An Introduction to Statistical Modelling of Extreme Values

Package

```
library(ismev)

## Loading required package: mgcv

## Loading required package: nlme

## This is mgcv 1.8-23. For overview type 'help("mgcv-package")'.

library(extRemes)

## Loading required package: Lmoments

## Loading required package: distillery

## Loading required package: car

##

## Attaching package: 'extRemes'

## The following objects are masked from 'package:stats':

##

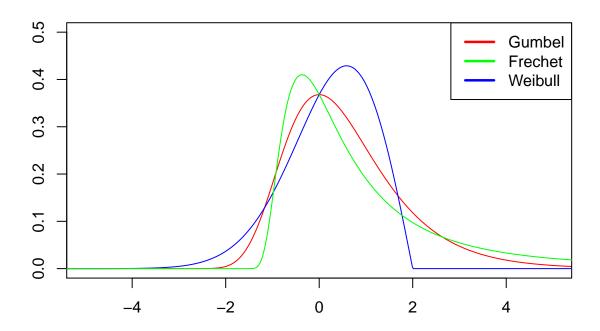
## qqnorm, qqplot

source("extreme_functions.r")
```

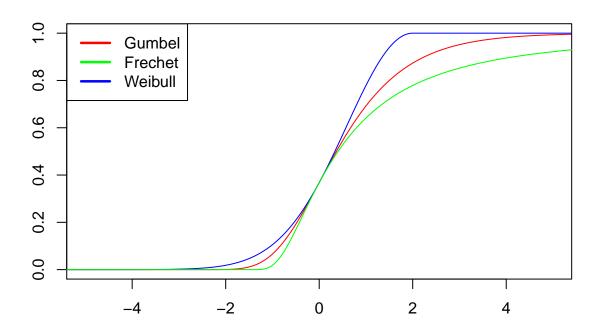
Chapter 3: Classical Extreme Value Theory and Models

