## STAT 349 Final Project Report

### Kenny Jin

```
## Warning: package 'tseries' was built under R version 3.5.3
## Warning: package 'TSA' was built under R version 3.5.3
##
## Attaching package: 'TSA'
## The following objects are masked from 'package:stats':
##
## acf, arima
## The following object is masked from 'package:utils':
##
## tar
```

Note: log based 10 is always used in this project Some of the dataset is replaced because of the length. Current datasets (\* means replaced):

\*BLL F \*LLY LOW MCO MSFT PFE ROK \*TRV \*TXT

### 1 Steps 1-3

#### 1.1 Dataset 1: Stock BLL

#### 1.1.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
## [1] 3740.503
## [1] 1 2
##
## Call:
## garch(x = BLL.return, order = best_params)
##
## Model:
## GARCH(1,2)
##
## Residuals:
##
       Min
                 1Q
                     Median
                                   ЗQ
                                            Max
## -8.12455 -0.49526 0.03929 0.57964 10.91165
##
## Coefficient(s):
##
      Estimate Std. Error t value Pr(>|t|)
## a0 4.892e-06
                 2.768e-07
                             17.675 < 2e-16 ***
## a1 7.677e-02
                 1.227e-02
                              6.258 3.90e-10 ***
## a2 1.441e-01
                 2.072e-02
                               6.954 3.54e-12 ***
## b1 6.926e-01
                 1.611e-02
                             42.980 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
##
## Diagnostic Tests:
## Jarque Bera Test
##
## data: Residuals
## X-squared = 7117.9, df = 2, p-value < 2.2e-16
##
##
## Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.070794, df = 1, p-value = 0.7902</pre>
```

Report the sum of the squared error of the final model.

```
## [1] 3740.503
```

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.1.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.95955, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 7117.9, df = 2, p-value < 2.2e-16
## skewness: 0.1792892
## kurtosis: 6.671514</pre>
```

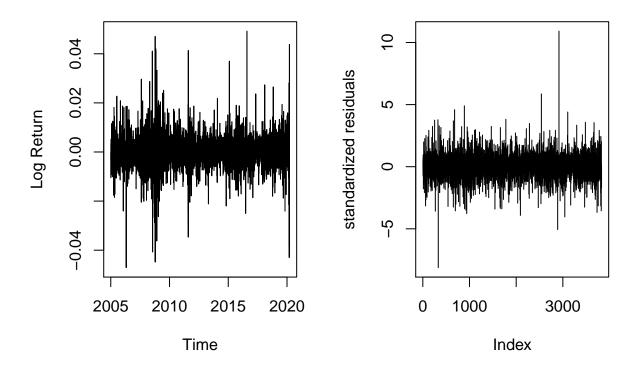


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)

# Series residuals(mpq)^2

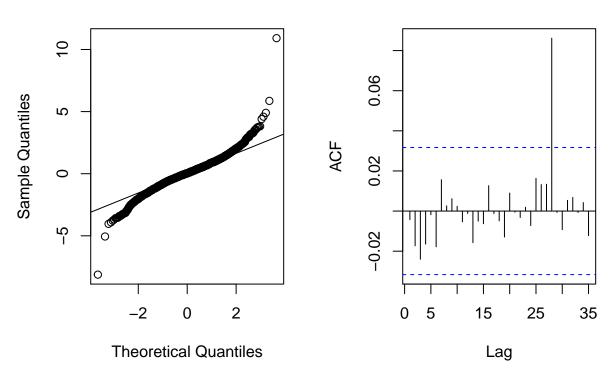


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.2-1.10 Same as 1.1

## 1.2 Dataset 1: Stock F

### 1.2.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
## [1] 3669.49
## [1] 1 2
##
## Call:
## garch(x = F.return, order = best_params)
##
## Model:
   GARCH(1,2)
##
##
   Residuals:
##
##
                                 3Q
       {\tt Min}
                 1Q
                    Median
                                         Max
   -7.0528 -0.5422
                     0.0000
                             0.5238
##
```

```
## Coefficient(s):
##
      Estimate Std. Error t value Pr(>|t|)
## a0 4.217e-06
                 3.797e-07
                              11.11
                                      <2e-16 ***
## a1 2.171e-01
                 1.711e-02
                              12.69
                                      <2e-16 ***
## a2 8.688e-08
                 2.070e-02
                               0.00
                                           1
## b1 7.777e-01
                 1.247e-02
                              62.37
                                      <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
  Jarque Bera Test
##
## data: Residuals
## X-squared = 3814.9, df = 2, p-value < 2.2e-16
##
##
##
  Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.30565, df = 1, p-value = 0.5804
```

Report the sum of the squared error of the final model.

#### ## [1] 3669.49

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

#### 1.2.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.95581, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 3814.9, df = 2, p-value < 2.2e-16
## skewness: -0.4561754
## kurtosis: 4.805406</pre>
```

