Risk Management — Assignment 5

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Autumn 2021

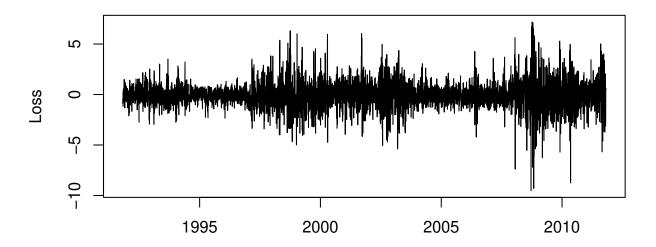
Question 1

```
## (a) values
values <- (prices$Netherlands+prices$Greece)/2

## (b) losses
losses <- 100*(1-Quot(values)) ## see package DescTools

## (c) plot
> plot(prices$Date[-1], losses, type="l", xlab="", ylab="Loss",
+ main ="Daily Percentage Losses", ylim=range(losses))
```

Daily Percentage Losses



The daily percentage losses do not appear to be an i.i.d sample given the volatility clustering that is apparent in the data.

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```
maxima <- NA
blocks <- split(losses, ceiling(seq_along(losses)/65))
> for(i in 1:80) {maxima[i] <- max(unlist(blocks[i]))}</pre>
```

Question 3

```
parameters <- fit.GEV(maxima)
xi.GEV <- parameters$par.ests[1]
mu.GEV <- parameters$par.ests[2]
sigma.GEV <- parameters$par.ests[3]</pre>
```

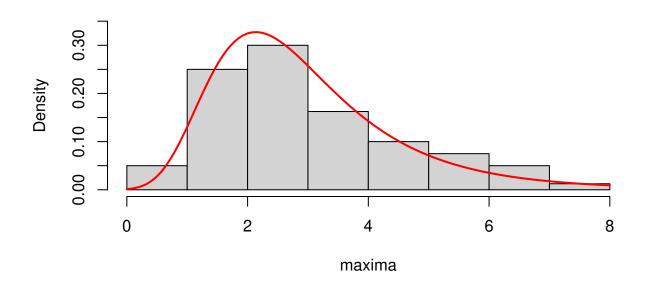
Our estimate of $\xi = 0.09469 > 0$, which would indicate that the distribution of the percentage losses is heavy-tailed.

Question 4

P.T.O

- > hist(maxima, prob=T, ylim=c(0,0.35))
- > curve(dGEV(x, xi=xi.GEV, mu=mu.GEV, sigma=sigma.GEV), lwd=2, col="red", add=T)

Histogram of maxima

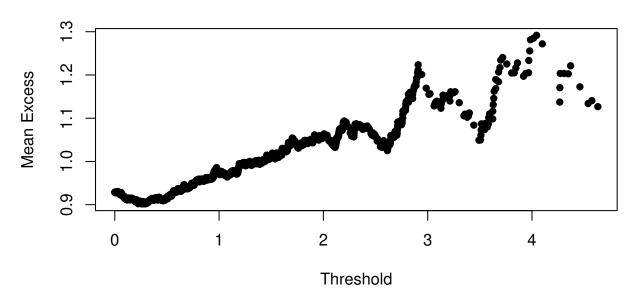


The GEV distribution provides a reasonable fit, although our estimate of ξ may need to be scrutinized given the excess empirical density towards the tail of the distribution.

Question 6

We see that the return period (in quarter-years) of a maximum daily loss of 10% or more is 200, hence we expect this to happen approximately once every 50 years.

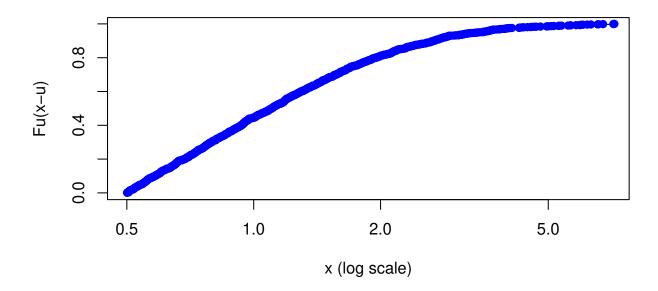




```
fitted.gpd <- fit.GPD(losses, threshold=0.5)
xi.GPD <- fitted.gpd$par.ests[1] ## 0.08468195
beta.GPD <- fitted.gpd$par.ests[2] ## 0.8380901</pre>
```

Our new estimate of $\xi = 0.0847$, which is slightly lower than the GEV maximum likelihood estimate. However, we would still consider this to be indicative of a distribution that exhibits heavy tails.

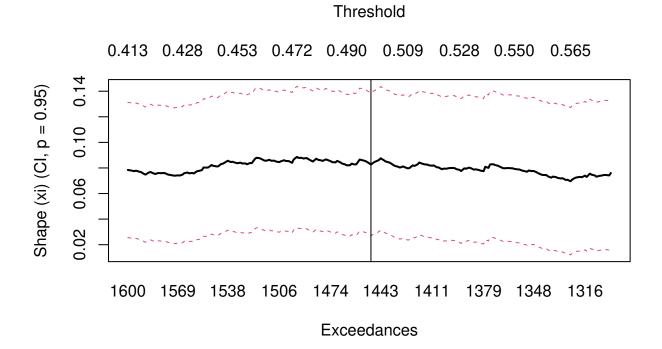
Question 8



The GPD appears to fit the empirical distribution of the excesses above the threshold u = 0.5 very well.

Question 10

```
## (a)
u \leftarrow 0.5; xi \leftarrow xi.GPD; beta \leftarrow beta.GPD; p \leftarrow 0.995;
n <- length(losses); cap.N <- length(losses[losses>u])
val.risk \leftarrow u + (beta/xi)*((n*(1-p)/cap.N)^(-xi)-1)
> val.risk
[1] 4.510376
> quantile(losses, probs=.995)
                                       ## verify
[1] 4.571094
exp.short <- as.numeric(val.risk/(1-xi) + (beta-xi*u)/(1-xi))</pre>
> exp.short
[1] 5.797029
q <- as.numeric(quantile(losses, probs=.995)) ## verify</pre>
empirical.es <- mean(losses[losses>q])
> empirical.es
[1] 5.672348
```



We note that, over the plotted range, our estimator $\hat{\xi}$ is insensitive to the chosen threshold (u=0.5) *i.e.* it does not vary much at all, which is a positive result given the heuristic method we used to determine u.