

Modeling extreme values with a GEV mixture probability distributions

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September 28th, 2023

```
# 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/plot_several_standardized_block_maxima_mean.R"))
xfun::in_dir(dir = path, expr = source("./src/estimate_gev_mixture_model_quantile.R"))

n <- 100000

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

nlargest <- 1000

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] 9 10 11 12 13 14 15 16 17 18 19 20
```

```
gev_mixture_model$normalized_gev_parameters_object
```

```
##           loc_star      scale_star      shape_star
## 9  2.45628443725258 0.256440743198709 0.04256378669663031
## 10 2.47637313395835 0.258234275166653 0.02738164560120679
## 11 2.41543340306138 0.290686060651678 0.00117924257969587
## 12 2.36210335865845 0.335347611311548 -0.04123424946548924
## 13 2.39423783521682 0.321979960277751 -0.03308091180478364
## 14 2.23713203017343 0.403833891594456 -0.08456762308699213
## 15 2.52832405559420 0.237112037607216 0.04876518655428982
## 16 2.38493413827506 0.308693952100226 -0.01572077091047429
## 17 2.55394533700493 0.245046724739142 0.02928797698600330
## 18 2.50689633726953 0.239545317872480 0.04979310041600821
## 19 2.43795358324203 0.290534269252053 -0.00360412083143491
## 20 2.55804489122312 0.210755847668614 0.08137171032744203
```