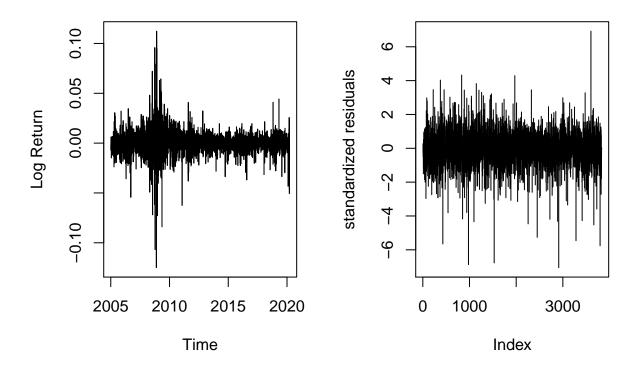


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)

# Series residuals(mpq)^2

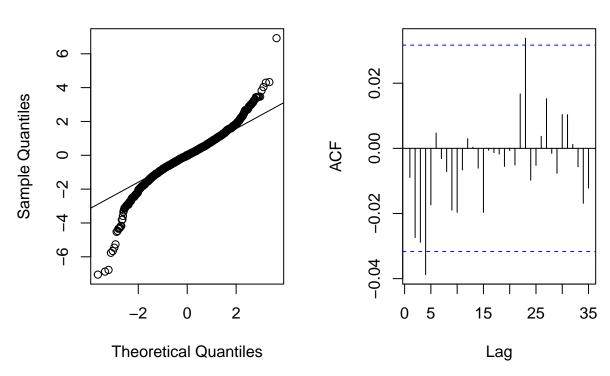


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.3 Dataset 1: Stock LLY

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.3.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
##
   [1] 3684.924
  [1] 1 2
##
## Call:
## garch(x = LLY.return, order = best_params)
##
## Model:
## GARCH(1,2)
##
## Residuals:
##
                  1Q
                       Median
                                     ЗQ
                                             Max
                      0.04902 0.57247
## -8.41665 -0.51273
                                         4.90148
## Coefficient(s):
```

```
##
      Estimate Std. Error t value Pr(>|t|)
## a0 2.276e-06
                 1.848e-07
                              12.31
                                      <2e-16 ***
## a1 1.912e-01
                 2.101e-02
                               9.10
                                      <2e-16 ***
## a2 2.447e-10
                 2.276e-02
                               0.00
                                           1
## b1 7.736e-01
                 1.170e-02
                              66.11
                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
##
  Jarque Bera Test
##
## data: Residuals
## X-squared = 2327.6, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 1.8415, df = 1, p-value = 0.1748
```

Report the sum of the squared error of the final model.

#### ## [1] 3684.924

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.3.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.96634, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 2327.6, df = 2, p-value < 2.2e-16
## skewness: -0.2212471
## kurtosis: 3.794858</pre>
```

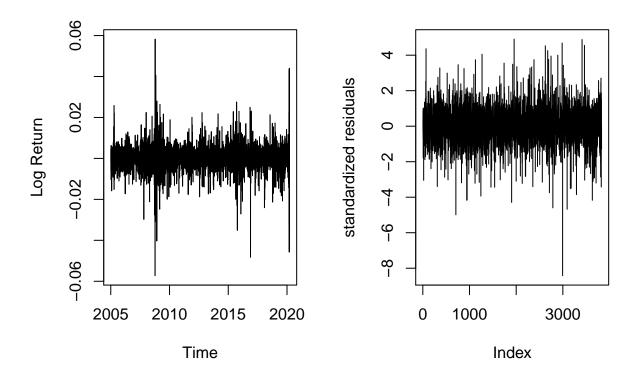
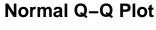


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)



# Series residuals(mpq)^2

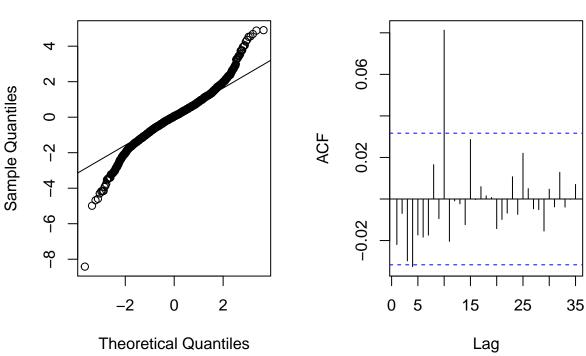


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.4 Dataset 1: Stock LOW

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.4.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
## [1] 3752.975
## [1] 2 2
## Warning in garch(x = LOW.return, order = best_params): singular information
##
## Call:
## garch(x = LOW.return, order = best_params)
##
## Model:
## GARCH(2,2)
##
## Residuals:
```

