

# use extreme model

Comfort

8/17/2020

```
library(readxl)
library(extRemes)

## Loading required package: Lmoments
## Loading required package: distillery
##
## Attaching package: 'extRemes'
## The following objects are masked from 'package:stats':
##
##      qqnorm, qqplot
DODOMA_RAIN <- read_excel("DODOMA_RAIN.xlsx")
View(DODOMA_RAIN)
```

## fitting a model assuming stationarity

```
attach(DODOMA_RAIN)
fit1 = fevd(max_rain, type="GEV", method = "MLE")
fit1

##
## fevd(x = max_rain, type = "GEV", method = "MLE")
##
## [1] "Estimation Method used: MLE"
##
##
## Negative Log-Likelihood Value: 337.1421
##
##
## Estimated parameters:
##      location      scale      shape
## 61.82464460 14.90230824 -0.03147776
##
## Standard Error Estimates:
##      location      scale      shape
## 1.9392155 1.4298998 0.1004641
##
## Estimated parameter covariance matrix.
##      location      scale      shape
## location 3.76055683 1.08875506 -0.08211196
```

```
## scale      1.08875506  2.04461335 -0.06009078
## shape     -0.08211196 -0.06009078  0.01009303
##
## AIC = 680.2842
##
## BIC = 687.3925
```

## summary of the fitted model

```
summary.fevd(fit1)
```

```
##
## fevd(x = max_rain, type = "GEV", method = "MLE")
##
## [1] "Estimation Method used: MLE"
##
##
## Negative Log-Likelihood Value: 337.1421
##
##
## Estimated parameters:
##   location      scale      shape
## 61.82464460 14.90230824 -0.03147776
##
## Standard Error Estimates:
##   location      scale      shape
## 1.9392155 1.4298998 0.1004641
##
## Estimated parameter covariance matrix.
##           location      scale      shape
## location  3.76055683  1.08875506 -0.08211196
## scale     1.08875506  2.04461335 -0.06009078
## shape    -0.08211196 -0.06009078  0.01009303
##
## AIC = 680.2842
##
## BIC = 687.3925
```

## Confidence interval for parameters

```
ci(fit1, type = "parameter")
```

```
## fevd(x = max_rain, type = "GEV", method = "MLE")
##
## [1] "Normal Approx."
##
##           95% lower CI      Estimate 95% upper CI
## location  58.0238520 61.82464460 65.6254372
## scale     12.0997562 14.90230824 17.7048603
## shape    -0.2283837 -0.03147776  0.1654282
```

```
distill(fit1, cov=TRUE, FUN="mean")
```

```
##          location          scale          shape          nllh
##    61.82464460    14.90230824    -0.03147776    337.14207708
## location.location  scale.location  shape.location  location.scale
##    3.76055683     1.08875506    -0.08211196     1.08875506
##    scale.scale    shape.scale    location.shape  scale.shape
##    2.04461335    -0.06009078    -0.08211196    -0.06009078
##    shape.shape
##    0.01009303
```

To confirm if the fitted model is stationary

```
is.fixedfevd(fit1)
```

```
## [1] TRUE
```

fitting a non-stationary model

```
attach(DODOMA_RAIN)
```

```
## The following objects are masked from DODOMA_RAIN (pos = 3):
```

```
##
```

```
##    max_rain, year
```

```
fit2 = fevd(max_rain, location.fun=~year,type="GEV", method = "MLE")
```

```
fit2
```

```
##
```

```
## fevd(x = max_rain, location.fun = ~year, type = "GEV", method = "MLE")
```

```
##
```

```
## [1] "Estimation Method used: MLE"
```

```
##
```

```
##
```

```
## Negative Log-Likelihood Value: 336.8901
```

```
##
```

```
##
```

```
## Estimated parameters:
```

```
##          mu0          mu1          scale          shape
```

```
## 25.68592787  0.01831569 15.01216832 -0.03507236
```

```
##
```

```
## Standard Error Estimates:
```

```
##          mu0          mu1          scale          shape
```

```
## 195.44556694  0.09928921  1.57989882  0.12147849
```

```
##
```

```
## Estimated parameter covariance matrix.
```

```
##          mu0          mu1          scale          shape
```

```
## mu0  38198.96964 -19.404680670 -112.90664014 12.769984143
```

```
## mu1   -19.40468  0.009858347  0.05791028 -0.006528674
```

```
## scale -112.90664  0.057910276  2.49608029 -0.103292028
```

```
## shape  12.76998 -0.006528674 -0.10329203  0.014757022
```

```
##
```

```
## AIC = 681.7801
```

```
##  
## BIC = 691.2579
```

