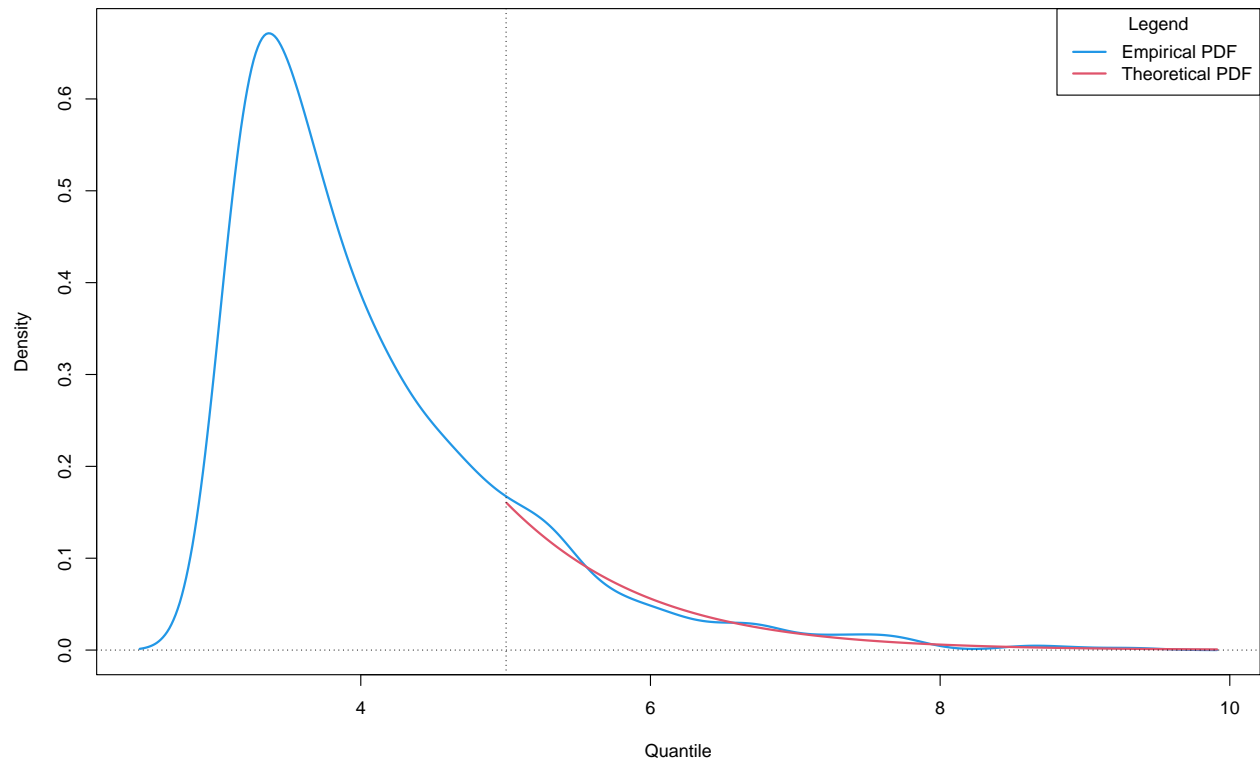


```
plot_gev_mixture_model_pdf(gev_mixture_model,  
    type = "automatic_weights",  
    model_wise = FALSE,  
    zoom = TRUE,  
    xlab = "Quantile",  
    ylab = "Density",  
    main = "Probability Density Function (PDF) Plot")
```