```
##
                  1Q
                      Median
## -8.54097 -0.53378 0.01902 0.55716 6.63831
##
## Coefficient(s):
##
       Estimate Std. Error t value Pr(>|t|)
## a0 5.449e-06
                                  NA
                         NA
## a1 1.123e-01
                         NA
                                  NA
                                           NA
## a2 1.304e-01
                                  NA
                         NA
                                           NA
## b1 2.257e-01
                         NA
                                  NA
                                           NA
## b2 4.718e-01
                                  NA
                         NA
                                           NA
## Diagnostic Tests:
   Jarque Bera Test
##
## data: Residuals
## X-squared = 3638.7, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.47682, df = 1, p-value = 0.4899
```

Report the sum of the squared error of the final model.

#### ## [1] 3752.975

State the reason why this model is the final model.

GARCH(2,2) gives the smallest sum of squared residuals for this dataset.

### 1.4.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.95904, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 3638.7, df = 2, p-value < 2.2e-16
## skewness: -0.05093237
## kurtosis: 4.775853</pre>
```

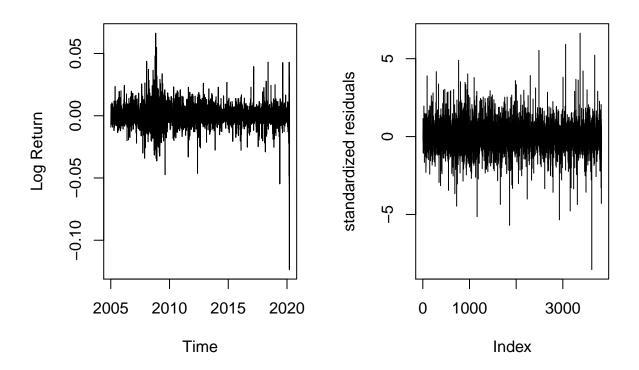


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)



# Series residuals(mpq)^2

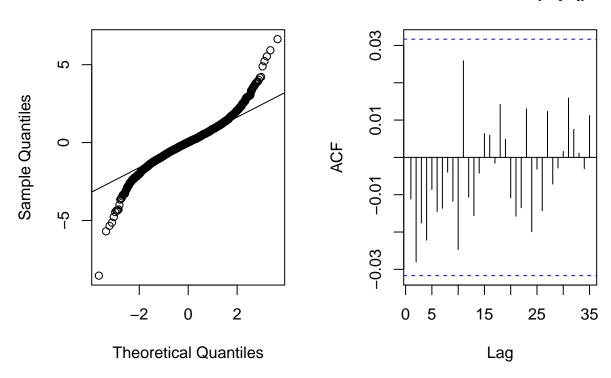


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.5 Dataset 1: Stock MCO

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.5.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
   [1] 3773.204
   [1] 1 2
##
##
## Call:
## garch(x = MCO.return, order = best_params)
##
## Model:
## GARCH(1,2)
##
## Residuals:
##
         Min
                    1Q
                           Median
                                          3Q
                                                   Max
## -10.72984
              -0.46826
                          0.05438
                                    0.58179
                                               7.25452
```

```
##
## Coefficient(s):
##
      Estimate Std. Error t value Pr(>|t|)
## a0 3.643e-06 3.221e-07
                              11.31
                                      <2e-16 ***
## a1 1.645e-01
                 1.309e-02
                              12.56
                                      <2e-16 ***
## a2 5.127e-09
                1.776e-02
                               0.00
                                           1
## b1 8.064e-01
                1.427e-02
                              56.49
                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
  Jarque Bera Test
##
##
## data: Residuals
## X-squared = 14350, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.026958, df = 1, p-value = 0.8696
```

Report the sum of the squared error of the final model.

#### ## [1] 3773.204

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.5.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.93483, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 14350, df = 2, p-value < 2.2e-16
## skewness: -0.5428057
## kurtosis: 9.424122</pre>
```