The Generalsed extreme value distribution

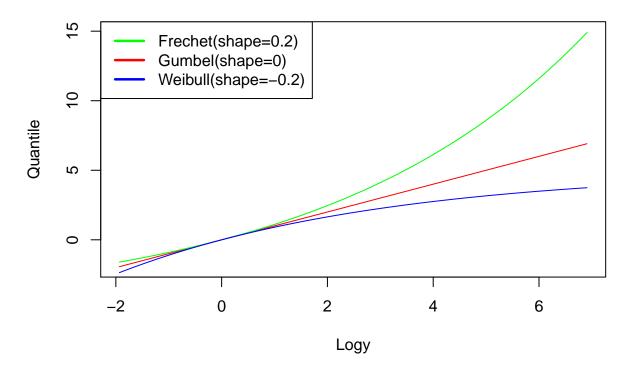
PDF



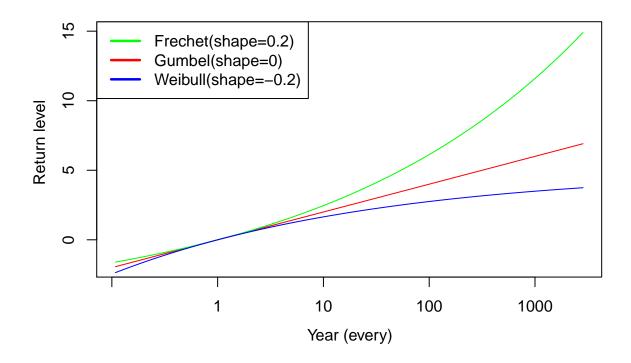
\mathbf{CDF}



Return level

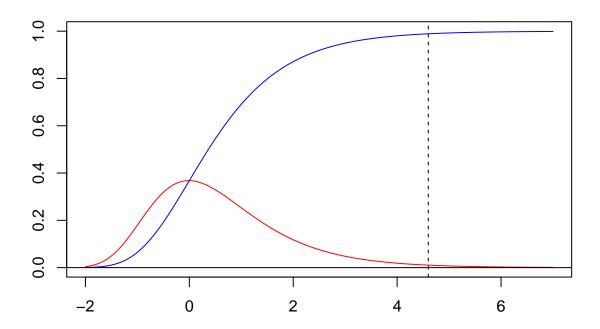


```
p \leftarrow seq(0.001, 0.999, length.out = 10000)
r.p <- -1/log(1-p)
rl.gum \leftarrow log(r.p)
rl.fre <- (r.p^(0.2)-1)/(0.2)
rl.wei \leftarrow (r.p^{-(-0.2)-1)/(-0.2)}
plot(x=log(r.p), y=rl.gum, t="l", col="red",
     ylim=c(-2,15), xaxt='n', ann=FALSE,
     ylab="Return level", xlab="Logy")
lines(x=log(r.p), y=rl.fre, col="green")
lines(x=log(r.p), y=rl.wei, col="blue")
legend("topleft", legend =
         c("Frechet(shape=0.2)", "Gumbel(shape=0)", "Weibull(shape=-0.2)"),
       lwd=c(2.5,2.5,2.5),col=c("green","red","blue"))
axis(side = 1, at = c(-2,0,2,4,6), labels= c("",1,10,100,1000))
mtext("Year (every)", side = 1, line=2.5, at=2.5)
mtext("Return level", side = 2, line=2.5, at=7)
```



```
library(extRemes)
xx <- seq(-2, 7, length=1000)
pdf <- devd(xx, loc = 0, scale = 1, shape = 0.01, type = c("GEV"))
cdf <- pevd(xx, loc = 0, scale = 1, shape = 0.01, type = c("GEV"))
plot(x=xx, y=pdf, t="l", col="red", ylab="", xlab="", ylim = c(0,1))
lines(x=xx, cdf, col="blue")
abline(h=0)

return.level <- -log(-log(1-0.01))
abline(v=return.level, lty=2)</pre>
```



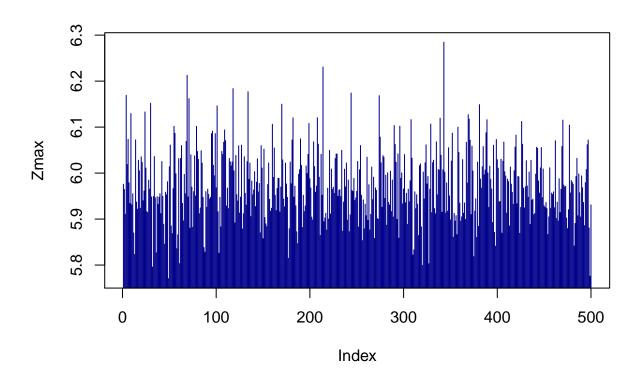
Inference for return levels

Case for xi < 0

Random variates from beta distribution

```
set.seed(100)
n <- 500
Z <- matrix(rbeta(100*n, shape1=1, shape2=2), nrow=100, ncol=n, byrow=T) + 5 +
    rnorm(n, mean = 0, sd = 0.1)

Zmax <- apply(Z, 2, max)
plot(Zmax, t="h", col = "darkblue")</pre>
```