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

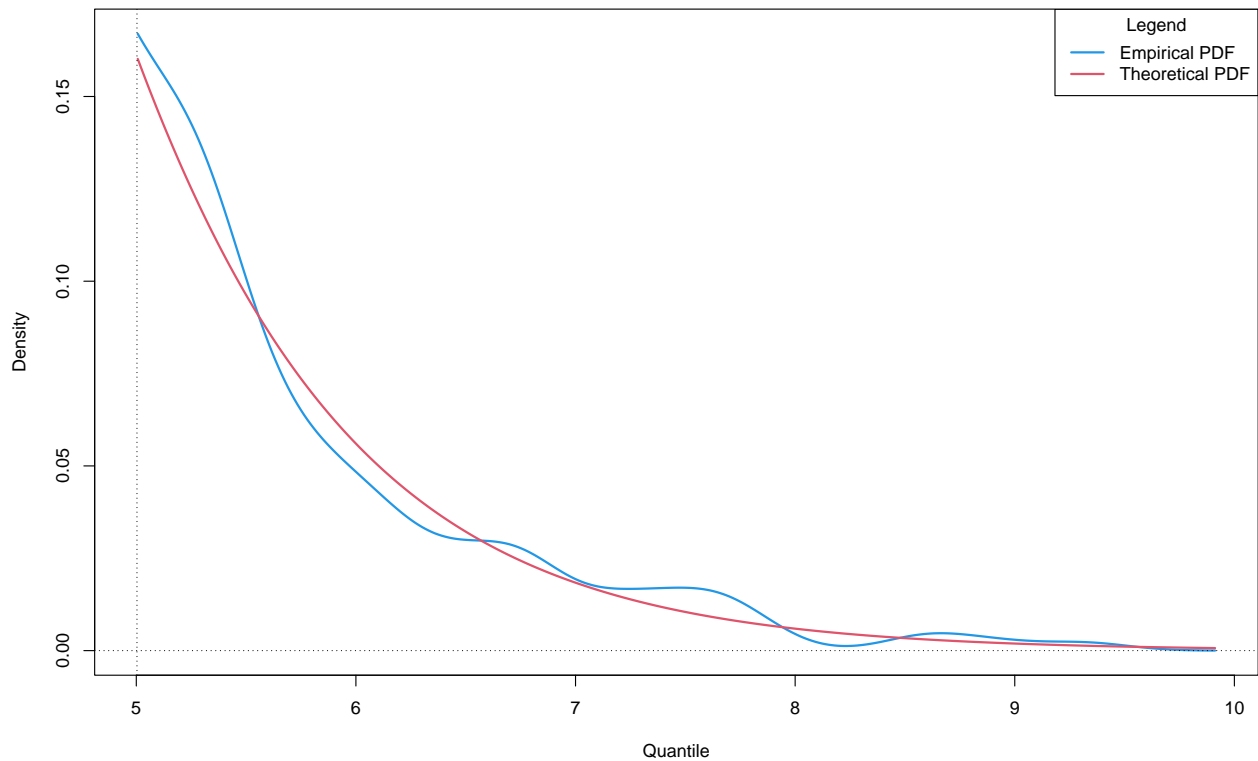
Probability Density Function (PDF) Plot : automatic\_weights – model\_wise = TRUE : zoom = FALSE



```
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



```
estimator_types <- c("automatic_weights_mw",
  "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)
```

```
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
```

```
## lower estimate upper
## 1 NA 36.9414820328362 NA
```

```
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

##      lower      estimate upper
## 1      NA 48.7652905178263    NA

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

rl_empirical

##      lower      estimate upper
## 1      NA 9.32637366225866    NA

true_rl <- qexp(p = 1 - alpha)
true_rl

## [1] 32.2369908993468

est_rl_pw <- estimate_gev_mixture_model_quantile(gev_mixture_model,
                                                  alpha = alpha,
                                                  confidence_level = 0.95,
                                                  do.ci = TRUE,
                                                  estimator_type = estimator_types[9])

est_rl_pw

##           lower      estimate      upper
## 11 -5.33725533795543 20.8088856145853 46.9550265671261
## 12 -54.6851376503337 37.2083453260977 129.101828302529
## 13 -7.20051572831085 20.333368467944 47.8672526641989
## 14 -20.7925982801311 23.249751888565 67.2921020572611
## 15 3.25947733267602 14.6973568099349 26.1352362871937
## 16 -1.04892346751356 16.1923500364021 33.4336235403178
## 17 3.47565308886628 13.7806549395963 24.0856567903263
## 18 -55.5777942109169 29.5201084977206 114.618011206358
## 19 -122.562697705713 46.0134907413773 214.589679188468
## 20 -5.0597127061962 16.5338799886942 38.1274726835847

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

## [1] -122.562697705713 214.589679188468

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

est_rl_mw

##           lower      estimate      upper
## 12 -54.6851376503337 37.2083453260977 129.101828302529
## 16 -1.04892346751356 16.1923500364021 33.4336235403178

```

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

```
## [1] -54.6851376503337 129.1018283025291
```

```
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 = "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[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)
```

