

VaR Prediction

[Code ▾](#)

Casey Tirshfield

The file `intel_d_logret.txt` contains daily log returns of Intel stock from July 9, 1986 to June 29, 2007. Compute the 99% 1-day and 10-day VaR for a long position of \$1 million using the following methods (as internal models in Section 12.1.2 of the textbook):

[Hide](#)

```
df <- read.table('intel_d_logret.txt', header=FALSE)
colnames(df) <- c('Date', 'logret')
# make dates column into rownames
df$Date <- as.Date(df$Date, '%m/%d/%Y')
row.names(df) <- df$Date
df[1] <- NULL
```

(a) GARCH(1,1) model with standard normal ϵ_t ;

[Hide](#)

```
# first we fit the GARCH(1,1) model to our return series
# including mean
model <- ugarchspec(variance.model=list(model='sGARCH', garchOrder=c(1, 1)), mean.model=list(armaOrder=c(0, 0), include.mean=TRUE), distribution.model='norm')
model_fit <- ugarchfit(spec=model, data=df, solver='hybrid')
print(model_fit)
```

```
*-----*
*          GARCH Model Fit          *
*-----*

Conditional Variance Dynamics
-----
GARCH Model : sGARCH(1,1)
Mean Model  : ARFIMA(0,0,0)
Distribution : norm
```

Optimal Parameters

```
-----
      Estimate Std. Error  t value Pr(>|t|)
mu      0.001159   0.000321   3.6132 0.000302
omega    0.000009   0.000004   2.3628 0.018136
alpha1   0.055866   0.000545 102.5766 0.000000
beta1    0.933970   0.005115 182.5816 0.000000
```

Robust Standard Errors:

```
      Estimate Std. Error  t value Pr(>|t|)
mu      0.001159   0.000330   3.50979 0.000448
omega    0.000009   0.000024   0.36556 0.714690
alpha1   0.055866   0.028248   1.97767 0.047965
beta1    0.933970   0.021394  43.65549 0.000000
```

LogLikelihood : 11906.77

Information Criteria

```
-----
Akaike      -4.4984
Bayes       -4.4934
Shibata     -4.4984
Hannan-Quinn -4.4967
```

Weighted Ljung-Box Test on Standardized Residuals

```
-----
                        statistic p-value
Lag[1]                  3.460 0.06289
Lag[2*(p+q)+(p+q)-1][2] 3.668 0.09261
Lag[4*(p+q)+(p+q)-1][5] 4.693 0.17938
d.o.f=0
H0 : No serial correlation
```

Weighted Ljung-Box Test on Standardized Squared Residuals

```
-----
                        statistic p-value
Lag[1]                  0.01313 0.9088
Lag[2*(p+q)+(p+q)-1][5] 2.12384 0.5895
Lag[4*(p+q)+(p+q)-1][9] 3.81544 0.6203
d.o.f=2
```

Weighted ARCH LM Tests

```
-----
      Statistic Shape Scale P-Value
ARCH Lag[3]      3.179 0.500 2.000 0.07457
ARCH Lag[5]      3.703 1.440 1.667 0.20279
ARCH Lag[7]      4.727 2.315 1.543 0.25358
```

Nyblom stability test

Joint Statistic: 3.5335

Individual Statistics:

mu 0.5698

omega 1.0057

alpha1 0.1989

beta1 0.2134

Asymptotic Critical Values (10% 5% 1%)

Joint Statistic: 1.07 1.24 1.6

Individual Statistic: 0.35 0.47 0.75

Sign Bias Test

 t-value prob sig
Sign Bias 0.5675 0.5704
Negative Sign Bias 0.8468 0.3972
Positive Sign Bias 0.4907 0.6237
Joint Effect 4.1733 0.2433

Adjusted Pearson Goodness-of-Fit Test:

 group statistic p-value(g-1)
1 20 303.5 3.802e-53
2 30 523.1 5.113e-92
3 40 657.5 2.334e-113
4 50 502.2 1.905e-76

Elapsed time : 0.389076

Hide

```
# next we forecast the conditional mean and conditional variance of the assumed param  
etric density 10 days ahead  
forc <- ugarchforecast(model_fit, n.ahead = 10, include.mean=TRUE)  
print(forc)
```

```
*-----*
*           GARCH Model Forecast           *
*-----*
```

```
Model: sGARCH
Horizon: 10
Roll Steps: 0
Out of Sample: 0
```

```
0-roll forecast [T0=2007-07-02]:
```