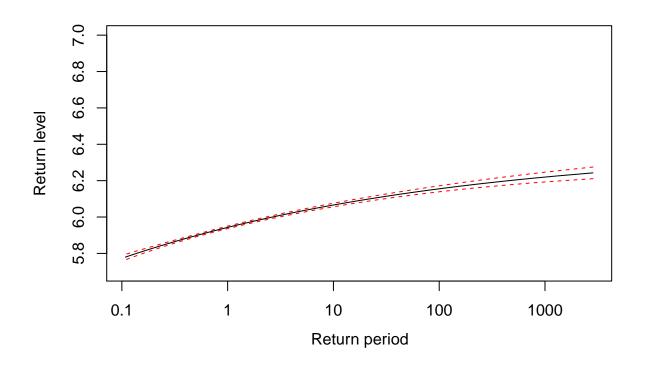
```
fit <- gev.fit(Zmax)</pre>
## $conv
## [1] 0
##
## $nllh
## [1] -576.9642
##
## $mle
       5.94301637 0.07199454 -0.16131124
## [1]
##
## $se
## [1] 0.003497106 0.002394182 0.021856572
mu <- fit$mle[1]; sigma <- fit$mle[2]; xi <- fit$mle[3]; cov <- fit$cov</pre>
# Return level data
p \leftarrow seq(0, 1, by = 0.001)
y.p \leftarrow -log(1-p)
rl <- mu-sigma*(1-y.p^(-xi))/(xi)
rl.data <- data.frame(prob=p, y.p=y.p, return.level=rl)</pre>
# Variance fdata
var <- matrix(nrow = length(y.p), ncol = 1)</pre>
for(i in 1:length(y.p)){
  dz1 <- 1
  dz2 \leftarrow -xi^{-1}*(1-y.p[i]^{-xi})
```

```
dz3 <- sigma*xi^(-2)*(1-y.p[i]^(-xi)) - sigma*xi^(-1)*y.p[i]^(-xi)*log(y.p[i])
dz <- matrix(c(dz1,dz2,dz3), nrow=3, byrow=T)
  each.var <- t(dz) %*% cov %*% dz
  var[i] <- each.var
}
var.data <- data.frame(prob=p, y.p=y.p, var=var)

plot(x=-log(y.p), y=rl.data$return.level, t="1",
      ylim = c(5.7, 7),
      xaxt='n', ylab="Return level", xlab="")
axis(side = 1, at = c(-2,0,2,4,6), labels= c(0.1,1,10,100,1000))
mtext("Return period", side = 1, line=2.5, at=2.5)

lower.line <- rl.data$return.level-1.96*sqrt(var.data$var)
upper.line <- rl.data$return.level+1.96*sqrt(var.data$var)

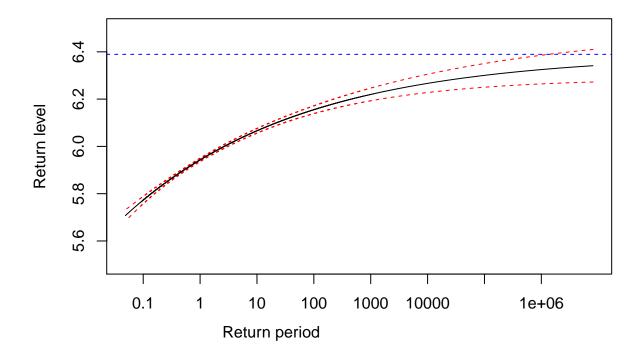
lines(x=-log(y.p), y=lower.line, col="red", lty=2)
lines(x=-log(y.p), y=upper.line, col="red", lty=2)</pre>
```



```
# upper <- mu-sigma/xi
# abline(h=upper, col="blue", lty=2)

Precise level
mu <- fit$mle[1]; sigma <- fit$mle[2]; xi <- fit$mle[3]; cov <- fit$cov
# Return level data</pre>
```

```
p \leftarrow seq(0, 1, by = 0.000001)
y.p \leftarrow -log(1-p)
rl <- mu-sigma*(1-y.p^(-xi))/(xi)
rl.data <- data.frame(prob=p, y.p=y.p, return.level=rl)</pre>
# Variance fdata
var <- matrix(nrow = length(y.p), ncol = 1)</pre>
for(i in 1:length(y.p)){
  dz1 <- 1
  dz2 <- -xi^(-1)*(1-y.p[i]^(-xi))
 dz3 <- sigma*xi^(-2)*(1-y.p[i]^(-xi)) - sigma*xi^(-1)*y.p[i]^(-xi)*log(y.p[i])
 dz <- matrix(c(dz1,dz2,dz3), nrow=3, byrow=T)</pre>
  each.var <- t(dz) %*% cov %*% dz
  var[i] <- each.var</pre>
var.data <- data.frame(prob=p, y.p=y.p, var=var)</pre>
upper <- mu - sigma/xi
plot(x=-log(y.p), y=rl.data$return.level, t="1",
     ylim = c(5.5, 6.5),
     xaxt='n', ylab="Return level", xlab="")
axis(side = 1, at = c(-2,0,2,4,6,8,10,12),
     labels= c(0.1,1,10,100,1000,10000,100000,1000000))
mtext("Return period", side = 1, line=2.5, at=2.5)
lower.line <- rl.data$return.level-1.96*sqrt(var.data$var)</pre>
upper.line <- rl.data$return.level+1.96*sqrt(var.data$var)</pre>
lines(x=-log(y.p), y=lower.line, col="red", lty=2)
lines(x=-log(y.p), y=upper.line, col="red", lty=2)
abline(h=upper, col="blue", lty=2)
```



Annual maximum sea-levels at port price (using ismev)

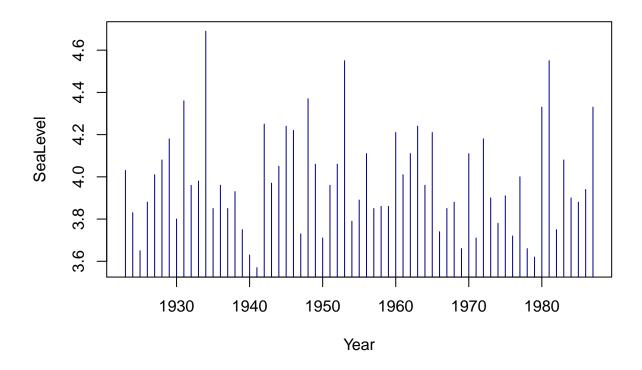
```
library(ismev)
library(extRemes)
source("extreme_functions.r")
```