The code below is important when we are fitting a non-stationary model

```
findpars(fit2)
```

```
## $location  
##      1      2      3      4      5      6      7      8  
## 61.12680 61.14511 61.16343 61.18174 61.20006 61.21837 61.23669 61.25501  
##      9     10     11     12     13     14     15     16  
## 61.27332 61.29164 61.30995 61.32827 61.34658 61.36490 61.38322 61.40153  
##     17     18     19     20     21     22     23     24  
## 61.41985 61.43816 61.45648 61.47479 61.49311 61.51143 61.52974 61.54806  
##     25     26     27     28     29     30     31     32  
## 61.56637 61.58469 61.60300 61.62132 61.63964 61.65795 61.67627 61.69458  
##     33     34     35     36     37     38     39     40  
## 61.71290 61.73121 61.74953 61.76785 61.78616 61.80448 61.82279 61.84111  
##     41     42     43     44     45     46     47     48  
## 61.85942 61.87774 61.89606 61.91437 61.93269 61.95100 61.96932 61.98763  
##     49     50     51     52     53     54     55     56  
## 62.00595 62.02427 62.04258 62.06090 62.07921 62.09753 62.11584 62.13416  
##     57     58     59     60     61     62     63     64  
## 62.15248 62.17079 62.18911 62.20742 62.22574 62.24405 62.26237 62.28069  
##     65     66     67     68     69     70     71     72  
## 62.29900 62.31732 62.33563 62.35395 62.37226 62.39058 62.40889 62.42721  
##     73     74     75     76     77     78     79  
## 62.44553 62.46384 62.48216 62.50047 62.51879 62.53710 62.55542  
##  
## $scale  
## [1] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [9] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [17] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [25] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [33] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [41] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [49] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [57] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [65] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
## [73] 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217 15.01217  
##  
## $shape  
## [1] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [7] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [13] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [19] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [25] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [31] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [37] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [43] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [49] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [55] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236  
## [61] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236
```

```
## [67] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236
## [73] -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236 -0.03507236
## [79] -0.03507236
```

## Covariate matrix for non-stationary EVD projections

```
make.qcov(fit2)
```

```
##      mu0 mu1 scale shape threshold
## [1,]  1  0     1     1           0
```

## Likelihood Ratio test

```
lr.test(fit1,fit2, alpha = 0.05)
```

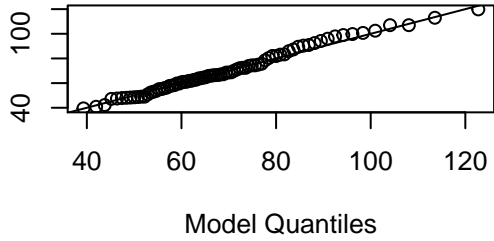
```
##
## Likelihood-ratio Test
##
## data: max_rainmax_rain
## Likelihood-ratio = 0.50402, chi-square critical value = 3.8415, alpha =
## 0.0500, Degrees of Freedom = 1.0000, p-value = 0.4777
## alternative hypothesis: greater
```

## Diagnostic plot for a model assuming stationarity

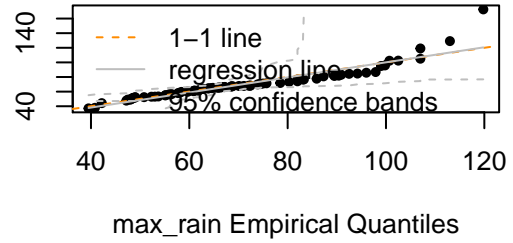
```
plot(fit1)
```

fevd(x = max\_rain, type = "GEV", method = "MLE")