```
gev_mixture_model$weighted_normalized_gev_parameters_object
```

```
##           loc_star      scale_star      shape_star
## identic_weights  2.44263854507749 0.283184224286711 0.00851124775517518
## pessimistic_weights 2.45020786975786 0.285873506997451 0.01051224767980413
## automatic_weights  2.33315317672659 0.363324234169863 -0.08456762308699213
```

```
gev_mixture_model$automatic_weights_mw_statistics
```

```
## $function_value
## [1] 0.00249655002489225
##
## $gradient_value
## [1] 9.98927512674275e-06
##
## $function_reduction
## [1] 0.0192397897654247
##
## $number_iterations
## [1] 2405
##
## $convergence
## [1] 0
```

```
##
## $message
## [1] "Successful convergence"
```

```
gev_mixture_model$automatic_weights_pw_statistics
```

```
## $function_value
## [1] 0.00225569403067106
##
## $gradient_value
## [1] 5.4161485770754e-06
##
## $function_reduction
## [1] 0.0172758264439327
##
## $number_iterations
## [1] 4496
##
## $convergence
## [1] 0
##
## $message
## [1] "Successful convergence"
```

```
gev_mixture_model$automatic_weights_mw
```

```
##          9          10          11          12
## 0.000000000000000 0.000000000000000 0.000000000000000 0.217645067134693
##          13          14          15          16
## 0.229859424904641 0.116978205638081 0.000000000000000 0.435517302322585
##          17          18          19          20
## 0.000000000000000 0.000000000000000 0.000000000000000 0.000000000000000
```