	Series	Sigma
T+1	0.001159	0.01833
T+2	0.001159	0.01847
T+3	0.001159	0.01861
T+4	0.001159	0.01874
T+5	0.001159	0.01888
T+6	0.001159	0.01901
T+7	0.001159	0.01914
T+8	0.001159	0.01927
T+9	0.001159	0.01939
T+10	0.001159	0.01951

Hide

```
# set 1 and 10 day ahead means and standard deviations from forecast
sigma1 <- 0.01833
sigma10 <- (0.01833^2 + 0.01847^2 + 0.01861^2 + 0.01874^2 + 0.01888^2 + 0.01901^2 + 0.01914^2 + 0.01927^2 + 0.01939^2 + 0.01951^2) ^ (1 / 2)
mu <- 0.001159
long <- 1000000
# compute the 1-day VaR
VaR1 <- (mu - qnorm(0.99) * sigma1) * long
# compute the 10-day VaR
VaR10 <- (10 * mu - qnorm(0.99) * sigma10) * long
# formatting for printing
VaR <- matrix(c(VaR1, VaR10), nrow=1, ncol=2)
rownames(VaR) <- 'VaR'
colnames(VaR) <- c('1-day', '10-day')
print('The 99% 1-day and 10-day VaR for a long position of $1 million using a GARCH(1,1) model with a standard normal residual is:')
```

```
[1] "The 99% 1-day and 10-day VaR for a long position of $1 million using a GARCH(1,1) model with a standard normal residual is:"
```

Hide

```
print(VaR)
```

```
      1-day    10-day
VaR -41482.96 -127734.1
```

[Hide](#)

```
# excluding mean
model_nomean <- ugarchspec(variance.model=list(model='sGARCH', garchOrder=c(1, 1)), m
ean.model=list(armaOrder=c(0, 0), include.mean=FALSE), distribution.model='norm')
model_fit_nomean <- ugarchfit(spec=model_nomean, data=df, solver='hybrid')
print(model_fit)
```

```
*-----*
*          GARCH Model Fit          *
*-----*

Conditional Variance Dynamics
-----
GARCH Model : sGARCH(1,1)
Mean Model  : ARFIMA(0,0,0)
Distribution : norm

Optimal Parameters
-----

      Estimate Std. Error t value Pr(>|t|)
mu      0.001159   0.000321   3.6132 0.000302
omega    0.000009   0.000004   2.3628 0.018136
alpha1   0.055866   0.000545 102.5766 0.000000
beta1    0.933970   0.005115 182.5816 0.000000

Robust Standard Errors:
      Estimate Std. Error t value Pr(>|t|)
mu      0.001159   0.000330   3.50979 0.000448
omega    0.000009   0.000024   0.36556 0.714690
alpha1   0.055866   0.028248   1.97767 0.047965
beta1    0.933970   0.021394 43.65549 0.000000

LogLikelihood : 11906.77

Information Criteria
-----

Akaike      -4.4984
Bayes       -4.4934
Shibata     -4.4984
Hannan-Quinn -4.4967
```

```
Weighted Ljung-Box Test on Standardized Residuals
-----
```

	statistic	p-value
Lag[1]	3.460	0.06289
Lag[2*(p+q)+(p+q)-1][2]	3.668	0.09261
Lag[4*(p+q)+(p+q)-1][5]	4.693	0.17938

d.o.f=0
H0 : No serial correlation

Weighted Ljung-Box Test on Standardized Squared Residuals

	statistic	p-value
Lag[1]	0.01313	0.9088
Lag[2*(p+q)+(p+q)-1][5]	2.12384	0.5895
Lag[4*(p+q)+(p+q)-1][9]	3.81544	0.6203

d.o.f=2

Weighted ARCH LM Tests

	Statistic	Shape	Scale	P-Value
ARCH Lag[3]	3.179	0.500	2.000	0.07457
ARCH Lag[5]	3.703	1.440	1.667	0.20279
ARCH Lag[7]	4.727	2.315	1.543	0.25358

Nyblom stability test

Joint Statistic: 3.5335

Individual Statistics:

mu 0.5698
omega 1.0057
alpha1 0.1989
beta1 0.2134

Asymptotic Critical Values (10% 5% 1%)

Joint Statistic: 1.07 1.24 1.6
Individual Statistic: 0.35 0.47 0.75

Sign Bias Test

	t-value	prob	sig
Sign Bias	0.5675	0.5704	
Negative Sign Bias	0.8468	0.3972	
Positive Sign Bias	0.4907	0.6237	
Joint Effect	4.1733	0.2433	