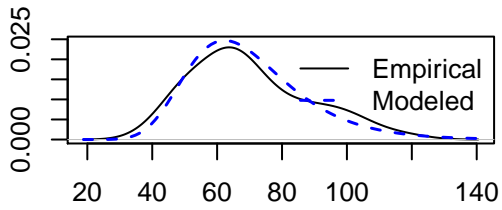
Empirical Quantiles



Quantiles from Model Simulated D



Density

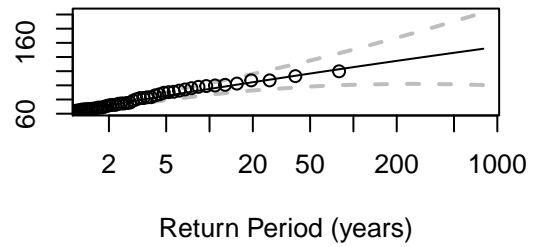


N = 79 Bandwidth = 6.897

## Diagnostic plot for non-stationary model

```
plot(fit2)
```

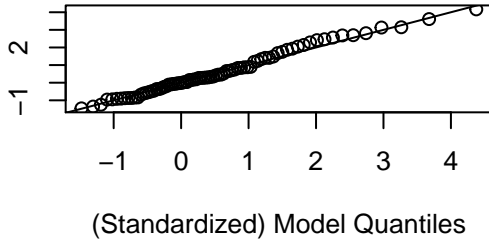
Return Level



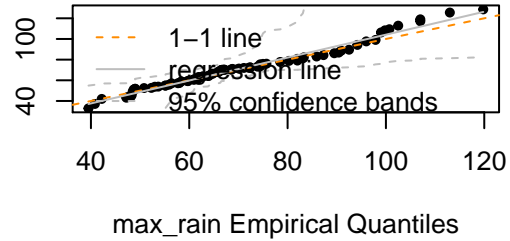
```
fevd(x = max_rain, location.fun = ~year, type = "GEV", method = "MLE")
```

Empirical Residual Quantiles

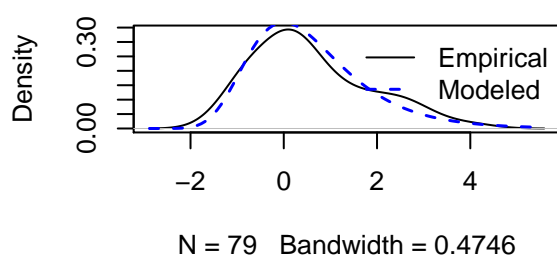
### (Gumbel Scale)



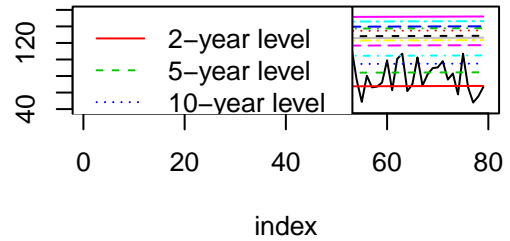
Quantiles from Model Simulated D



### Transformed Data



Return Level



##

Return level estimate for a model assuming stationarity

```
return.level(fit1,return.period = c(10,20,50,100,200))
```

```
## fevd(x = max_rain, type = "GEV", method = "MLE")
## get(paste("return.level.fevd.", newcl, sep = ""))(x = x, return.period = return.period)
##
## GEV model fitted to max_rain
## Data are assumed to be stationary
## [1] "Return Levels for period units in years"
## 10-year level 20-year level 50-year level 100-year level 200-year level
## 94.20009 104.08124 116.54337 125.64532 134.51727
```

```
ci(fit1, alpha=0.05,return.period =10 ,qcov=NULL)
```

```
## fevd(x = max_rain, type = "GEV", method = "MLE")
##
## [1] "Normal Approx."
##
## [1] "10-year return level: 94.2"
##
## [1] "95% Confidence Interval: (86.3823, 102.0179)"
```