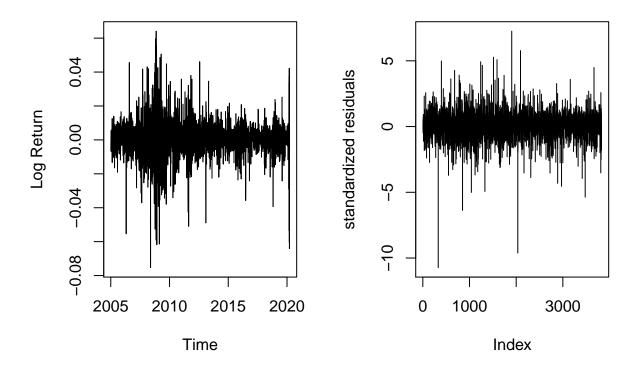


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)



# Series residuals(mpq)^2

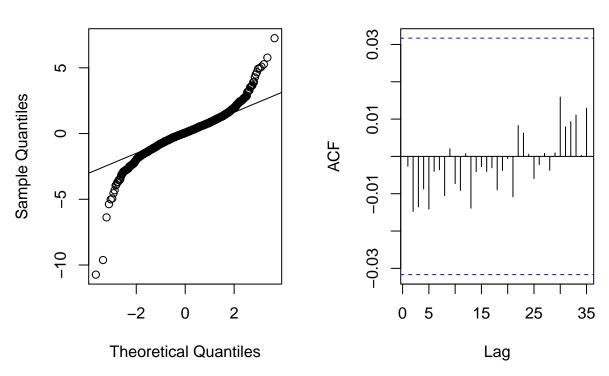


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.6 Dataset 1: Stock MSFT

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.6.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
   [1] 3772.626
   [1] 1 2
##
##
## Call:
## garch(x = MSFT.return, order = best_params)
##
## Model:
## GARCH(1,2)
##
## Residuals:
##
         Min
                    1Q
                           Median
                                          3Q
                                                   Max
## -11.58003
                          0.02266
                                    0.56984
                                              7.05862
              -0.48279
```

```
##
## Coefficient(s):
##
      Estimate Std. Error t value Pr(>|t|)
## a0 4.311e-06 4.075e-07
                              10.58
                                      <2e-16 ***
## a1 1.791e-01
                 1.604e-02
                              11.17
                                      <2e-16 ***
## a2 6.559e-10
                1.890e-02
                               0.00
                                           1
## b1 7.525e-01
                2.019e-02
                              37.28
                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
  Jarque Bera Test
##
##
## data: Residuals
## X-squared = 17052, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.82838, df = 1, p-value = 0.3627
```

Report the sum of the squared error of the final model.

#### ## [1] 3772.626

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.6.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.9334, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 17052, df = 2, p-value < 2.2e-16
## skewness: -0.3523854
## kurtosis: 10.31701</pre>
```

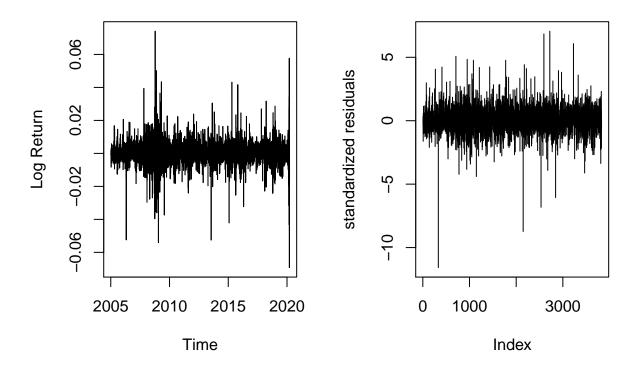


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)



# Series residuals(mpq)^2

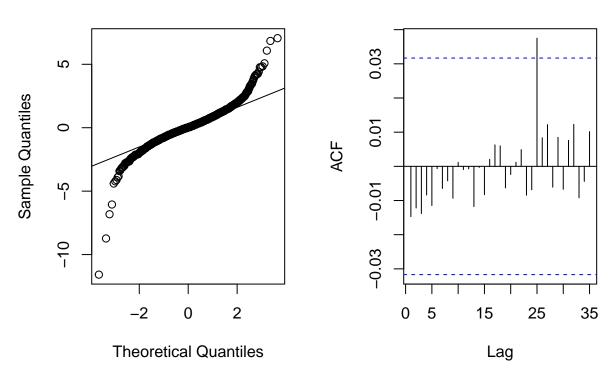


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.7 Dataset 1: Stock PFE

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.7.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
   [1] 3802.875
  [1] 1 2
##
##
## Call:
## garch(x = PFE.return, order = best_params)
##
## Model:
## GARCH(1,2)
##
## Residuals:
##
       Min
                1Q
                    Median
                                        Max
## -9.6496 -0.5520 0.0000
                            0.5713
                                    5.5237
```