Adjusted Pearson Goodness-of-Fit Test:

group	statistic	p-value(g-1)
1 20	303.5	3.802e-53
2 30	523.1	5.113e-92
3 40	657.5	2.334e-113

```
4      50      502.2      1.905e-76
```

Elapsed time : 0.389076

Hide

```
# next we forecast the conditional mean and conditional variance of the assumed param  
etric density 10 days ahead  
forc_nomean <- ugarchforecast(model_fit_nomean, n.ahead = 10, include.mean=FALSE)  
print(forc_nomean)
```

```
*-----*  
*      GARCH Model Forecast      *  
*-----*
```

```
Model: sGARCH  
Horizon: 10  
Roll Steps: 0  
Out of Sample: 0
```

```
0-roll forecast [T0=2007-07-02]:
```

	Series	Sigma
T+1	0	0.01839
T+2	0	0.01852
T+3	0	0.01864
T+4	0	0.01877
T+5	0	0.01889
T+6	0	0.01901
T+7	0	0.01913
T+8	0	0.01925
T+9	0	0.01936
T+10	0	0.01948

Hide

```
# set 1 and 10 day ahead means and standard deviations from forecast
sigma1_nomean <- 0.01839
sigma10_nomean <- (0.01839^2 + 0.01852^2 + 0.01864^2 + 0.01877^2 + 0.01889^2 + 0.01901^2 + 0.01913^2 + 0.01925^2 + 0.01936^2 + 0.01948^2) ^ (1 / 2)
mu <- 0
long <- 1000000
# compute the 1-day VaR
VaR1_nomean <- (mu - qnorm(0.99) * sigma1_nomean) * long
# compute the 10-day VaR
VaR10_nomean <- (10 * mu - qnorm(0.99) * sigma10_nomean) * long
# formatting for printing
VaR_nomean <- matrix(c(VaR1_nomean, VaR10_nomean), nrow=1, ncol=2)
rownames(VaR_nomean) <- 'VaR'
colnames(VaR_nomean) <- c('1-day', '10-day')
print('The 99% 1-day and 10-day VaR for a long position of $1 million using a GARCH(1,1) model (no mean) with a standard normal residual is:')
```

```
[1] "The 99% 1-day and 10-day VaR for a long position of $1 million using a GARCH(1,1) model (no mean) with a standard normal residual is:"
```

Hide

```
print(VaR_nomean)
```

```
      1-day  10-day
VaR -42781.54 -139386
```

(b) EGARCH(1, 1) model with standard normal ϵ_t ;

Hide

```
# first we fit the EGARCH(1,1) model to our return series
# including mean
model_e <- ugarchspec(variance.model=list(model='eGARCH', garchOrder=c(1, 1)), mean.model=list(armaOrder=c(0, 0), include.mean=TRUE), distribution.model='norm')
model_fit_e <- ugarchfit(spec=model_e, data=df, solver='hybrid')
print(model_fit_e)
```

```
*-----*
*          GARCH Model Fit          *
*-----*
```

Conditional Variance Dynamics

```
-----
GARCH Model : eGARCH(1,1)
Mean Model  : ARFIMA(0,0,0)
```


Distribution : norm

Optimal Parameters

```
-----
      Estimate Std. Error  t value Pr(>|t|)
mu      0.001051   0.000256   4.1066  4e-05
omega  -0.120069   0.007587 -15.8253  0e+00
alpha1 -0.039050   0.006599  -5.9173  0e+00
beta1   0.983048   0.000979 1003.7206  0e+00
gamma1  0.130575   0.012778  10.2189  0e+00
```

Robust Standard Errors:

```
      Estimate Std. Error  t value Pr(>|t|)
mu      0.001051   0.000221   4.7635 0.000002
omega  -0.120069   0.015158  -7.9211 0.000000
alpha1 -0.039050   0.012456  -3.1350 0.001718
beta1   0.983048   0.001869 525.9191 0.000000
gamma1  0.130575   0.035564   3.6715 0.000241
```

LogLikelihood : 11941.21

Information Criteria

```
-----
Akaike      -4.5110
Bayes       -4.5048
Shibata     -4.5110
Hannan-Quinn -4.5089
```