```
gev_mixture_model$pessimistic_weights_pw_shape
```

```
##          9          10          11          12
## 0.0861334227071130 0.0848356096494699 0.0826415827308796 0.0792097569225287
##          13          14          15          16
## 0.0798582207968855 0.0758506375433885 0.0866692301634263 0.0812566743152026
##          17          18          19          20
## 0.0849974886832106 0.0867583644698420 0.0822472219452161 0.0895417900728373
```

```
gev_mixture_model$pessimistic_weights_pw_scale
```

```
##          9          10          11          12
## 0.0810259699353189 0.0811714230010026 0.0838487882886427 0.0876784888205607
##          13          14          15          16
## 0.0865142324026847 0.0938936602694034 0.0794748813743538 0.0853724055599297
##          17          18          19          20
## 0.0801079981630140 0.0796685015052749 0.0838360617296270 0.0774075889501875
```

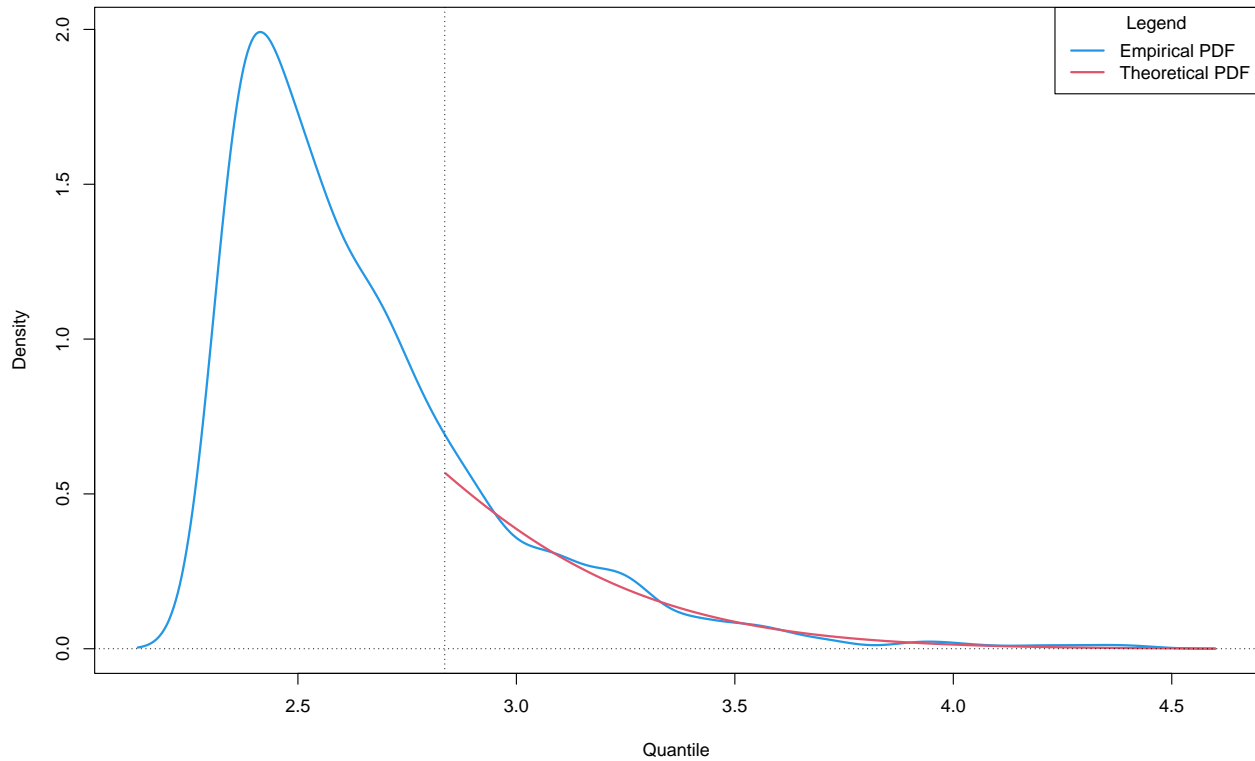
```
gev_mixture_model$pessimistic_weights_pw_loc
```

```
##          9          10          11          12
## 0.0841558754528664 0.0858635523846430 0.0807872943642429 0.0765917723797842
##          13          14          15          16
## 0.0790929811812481 0.0675939398815008 0.0904421443865863 0.0783605365587992
##          17          18          19          20
```