Data

The portpirie data frame has 65 rows and 2 columns. The second column gives annual maximimum sea levels recorded at Port Pirie, South Australia, from 1923 to 1987. The first column gives the corresponding years.

```
data("portpirie")
str(portpirie)

## 'data.frame': 65 obs. of 2 variables:
## $ Year : num 1923 1924 1925 1926 1927 ...
## $ SeaLevel: num 4.03 3.83 3.65 3.88 4.01 4.08 4.18 3.8 4.36 3.96 ...
plot(portpirie, t="h", col="darkblue")
```

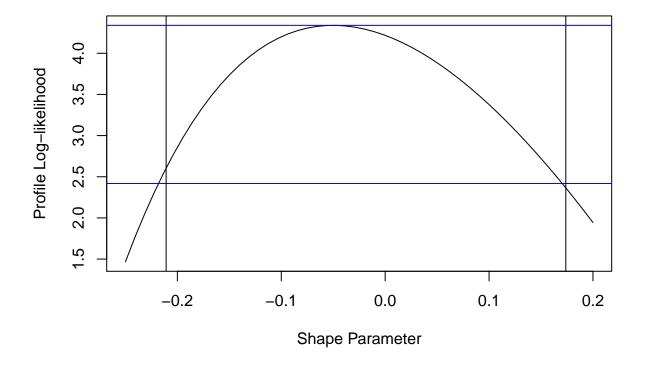


GEV fit

```
options(digits = 3)
fitgev <- gev.fit(portpirie$SeaLevel)</pre>
## $conv
## [1] 0
##
## $nllh
## [1] -4.34
##
## $mle
## [1]
       3.8747 0.1980 -0.0501
##
## $se
## [1] 0.0279 0.0202 0.0983
#log-likelihood
-fitgev$nllh
## [1] 4.34
# Covariance
fitgev$cov
             [,1]
                        [,2]
                                  [,3]
##
```

```
## [1,] 0.000780 0.000197 -0.001074
## [2,] 0.000197 0.000410 -0.000777
## [3,] -0.001074 -0.000777 0.009654
# Standard errors
fitgev$se
## [1] 0.0279 0.0202 0.0983
# confidential interval
ci <- cbind(fitgev$mle-1.96*fitgev$se, fitgev$mle+1.96*fitgev$se)</pre>
colnames(ci) <- c("lower", "upper")</pre>
rownames(ci) <- c("mu", "sigma", "xi")</pre>
ci
##
          lower upper
## mu
          3.820 3.929
## sigma 0.158 0.238
         -0.243 0.142
## xi
# Profile likelihood for shape
cgev.profxi(fitgev, xlow = -0.25, xup = 0.2, nint = 50)
```

If routine fails, try changing plotting interval

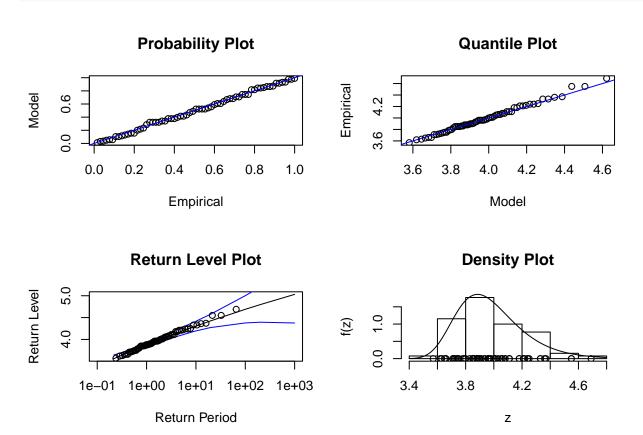


```
## $optimal.value
## [1] 4.34
##
## $optimal.x
```

```
## [1] -0.048
##
## $confint.lower
## [1] -0.211
##
## $confint.upper
## [1] 0.174
```

Diagnosis

gev.diag(fitgev)