Weighted Ljung-Box Test on Standardized Residuals

```
-----
                        statistic p-value
Lag[1]                  4.280 0.03856
Lag[2*(p+q)+(p+q)-1][2] 4.532 0.05428
Lag[4*(p+q)+(p+q)-1][5] 5.848 0.09763
d.o.f=0
H0 : No serial correlation
```

Weighted Ljung-Box Test on Standardized Squared Residuals

```
-----
                        statistic p-value
Lag[1]                  0.1203 0.7287
Lag[2*(p+q)+(p+q)-1][5] 2.0842 0.5987
Lag[4*(p+q)+(p+q)-1][9] 3.7439 0.6325
d.o.f=2
```

Weighted ARCH LM Tests

```
-----
      Statistic Shape Scale P-Value
ARCH Lag[3]      3.019 0.500 2.000 0.0823
```

```
ARCH Lag[5]      3.461 1.440 1.667 0.2295
ARCH Lag[7]      4.481 2.315 1.543 0.2824
```

Nyblom stability test

Joint Statistic: 1.9286

Individual Statistics:

mu 0.9913

omega 0.4006

alpha1 0.2223

beta1 0.4529

gamma1 0.1430

Asymptotic Critical Values (10% 5% 1%)

Joint Statistic: 1.28 1.47 1.88

Individual Statistic: 0.35 0.47 0.75

Sign Bias Test

	t-value	prob	sig
Sign Bias	0.68740	0.4919	
Negative Sign Bias	0.13314	0.8941	
Positive Sign Bias	0.05792	0.9538	
Joint Effect	1.31774	0.7249	

Adjusted Pearson Goodness-of-Fit Test:

group	statistic	p-value(g-1)
1 20	274.8	2.806e-47
2 30	481.5	1.840e-83
3 40	677.1	2.250e-117
4 50	576.3	3.854e-91

Elapsed time : 0.5143549

Hide

```
# next we forecast the conditional mean and conditional variance of the assumed param
etric density 10 days ahead
forc_e <- ugarchforecast(model_fit_e, n.ahead = 10, include.mean=TRUE)
print(forc_e)
```

```
*-----*
*           GARCH Model Forecast           *
*-----*
```

```
Model: eGARCH
Horizon: 10
Roll Steps: 0
Out of Sample: 0
```

```
0-roll forecast [T0=2007-07-02]:
```

	Series	Sigma
T+1	0.001051	0.01800
T+2	0.001051	0.01815
T+3	0.001051	0.01829
T+4	0.001051	0.01844
T+5	0.001051	0.01858
T+6	0.001051	0.01872
T+7	0.001051	0.01886
T+8	0.001051	0.01899
T+9	0.001051	0.01913
T+10	0.001051	0.01927

Hide

```
# set 1 and 10 day ahead means and standard deviations from forecast
sigma1_e <- 0.01800
sigma10_e <- (0.01800^2 + 0.01815^2 + 0.01829^2 + 0.01844^2 + 0.01858^2 + 0.01872^2 +
0.01886^2 + 0.01899^2 + 0.01913^2 + 0.01927^2) ^ (1 / 2)
mu_e <- 0.001051
long <- 1000000
# compute the 1-day VaR
VaR1_e <- (mu_e - qnorm(0.99) * sigma1_e) * long
# compute the 10-day VaR
VaR10_e <- (10 * mu_e - qnorm(0.99) * sigma10_e) * long
# formatting for printing
VaR_e <- matrix(c(VaR1_e, VaR10_e), nrow=1, ncol=2)
rownames(VaR_e) <- 'VaR'
colnames(VaR_e) <- c('1-day', '10-day')
print('The 99% 1-day and 10-day VaR for a long position of $1 million using a EGARCH(
1,1) model with a standard normal residual is:')
```

```
[1] "The 99% 1-day and 10-day VaR for a long position of $1 million using a EGARCH(1,
1) model with a standard normal residual is:"
```

Hide

```
print(VaR_e)
```

```
      1-day    10-day
VaR -40823.26 -126670.5
```

[Hide](#)

```
# excluding mean
model_e_nomean <- ugarchspec(variance.model=list(model='eGARCH', garchOrder=c(1, 1)),
mean.model=list(armaOrder=c(0, 0), include.mean=FALSE), distribution.model='norm')
model_fit_e_nomean <- ugarchfit(spec=model_e_nomean, data=df, solver='hybrid')
print(model_fit_e_nomean)
```

```
*-----*
*          GARCH Model Fit          *
*-----*
```