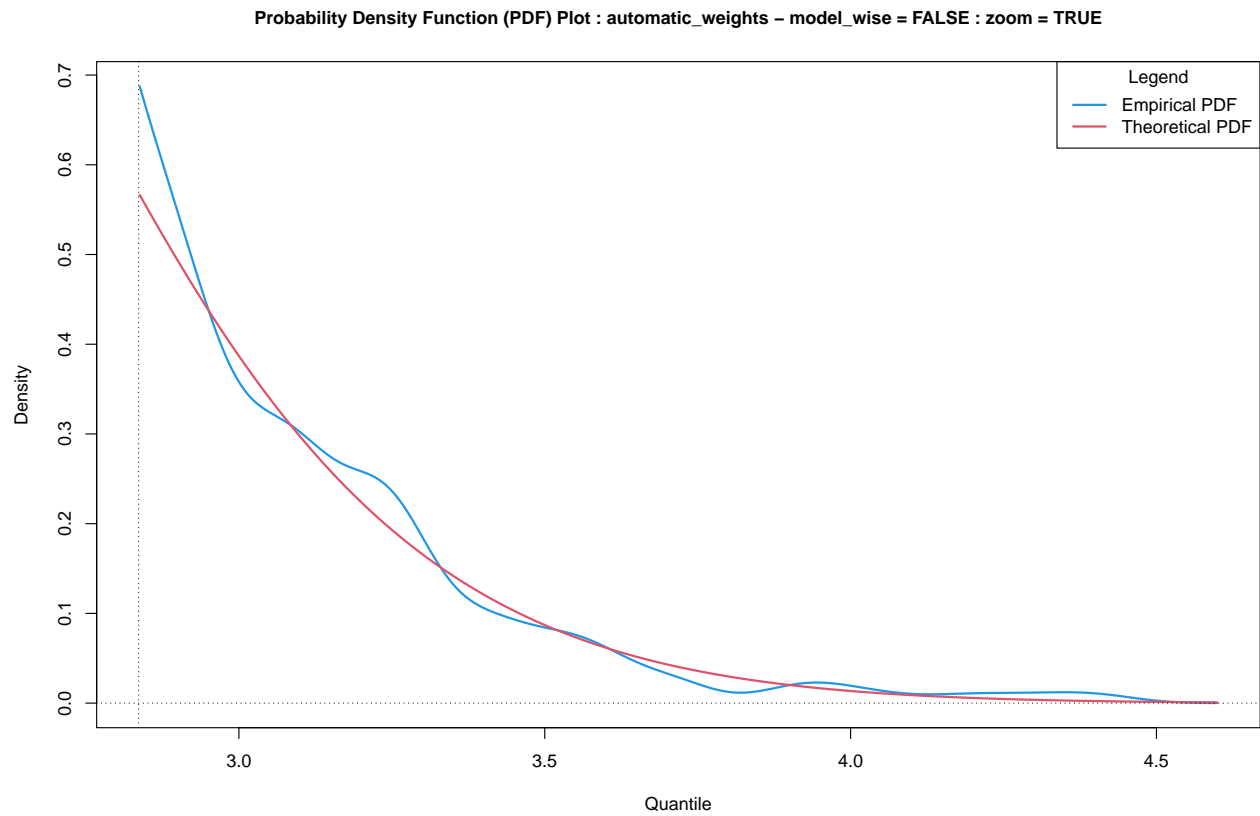
```
## 0.0927893285529685 0.0885247912056398 0.0826272794241551 0.0931705042275658
```

```
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")
```

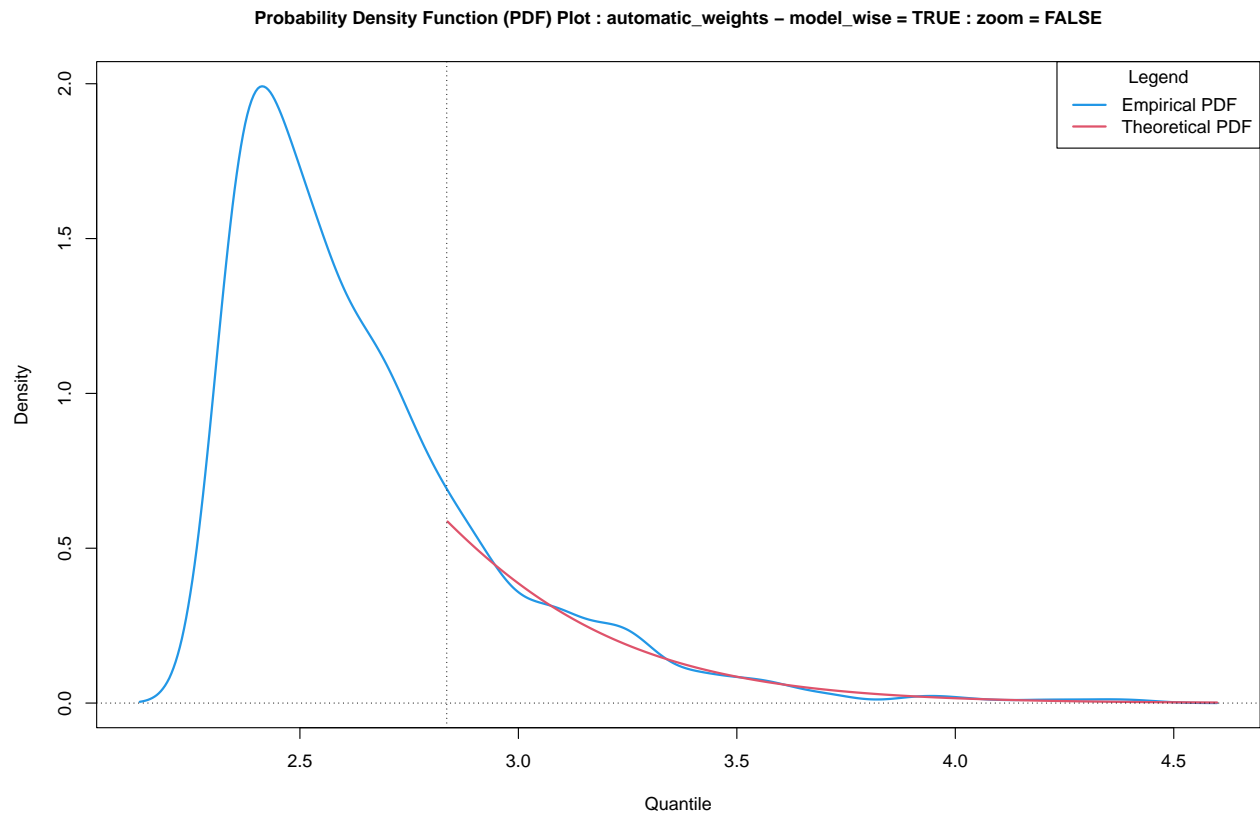
Probability Density Function (PDF) Plot : automatic_weights – model_wise = FALSE : zoom = FALSE



```
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")
```

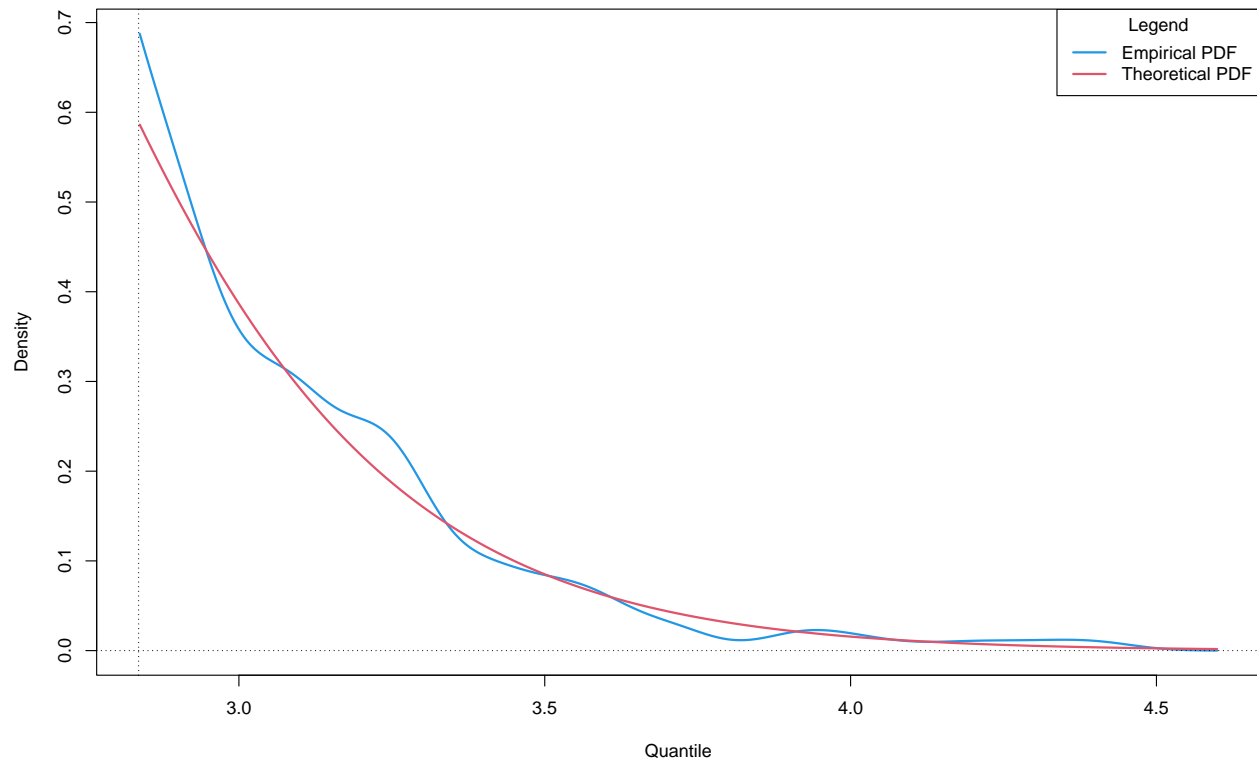


```
plot_gev_mixture_model_pdf(gev_mixture_model,  
    type = "automatic_weights",  
    model_wise = TRUE,  
    zoom = FALSE,  
    xlab = "Quantile",  
    ylab = "Density",  
    main = "Probability Density Function (PDF) Plot")
```



```
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)

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

results_mw

## lower estimate upper
## 1 NA 9.1361288596493 NA

results_pw <- estimate_gev_mixture_model_quantile(gev_mixture_model,
  alpha = alpha,
  confidence_level = 0.95,
  do.ci = TRUE,
  estimator_type = estimator_types[4])
```

```

results_pw

##      lower      estimate upper
## 1      NA 6.2141903736901      NA
quantile(x = x, probs = 1 - alpha)

##      100%
## 4.41549649952614
true_rl <- qnorm(p = 1 - alpha)
true_rl

## [1] 7.65073090515564
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