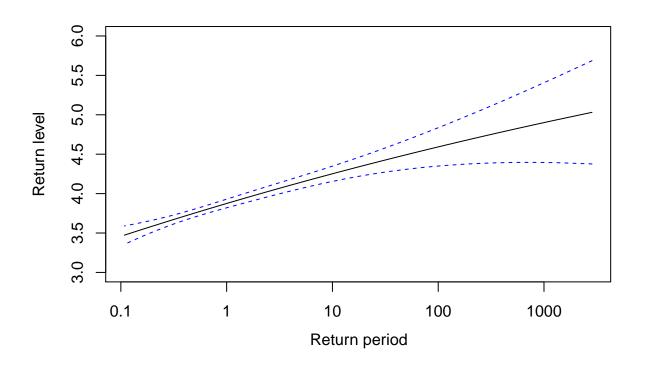
Return level

```
mu <- fitgev$mle[1]; sigma <- fitgev$mle[2]; xi <- fitgev$mle[3]; cov <- fitgev$cov

# Return level data
p <- seq(0, 1, by = 0.001)
y.p <- -log(1-p)
rl <- mu-sigma*(1-y.p^(-xi))/(xi)
rl.data <- data.frame(prob=p, y.p=y.p, return.level=rl)

# Variance fdata
var <- matrix(nrow = length(y.p), ncol = 1)</pre>
```

```
for(i in 1:length(y.p)){
  dz1 <- 1
  dz2 \leftarrow -xi^{-1}*(1-y.p[i]^{-xi})
  dz3 <- sigma*xi^(-2)*(1-y.p[i]^(-xi)) - sigma*xi^(-1)*y.p[i]^(-xi)*log(y.p[i])
  dz <- matrix(c(dz1,dz2,dz3), nrow=3, byrow=T)</pre>
  each.var <- t(dz) %*% cov %*% dz
  var[i] <- each.var</pre>
var.data <- data.frame(prob=p, y.p=y.p, var=var)</pre>
plot(x=-log(y.p), y=rl.data$return.level, t="l",
     ylim = c(3, 6),
     xaxt='n', ylab="Return level", xlab="")
axis(side = 1, at = c(-2,0,2,4,6), labels= c(0.1,1,10,100,1000))
mtext("Return period", side = 1, line=2.5, at=2.5)
lower.line <- rl.data$return.level-1.96*sqrt(var.data$var)</pre>
upper.line <- rl.data$return.level+1.96*sqrt(var.data$var)</pre>
lines(x=-log(y.p), y=lower.line, col="blue", lty=2)
lines(x=-log(y.p), y=upper.line, col="blue", lty=2)
```



10-year return

```
#return level
rl <- rl.data$return.level[rl.data$prob==0.1]

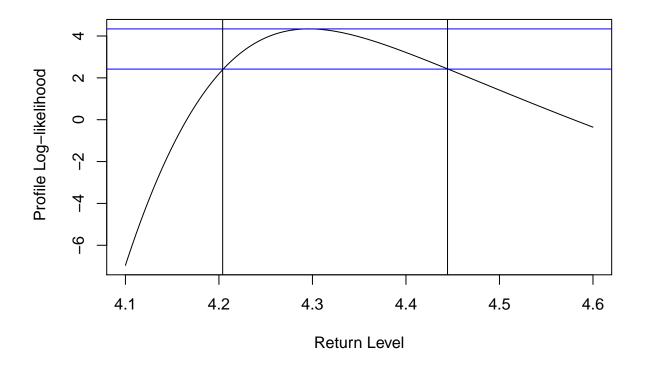
#standard deviation
sd <- sqrt(var.data$var[var.data$prob==0.1])

#CI at 95%
rl-1.96*sd

## [1] 4.19
rl+1.96*sd

## [1] 4.4
Priofile likelihood
cgev.prof(fitgev, m = 10, xlow = 4.1, xup = 4.6)</pre>
```

If routine fails, try changing plotting interval



```
## $optimal.value
## [1] 4.34
##
## $optimal.x
## [1] 4.3
##
## $confint.lower
## [1] 4.2
```

```
##
## $confint.upper
## [1] 4.44
```