Conditional Variance Dynamics

GARCH Model : eGARCH(1,1)
Mean Model : ARFIMA(0,0,0)
Distribution : norm

Optimal Parameters

	Estimate	Std. Error	t value	Pr(> t)
omega	-0.122277	0.001901	-64.3318	0
alpha1	-0.042214	0.006285	-6.7169	0
beta1	0.982470	0.000442	2222.2968	0
gamma1	0.127764	0.008345	15.3099	0

Robust Standard Errors:

	Estimate	Std. Error	t value	Pr(> t)
omega	-0.122277	0.008244	-14.8314	0.000000
alpha1	-0.042214	0.012193	-3.4623	0.000536
beta1	0.982470	0.001157	849.0387	0.000000
gamma1	0.127764	0.013423	9.5181	0.000000

LogLikelihood : 11935.96

Information Criteria

Akaike	-4.5094
Bayes	-4.5045
Shibata	-4.5094
Hannan-Quinn	-4.5077

Weighted Ljung-Box Test on Standardized Residuals

	statistic	p-value
Lag[1]	4.558	0.03277
Lag[2*(p+q)+(p+q)-1][2]	4.761	0.04714
Lag[4*(p+q)+(p+q)-1][5]	5.838	0.09817

d.o.f=0
H0 : No serial correlation

Weighted Ljung-Box Test on Standardized Squared Residuals

	statistic	p-value
Lag[1]	0.07924	0.7783
Lag[2*(p+q)+(p+q)-1][5]	2.20551	0.5707
Lag[4*(p+q)+(p+q)-1][9]	3.98333	0.5917

d.o.f=2

Weighted ARCH LM Tests

	Statistic	Shape	Scale	P-Value
ARCH Lag[3]	3.248	0.500	2.000	0.07153
ARCH Lag[5]	3.774	1.440	1.667	0.19546
ARCH Lag[7]	4.818	2.315	1.543	0.24358

Nyblom stability test

Joint Statistic: 1.5477

Individual Statistics:

omega 0.4879
alpha1 0.1662
beta1 0.5486
gamma1 0.1720

Asymptotic Critical Values (10% 5% 1%)

Joint Statistic: 1.07 1.24 1.6
Individual Statistic: 0.35 0.47 0.75

Sign Bias Test

	t-value	prob	sig
Sign Bias	0.1892	0.8500	
Negative Sign Bias	0.3380	0.7354	
Positive Sign Bias	0.3378	0.7355	
Joint Effect	0.8146	0.8460	

Adjusted Pearson Goodness-of-Fit Test:

group	statistic	p-value(g-1)
1 20	370.4	6.139e-67
2 30	534.7	2.047e-94
3 40	748.6	4.290e-132

```
4      50      972.9    6.397e-172
```

Elapsed time : 0.750628

Hide

```
# next we forecast the conditional mean and conditional variance of the assumed param  
etric density 10 days ahead  
forc_e_nomean <- ugarchforecast(model_fit_e_nomean, n.ahead = 10, include.mean=FALSE)  
print(forc_e_nomean)
```

```
*-----*  
*      GARCH Model Forecast      *  
*-----*
```

Model: eGARCH

Horizon: 10

Roll Steps: 0

Out of Sample: 0

0-roll forecast [T0=2007-07-02]:

	Series	Sigma
T+1	0	0.01800
T+2	0	0.01817
T+3	0	0.01833
T+4	0	0.01850
T+5	0	0.01866
T+6	0	0.01883
T+7	0	0.01899
T+8	0	0.01915
T+9	0	0.01930
T+10	0	0.01946

Hide

```
# set 1 and 10 day ahead means and standard deviations from forecast
sigma1_e_nomean <- 0.01800
sigma10_e_nomean <- (0.01800^2 + 0.01817^2 + 0.01833^2 + 0.01850^2 + 0.01866^2 + 0.01883^2 + 0.01899^2 + 0.01915^2 + 0.01930^2 + 0.01946^2) ^ (1 / 2)
mu_e_nomean <- 0
long <- 1000000
# compute the 1-day VaR
VaR1_e_nomean <- (mu_e_nomean - qnorm(0.99) * sigma1_e_nomean) * long
# compute the 10-day VaR
VaR10_e_nomean <- (10 * mu_e_nomean - qnorm(0.99) * sigma10_e_nomean) * long
# formatting for printing
VaR_e_nomean <- matrix(c(VaR1_e_nomean, VaR10_e_nomean), nrow=1, ncol=2)
rownames(VaR_e_nomean) <- 'VaR'
colnames(VaR_e_nomean) <- c('1-day', '10-day')
print('The 99% 1-day and 10-day VaR for a long position of $1 million using a EGARCH(1,1) model (nomean) with a standard normal residual is:')
```

```
[1] "The 99% 1-day and 10-day VaR for a long position of $1 million using a EGARCH(1,1) model (nomean) with a standard normal residual is:"
```