## 9  -14.5448370657666 15.9618808542649 46.4685987742964
## 10 -7.87222655492214 13.1427696185319 34.1577657919859
## 11 -4.33187694955725 10.5798268362574 25.4915306220721
## 12  0.713285466003487 7.89190215772055 15.0705188494376
## 13 -0.306488605311648 8.2254566208075 16.7574018469266
## 14  2.37798722449619 6.55091820312705 10.7238491817579
## 15 -22.6533534491439 16.374257173082 55.401867795308
## 16 -3.40365404588741 9.30340745378741 22.0104689534622
## 17 -13.7916118860738 12.9811746142776 39.7539611146291
## 18 -27.4961046483109 16.7401148584929 60.9763343652966
## 19 -7.21233500698153 10.0791119203642 27.3705588477098
## 20 -52.8359050279607 24.505063861496 101.846032750953

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

## [1] -52.8359050279607 101.8460327509527

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  0.713285466003487 7.89190215772055 15.0705188494376
## 13 -0.306488605311648 8.2254566208075 16.7574018469266
## 14  2.37798722449619 6.55091820312705 10.7238491817579
## 16 -3.40365404588741 9.30340745378741 22.0104689534622

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

```



```
## [1] -3.40365404588741 22.01046895346223
```

```
matplot(x = rownames(est_rl_pw),  
        y = est_rl_pw,  
        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 = results_mw[2], col = 7, lwd = 2)  
abline(h = results_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)
```