100-year return

```
#return level
rl <- rl.data$return.level[rl.data$prob==0.01]

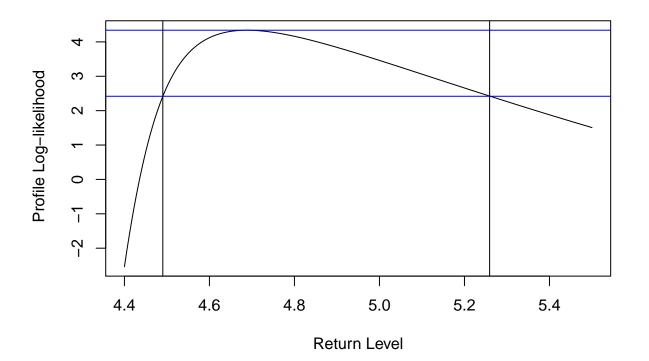
#standard deviation
sd <- sqrt(var.data$var[var.data$prob==0.01])

#CI at 95%
rl-1.96*sd

## [1] 4.38
rl+1.96*sd

## [1] 5
Priofile likelihood
cgev.prof(fitgev, m = 100, xlow = 4.4, xup = 5.5)</pre>
```

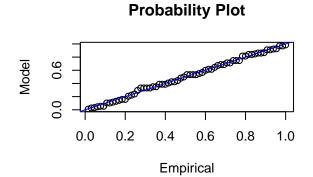
If routine fails, try changing plotting interval

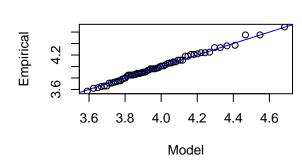


\$optimal.value ## [1] 4.34

```
##
## $optimal.x
## [1] 4.69
##
## $confint.lower
## [1] 4.49
## $confint.upper
## [1] 5.26
Gumbel fit
fitgum <- gum.fit(portpirie$SeaLevel)</pre>
## $conv
## [1] 0
##
## $nllh
## [1] -4.22
##
## $mle
## [1] 3.869 0.195
##
## $se
## [1] 0.0255 0.0189
# Covariance
fitgum$cov
##
             [,1]
                      [,2]
## [1,] 0.000650 0.000153
## [2,] 0.000153 0.000355
# Standard errors
fitgum$se
## [1] 0.0255 0.0189
# confidential interval
ci <- cbind(fitgum$mle-1.96*fitgum$se, fitgum$mle+1.96*fitgum$se)</pre>
colnames(ci) <- c("lower", "upper")</pre>
rownames(ci) <- c("mu", "sigma")</pre>
сi
##
         lower upper
## mu
         3.819 3.919
## sigma 0.158 0.232
Diagnosis
```

gum.diag(fitgum)



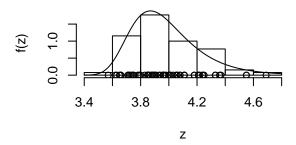




Return Period

Density Plot

Quantile Plot



Model comparison(Log ratio test)

```
#statistic
2*(-fitgev$nllh - -fitgum$nllh)

## [1] 0.243

#p-value
as.numeric(pchisq(-fitgev$nllh - -fitgum$nllh, df=1, lower.tail=FALSE))

## [1] 0.728
```

Annual maximum sea-levels at port price (using extRemes)