Hide

```
print(VaR_e_nomean)
```

```
      1-day      10-day
VaR -41874.26 -137897.2
```

(c) ARMA(1, 1)-GARCH(1, 1) model with ϵ_t having the standardized Student t-distribution whose degrees of freedom are to be estimated from the data;

Hide

```
# first we fit the ARMA(1,1)-GARCH(1,1) model to our return series
model_ag <- ugarchspec(variance.model=list(model='sGARCH', garchOrder=c(1, 1)), mean.model=list(armaOrder=c(1, 1), include.mean=TRUE), distribution.model='std')
model_fit_ag <- ugarchfit(spec=model_ag, data=df, solver='hybrid')
print(model_fit_ag)
```

```
*-----*
*          GARCH Model Fit          *
*-----*

Conditional Variance Dynamics
-----
```

GARCH Model : sGARCH(1,1)
Mean Model : ARFIMA(1,0,1)
Distribution : std

Optimal Parameters

	Estimate	Std. Error	t value	Pr(> t)
mu	0.001099	0.000281	3.9070	0.000093
arl	0.799277	0.103296	7.7377	0.000000
mal	-0.811452	0.100288	-8.0912	0.000000
omega	0.000003	0.000001	2.8625	0.004204
alpha1	0.033806	0.001632	20.7113	0.000000
beta1	0.961630	0.001157	831.1905	0.000000
shape	6.828087	0.547981	12.4604	0.000000

Robust Standard Errors:

	Estimate	Std. Error	t value	Pr(> t)
mu	0.001099	0.000303	3.6221	0.000292
arl	0.799277	0.041144	19.4261	0.000000
mal	-0.811452	0.039850	-20.3626	0.000000
omega	0.000003	0.000002	1.5951	0.110698
alpha1	0.033806	0.001274	26.5300	0.000000
beta1	0.961630	0.000672	1430.4572	0.000000
shape	6.828087	0.711653	9.5947	0.000000

LogLikelihood : 12100.3

Information Criteria

Akaike	-4.5704
Bayes	-4.5617
Shibata	-4.5704
Hannan-Quinn	-4.5674

Weighted Ljung-Box Test on Standardized Residuals

	statistic	p-value
Lag[1]	7.556	5.982e-03
Lag[2*(p+q)+(p+q)-1][5]	7.728	3.572e-08
Lag[4*(p+q)+(p+q)-1][9]	8.054	5.509e-02

d.o.f=2
H0 : No serial correlation

Weighted Ljung-Box Test on Standardized Squared Residuals

	statistic	p-value
Lag[1]	0.7687	0.38061
Lag[2*(p+q)+(p+q)-1][5]	7.4459	0.04038
Lag[4*(p+q)+(p+q)-1][9]	11.5639	0.02276

d.o.f=2

Weighted ARCH LM Tests

```
-----  
                Statistic Shape Scale  P-Value  
ARCH Lag[3]      8.087 0.500 2.000 0.004459  
ARCH Lag[5]     11.175 1.440 1.667 0.003390  
ARCH Lag[7]     12.141 2.315 1.543 0.005742
```

Nyblom stability test

Joint Statistic: 20.6986

Individual Statistics:

```
mu      0.6192  
arl     0.1401  
mal     0.1460  
omega  1.9415  
alpha1 0.3869  
beta1   0.3951  
shape  0.1659
```

Asymptotic Critical Values (10% 5% 1%)

Joint Statistic: 1.69 1.9 2.35

Individual Statistic: 0.35 0.47 0.75

Sign Bias Test

```
-----  
                t-value  prob sig  
Sign Bias      0.57448 0.5657  
Negative Sign Bias 1.56242 0.1182  
Positive Sign Bias 0.03635 0.9710  
Joint Effect    5.99544 0.1118
```

Adjusted Pearson Goodness-of-Fit Test:

```
-----  
group statistic p-value(g-1)  
1    20    263.9    4.724e-45  
2    30    430.4    4.977e-73  
3    40    420.9    1.520e-65  
4    50    422.8    5.945e-61
```