### Return level estimate for non-stationary model

```
return.level(fit2,return.period = c(10,20,50,100,200))
```

```
## fevd(x = max_rain, location.fun = ~year, type = "GEV", method = "MLE")
## get(paste("return.level.fevd.", newcl, sep = ""))(x = x, return.period = return.period)
```

```

##
## GEV model fitted to max_rain
## Data are assumed to be non-stationary
## [1] "Return Levels for period units in years"
##      10-year level 20-year level 50-year level 100-year level 200-year level
## [1,] 93.61091      103.4720      115.8720      124.9020      133.6819
## [2,] 93.62923      103.4903      115.8903      124.9204      133.7002
## [3,] 93.64754      103.5086      115.9086      124.9387      133.7185
## [4,] 93.66586      103.5269      115.9269      124.9570      133.7368
## [5,] 93.68418      103.5453      115.9453      124.9753      133.7551
## [6,] 93.70249      103.5636      115.9636      124.9936      133.7735
## [7,] 93.72081      103.5819      115.9819      125.0119      133.7918
## [8,] 93.73912      103.6002      116.0002      125.0303      133.8101
## [9,] 93.75744      103.6185      116.0185      125.0486      133.8284
## [10,] 93.77575     103.6368      116.0368      125.0669      133.8467
## [11,] 93.79407     103.6552      116.0552      125.0852      133.8650
## [12,] 93.81239     103.6735      116.0735      125.1035      133.8834
## [13,] 93.83070     103.6918      116.0918      125.1218      133.9017
## [14,] 93.84902     103.7101      116.1101      125.1401      133.9200
## [15,] 93.86733     103.7284      116.1284      125.1585      133.9383
## [16,] 93.88565     103.7467      116.1467      125.1768      133.9566
## [17,] 93.90396     103.7650      116.1650      125.1951      133.9749
## [18,] 93.92228     103.7834      116.1834      125.2134      133.9933
## [19,] 93.94060     103.8017      116.2017      125.2317      134.0116
## [20,] 93.95891     103.8200      116.2200      125.2500      134.0299
## [21,] 93.97723     103.8383      116.2383      125.2684      134.0482
## [22,] 93.99554     103.8566      116.2566      125.2867      134.0665
## [23,] 94.01386     103.8749      116.2749      125.3050      134.0848
## [24,] 94.03217     103.8933      116.2933      125.3233      134.1031
## [25,] 94.05049     103.9116      116.3116      125.3416      134.1215
## [26,] 94.06881     103.9299      116.3299      125.3599      134.1398
## [27,] 94.08712     103.9482      116.3482      125.3783      134.1581
## [28,] 94.10544     103.9665      116.3665      125.3966      134.1764
## [29,] 94.12375     103.9848      116.3848      125.4149      134.1947
## [30,] 94.14207     104.0031      116.4032      125.4332      134.2130
## [31,] 94.16038     104.0215      116.4215      125.4515      134.2314
## [32,] 94.17870     104.0398      116.4398      125.4698      134.2497
## [33,] 94.19702     104.0581      116.4581      125.4881      134.2680
## [34,] 94.21533     104.0764      116.4764      125.5065      134.2863
## [35,] 94.23365     104.0947      116.4947      125.5248      134.3046
## [36,] 94.25196     104.1130      116.5130      125.5431      134.3229
## [37,] 94.27028     104.1314      116.5314      125.5614      134.3413
## [38,] 94.28859     104.1497      116.5497      125.5797      134.3596
## [39,] 94.30691     104.1680      116.5680      125.5980      134.3779
## [40,] 94.32523     104.1863      116.5863      125.6164      134.3962
## [41,] 94.34354     104.2046      116.6046      125.6347      134.4145
## [42,] 94.36186     104.2229      116.6229      125.6530      134.4328
## [43,] 94.38017     104.2413      116.6413      125.6713      134.4511
## [44,] 94.39849     104.2596      116.6596      125.6896      134.4695
## [45,] 94.41680     104.2779      116.6779      125.7079      134.4878
## [46,] 94.43512     104.2962      116.6962      125.7263      134.5061
## [47,] 94.45343     104.3145      116.7145      125.7446      134.5244
## [48,] 94.47175     104.3328      116.7328      125.7629      134.5427
## [49,] 94.49007     104.3511      116.7512      125.7812      134.5610

```



## [50,]	94.50838	104.3695	116.7695	125.7995	134.5794
## [51,]	94.52670	104.3878	116.7878	125.8178	134.5977
## [52,]	94.54501	104.4061	116.8061	125.8361	134.6160
## [53,]	94.56333	104.4244	116.8244	125.8545	134.6343
## [54,]	94.58164	104.4427	116.8427	125.8728	134.6526
## [55,]	94.59996	104.4610	116.8610	125.8911	134.6709
## [56,]	94.61828	104.4794	116.8794	125.9094	134.6893
## [57,]	94.63659	104.4977	116.8977	125.9277	134.7076
## [58,]	94.65491	104.5160	116.9160	125.9460	134.7259
## [59,]	94.67322	104.5343	116.9343	125.9644	134.7442
## [60,]	94.69154	104.5526	116.9526	125.9827	134.7625
## [61,]	94.70985	104.5709	116.9709	126.0010	134.7808
## [62,]	94.72817	104.5893	116.9893	126.0193	134.7991
## [63,]	94.74649	104.6076	117.0076	126.0376	134.8175
## [64,]	94.76480	104.6259	117.0259	126.0559	134.8358
## [65,]	94.78312	104.6442	117.0442	126.0742	134.8541
## [66,]	94.80143	104.6625	117.0625	126.0926	134.8724
## [67,]	94.81975	104.6808	117.0808	126.1109	134.8907
## [68,]	94.83806	104.6991	117.0992	126.1292	134.9090
## [69,]	94.85638	104.7175	117.1175	126.1475	134.9274
## [70,]	94.87470	104.7358	117.1358	126.1658	134.9457
## [71,]	94.89301	104.7541	117.1541	126.1841	134.9640
## [72,]	94.91133	104.7724	117.1724	126.2025	134.9823
## [73,]	94.92964	104.7907	117.1907	126.2208	135.0006
## [74,]	94.94796	104.8090	117.2090	126.2391	135.0189
## [75,]	94.96627	104.8274	117.2274	126.2574	135.0372
## [76,]	94.98459	104.8457	117.2457	126.2757	135.0556
## [77,]	95.00291	104.8640	117.2640	126.2940	135.0739
## [78,]	95.02122	104.8823	117.2823	126.3124	135.0922
## [79,]	95.03954	104.9006	117.3006	126.3307	135.1105