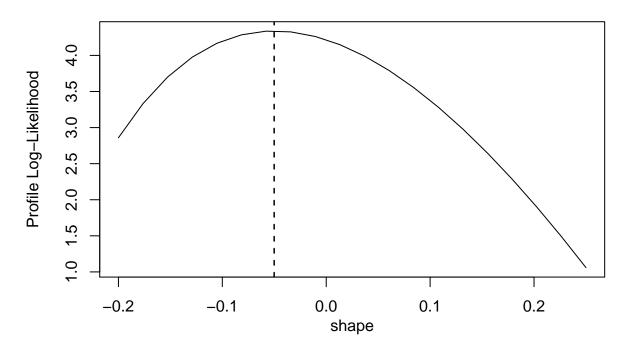
GEV fit

```
options(digits = 3)
fitgev <- fevd(portpirie$SeaLevel, type = "GEV")

# estimated paramters
par.gev <- distill(fitgev)[1:3]
par.gev

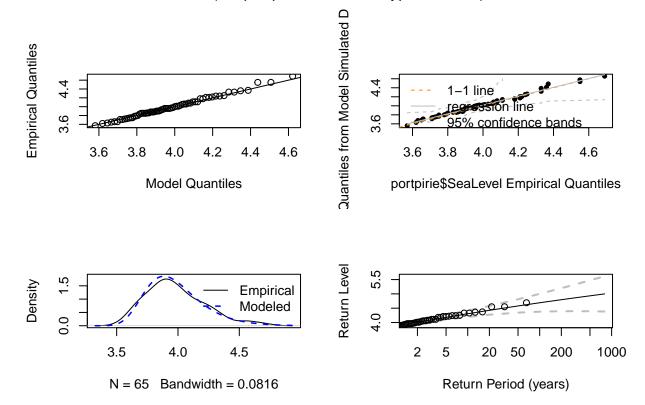
## location scale shape
## 3.8747 0.1980 -0.0501</pre>
```

```
# log-likelihood
loglik.gev <- -distill(fitgev)[4]</pre>
loglik.gev
## nllh
## 4.34
# covariance
cov.gev <- matrix(as.numeric(distill(fitgev)[5:13]),nrow = 3, byrow = T,</pre>
              dimnames =list(c("location", "scale", "shape"),
                             c("location", "scale", "shape")))
cov.gev
##
             location
                          scale
                                     shape
## location 0.000780 0.000197 -0.001074
## scale
             0.000197 0.000410 -0.000777
## shape
            -0.001074 -0.000777 0.009654
# Standard errors
se.gev <- sqrt(diag(cov.gev))</pre>
se.gev
## location
               scale
                        shape
   0.0279 0.0202
                       0.0983
# confidence interval
ci.fevd(fitgev, type = "parameter")
## fevd(x = portpirie$SeaLevel, type = "GEV")
## [1] "Normal Approx."
##
            95% lower CI Estimate 95% upper CI
##
## location
                  3.820 3.8747
                                          3.929
## scale
                   0.158
                          0.1980
                                          0.238
                                          0.142
## shape
                  -0.243 -0.0501
# Profile likelihood for shape
profliker(fitgev, type = "parameter", which.par = 3, xrange = c(-0.2, 0.25))
```



Diagnosis

plot(fitgev)



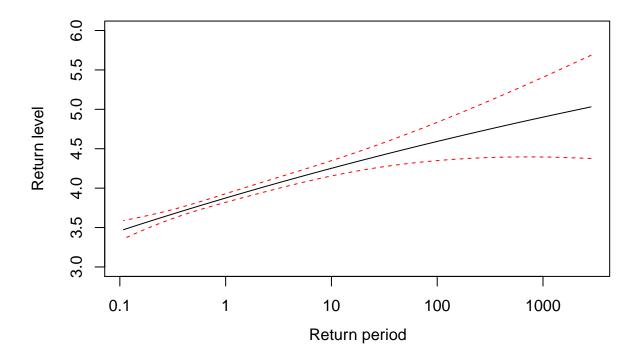
Return level

```
mu <- par.gev[1]; sigma <- par.gev[2]; xi <- par.gev[3]; cov <- cov.gev</pre>
# Return level data
p \leftarrow seq(0, 1, by = 0.001)
y.p <- -log(1-p)
rl \leftarrow mu-sigma*(1-y.p^(-xi))/(xi)
rl.data <- data.frame(prob=p, y.p=y.p, return.level=rl)</pre>
# Variance fdata
var <- matrix(nrow = length(y.p), ncol = 1)</pre>
for(i in 1:length(y.p)){
  dz1 \leftarrow 1
  dz2 <- -xi^(-1)*(1-y.p[i]^(-xi))
  dz3 <- sigma*xi^(-2)*(1-y.p[i]^(-xi)) - sigma*xi^(-1)*y.p[i]^(-xi)*log(y.p[i])
  dz <- matrix(c(dz1,dz2,dz3), nrow=3, byrow=T)</pre>
  each.var <- t(dz) %*% cov %*% dz
  var[i] <- each.var</pre>
}
var.data <- data.frame(prob=p, y.p=y.p, var=var)</pre>
plot(x=-log(y.p), y=rl.data$return.level, t="1",
     ylim = c(3, 6),
     xaxt='n', ylab="Return level", xlab="")
```

```
axis(side = 1, at = c(-2,0,2,4,6), labels= c(0.1,1,10,100,1000))
mtext("Return period", side = 1, line=2.5, at=2.5)

lower.line <- rl.data$return.level-1.96*sqrt(var.data$var)
upper.line <- rl.data$return.level+1.96*sqrt(var.data$var)

lines(x=-log(y.p), y=lower.line, col="red", lty=2)
lines(x=-log(y.p), y=upper.line, col="red", lty=2)</pre>
```