Elapsed time : 0.740963

Hide

```
# next we forecast the conditional mean and conditional variance of the assumed param  
etric density 10 days ahead  
forc_ag <- ugarchforecast(model_fit_ag, n.ahead = 10, include.mean=TRUE)  
print(forc_ag)
```

```
*-----*  
*          GARCH Model Forecast          *  
*-----*  
Model: sGARCH  
Horizon: 10  
Roll Steps: 0  
Out of Sample: 0  
  
0-roll forecast [T0=2007-07-02]:  
      Series  Sigma  
T+1  0.0008492 0.01671  
T+2  0.0008993 0.01676  
T+3  0.0009393 0.01682  
T+4  0.0009713 0.01687  
T+5  0.0009968 0.01692  
T+6  0.0010173 0.01698  
T+7  0.0010336 0.01703  
T+8  0.0010467 0.01708  
T+9  0.0010571 0.01713  
T+10 0.0010654 0.01718
```

Hide

```
# set 1 and 10 day ahead means and standard deviations from forecast
sigma1_ag <- 0.01671
sigma10_ag <- (0.01671^2 + 0.01676^2 + 0.01682^2 + 0.01687^2 + 0.01692^2 + 0.01698^2
+ 0.01703^2 + 0.01708^2 + 0.01713^2 + 0.01718^2) ^ (1 / 2)
mul <- 0.0008492
mul0 <- 0.0008492 + 0.0008993 + 0.0009393 + 0.0009713 + 0.0009968 + 0.0010173 + 0.001
0336 + 0.0010467 + 0.0010571 + 0.0010654
long <- 1000000
dof <- 6.828087
# compute the 1-day VaR
# we use the formula for VaR of a student-t found at the following link http://www.qu
antatrisk.com/2015/12/02/student-t-distributed-linear-value-at-risk/
VaR1_ag <- (mul - qt(0.99, dof) / sqrt(dof / (dof - 2))) * sigma1_ag) * long
# compute the 10-day VaR
VaR10_ag <- (mul0 - qt(0.99, dof) / sqrt(dof / (dof - 2))) * sigma10_ag) * long
# formating for printing
VaR_ag <- matrix(c(VaR1_ag, VaR10_ag), nrow=1, ncol=2)
rownames(VaR_ag) <- 'VaR'
colnames(VaR_ag) <- c('1-day', '10-day')
print('The 99% 1-day and 10-day VaR for a long position of $1 million using a ARMA(1,
1)-GARCH(1,1) model with a standard normal residual is:')
```

```
[1] "The 99% 1-day and 10-day VaR for a long position of $1 million using a ARMA(1,1)
-GARCH(1,1) model with a standard normal residual is:"
```

Hide

```
print(VaR_ag)
```

```
      1-day      10-day
VaR -41572.99 -126190.8
```

(d) the GEV distribution for extreme (negative) returns with subperiod length of 20 trading days.

Hide

```
# we fit the GEV distribution
GEV_fit <- gevFit(data.matrix(df), block=20, type='mle')
print(GEV_fit)
```

Title:

GEV Parameter Estimation

Call:

```
gevFit(x = data.matrix(df), block = 20, type = "mle")
```

Estimation Type:

gev mle

Estimated Parameters:

	xi	mu	beta
	0.10355161	0.03837724	0.01666132

Description

Mon Jul 16 23:29:49 2018

Hide

```
GEV_model <- gev(data.matrix(df), block=20)  
print(GEV_model)
```

\$n.all

[1] 5292

\$n

[1] 265

\$data

[1]	0.05884011	0.08701080	0.05406686	0.05406686	0.07598540	0.02409739
[7]	0.07847109	0.11211655	0.07197302	0.06291341	0.02381049	0.06371539
[13]	0.06136854	0.02272810	0.02927019	0.05358389	0.23638720	0.08829202
[19]	0.05635256	0.06062422	0.05195939	0.05588009	0.04546207	0.02325671
[25]	0.04495109	0.04167242	0.04546207	0.06453809	0.06805301	0.03125233
[31]	0.06453809	0.03125233	0.05884011	0.02777938	0.04445147	0.02469245
[37]	0.03390133	0.03593177	0.04598481	0.03550645	0.03468533	0.04348482
[43]	0.05715803	0.05406686	0.05770793	0.04566973	0.05129295	0.01785750
[49]	0.03745731	0.02247271	0.03745731	0.05505941	0.08167748	0.05348833
[55]	0.06385104	0.03846602	0.03846602	0.04317188	0.07095126	0.06952560
[61]	0.09193688	0.05981810	0.04688327	0.03149846	0.05091975	0.02553313
[67]	0.08777502	0.03419114	0.02553313	0.08438683	0.04598481	0.03815151
[73]	0.05001009	0.03947855	0.03610476	0.06407843	0.04255933	0.02932742
[79]	0.03409398	0.03815151	0.03580921	0.07864260	0.08152606	0.04985423
[85]	0.03593177	0.05089127	0.04240890	0.06221277	0.03922045	0.07347888
[91]	0.02739879	0.03182622	0.04220007	0.02597531	0.03922045	0.02590801
[97]	0.02914094	0.02316690	0.04458361	0.03468533	0.02898734	0.03647797
[103]	0.03342907	0.02638658	0.02950047	0.03217414	0.05866746	0.03580921
[109]	0.04890348	0.03684032	0.04221704	0.03864300	0.02473062	0.06106507
[115]	0.06557489	0.04469988	0.04164249	0.03789332	0.06045370	0.05326667

```
[121] 0.05016692 0.03016800 0.04687146 0.06650605 0.03699534 0.05438656
[127] 0.02469245 0.04434194 0.06171827 0.05328921 0.04173454 0.05872489
[133] 0.05941023 0.03416287 0.03205201 0.05386125 0.03746637 0.03954360
[139] 0.02766835 0.08817124 0.03468012 0.04014353 0.12822664 0.05223534
[145] 0.04847091 0.05111210 0.04050295 0.03325164 0.06271939 0.03724507
[151] 0.04307160 0.05357851 0.03532466 0.07179816 0.06615701 0.04764998
[157] 0.06647049 0.04964190 0.07484579 0.04352670 0.05327595 0.05120046
[163] 0.03714295 0.06969065 0.07192052 0.05760946 0.04577144 0.07135555
[169] 0.07046747 0.08591731 0.12382262 0.05125850 0.04659167 0.05263059
[175] 0.10713557 0.06742487 0.07959439 0.03336085 0.04801682 0.07856953
[181] 0.09374225 0.06978261 0.09634664 0.09716953 0.05082789 0.11571905
[187] 0.18321636 0.05240219 0.06221783 0.07001516 0.04317188 0.09920599
[193] 0.08291839 0.06070199 0.05196353 0.07347888 0.03555619 0.08163321
[199] 0.04849855 0.10293389 0.06477040 0.09624404 0.07608427 0.09080361
[205] 0.05811327 0.08961156 0.05856973 0.03532220 0.06977210 0.03919394
[211] 0.07027109 0.06498616 0.04987150 0.05375336 0.05318583 0.04461832
[217] 0.03694537 0.03957872 0.03899818 0.03440758 0.03255931 0.03494537
[223] 0.02204028 0.04672065 0.03748688 0.04507886 0.01877797 0.03258468
[229] 0.03256645 0.02350576 0.03844604 0.02218777 0.05143306 0.02692809
[235] 0.02794574 0.02609961 0.03031907 0.01497156 0.02191959 0.02661588
[241] 0.01702472 0.01664326 0.02169268 0.03150205 0.03543414 0.02403747
[247] 0.01323864 0.01550408 0.01216616 0.03002576 0.03035273 0.03511197
[253] 0.04013306 0.02629406 0.02477224 0.02821180 0.04115386 0.01411550
[259] 0.03953057 0.01783096 0.01957204 0.02852313 0.02175012 0.02476890
[265] 0.04234087
```

```
$block
```

```
[1] 20
```

```
$par.ests
```

```
      xi      sigma      mu
0.1032800 0.0166641 0.0383838
```

```
$par.ses
```

```
      xi      sigma      mu
0.0450268716 0.0008512921 0.0011510560
```

```
$varcov
```

```
      [,1]      [,2]      [,3]
[1,] 2.027419e-03 -5.652355e-06 -1.630883e-05
[2,] -5.652355e-06 7.246983e-07 4.725179e-07
[3,] -1.630883e-05 4.725179e-07 1.324930e-06
```

```
$converged
```

```
[1] 0
```

```
$nllh.final
```

```
[1] -651.7594
```

```
attr(,"class")
```

```
[1] "gev"
```

[Hide](#)

```
long <- 1000000  
VaR_GEV <- VaR(data.matrix(df), alpha=0.01, type='sample', tail='lower') * long  
print(VaR_GEV)
```

```
      1%  
-75348.93
```