10-year return

```
#return level
rl <- rl.data$return.level[rl.data$prob==0.1]

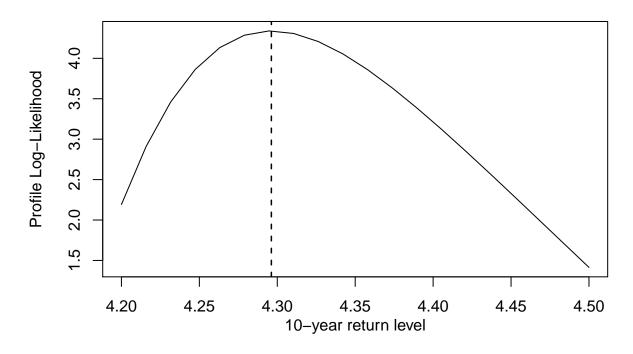
#standard deviation
sd <- sqrt(var.data$var[var.data$prob==0.1])

#CI at 95%
rl-1.96*sd

## [1] 4.19
rl+1.96*sd

## [1] 4.4</pre>
```

Priofile likelihood



100-year return

```
#return level
rl <- rl.data$return.level[rl.data$prob==0.01]

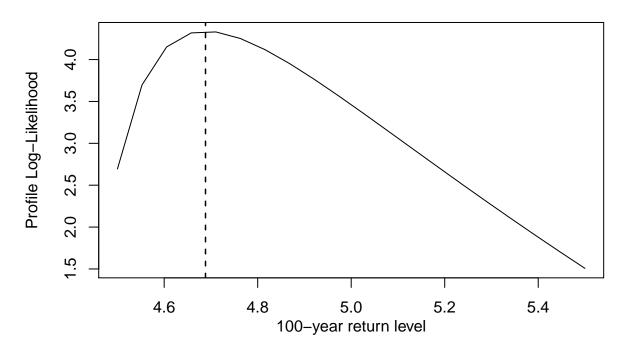
#standard deviation
sd <- sqrt(var.data$var[var.data$prob==0.01])

#CI at 95%
rl-1.96*sd

## [1] 4.38
rl+1.96*sd

## [1] 5</pre>
```

Priofile likelihood



Gumbel fit

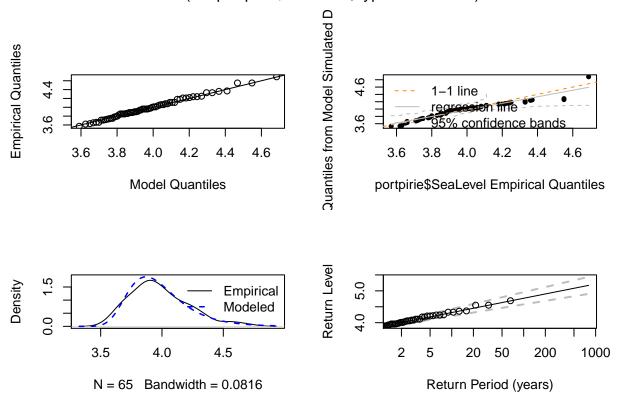
```
fitgum <- fevd(portpirie$SeaLevel, type = "Gumbel")</pre>
# estimated paramters
par.gum <- distill(fitgum)[1:2]</pre>
par.gum
## location
                scale
      3.869
                0.195
# log-likelihood
loglik.gum <- -distill(fitgum)["nllh"]</pre>
loglik.gum
## nllh
## 4.22
cov.gum <- matrix(as.numeric(distill(fitgum)[4:7]),nrow = 2, byrow = T,</pre>
               dimnames =list(c("location", "scale"), c("location", "scale")))
cov.gum
##
             location
                          scale
```

```
## location 0.000650 0.000153
## scale
            0.000153 0.000355
# Standard errors
se.gum <- sqrt(diag(cov.gum))</pre>
se.gum
## location
               scale
     0.0255
##
              0.0189
# confidential interval
ci.fevd(fitgum, type = "parameter")
## fevd(x = portpirie$SeaLevel, type = "Gumbel")
## [1] "Normal Approx."
##
##
            95% lower CI Estimate 95% upper CI
                    3.819
                             3.869
                                           3.919
## location
                    0.158
                                           0.232
## scale
                             0.195
```

Diagnosis

plot(fitgum)

fevd(x = portpirie\$SeaLevel, type = "Gumbel")



Model comparison(Log ratio test)

```
#statistic
2*(loglik.gev - loglik.gum)

## nllh
## 0.243

#p-value
as.numeric(pchisq(loglik.gev - loglik.gum, df=1, lower.tail=FALSE))

## [1] 0.728
```