```
##
## Coefficient(s):
##
      Estimate Std. Error t value Pr(>|t|)
## a0 1.353e-06 1.609e-07
                                      <2e-16 ***
                              8.411
## a1 1.427e-01
                1.509e-02
                              9.457
                                      <2e-16 ***
## a2 3.126e-10 1.653e-02
                              0.000
                                          1
## b1 8.309e-01
                1.298e-02
                             64.004
                                      <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
  Jarque Bera Test
##
##
## data: Residuals
## X-squared = 5047.9, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.021347, df = 1, p-value = 0.8838
```

Report the sum of the squared error of the final model.

```
## [1] 3802.875
```

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.7.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.96319, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 5047.9, df = 2, p-value < 2.2e-16
## skewness: -0.348042
## kurtosis: 5.583209</pre>
```

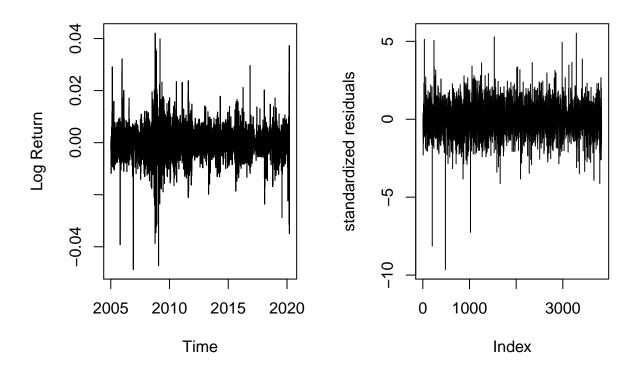


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)



# Series residuals(mpq)^2

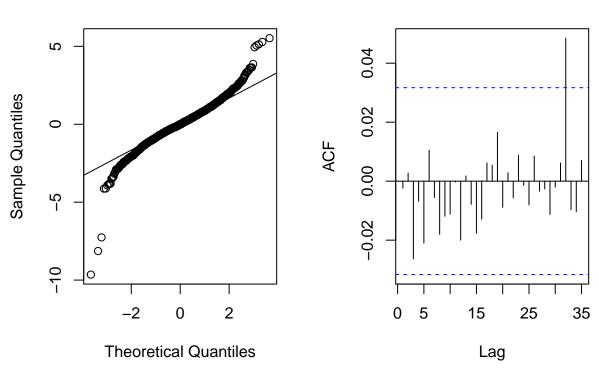


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.8 Dataset 1: Stock ROK

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.8.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
##
   [1] 3706.784
  [1] 2 1
##
## Call:
## garch(x = ROK.return, order = best_params)
##
## Model:
## GARCH(2,1)
##
## Residuals:
                  1Q
                       Median
                                     3Q
                                             Max
                      0.02852 0.55477
## -7.02956 -0.51420
                                         6.39944
## Coefficient(s):
```

```
##
      Estimate Std. Error t value Pr(>|t|)
## a0 6.824e-06
                5.191e-07
                             13.146 < 2e-16 ***
                             16.983 < 2e-16 ***
## a1 2.990e-01
                 1.760e-02
## b1 2.541e-01
                 4.079e-02
                              6.229 4.71e-10 ***
## b2 4.061e-01
                 3.696e-02
                             10.990 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
##
  Jarque Bera Test
##
## data: Residuals
## X-squared = 3282, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 7.5971, df = 1, p-value = 0.005846
```

Report the sum of the squared error of the final model.

#### ## [1] 3706.784

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.8.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.95807, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 3282, df = 2, p-value < 2.2e-16
## skewness: -0.08604433
## kurtosis: 4.533523</pre>
```

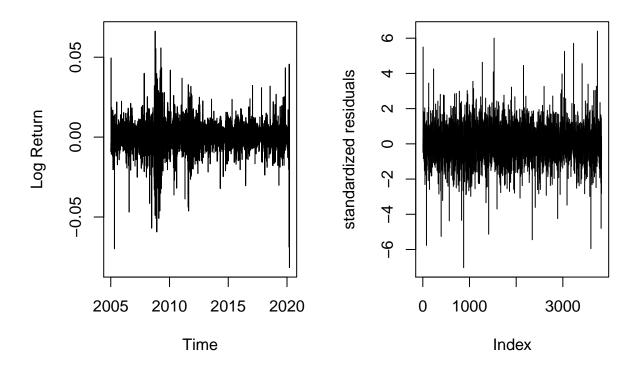


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)

# Series residuals(mpq)^2

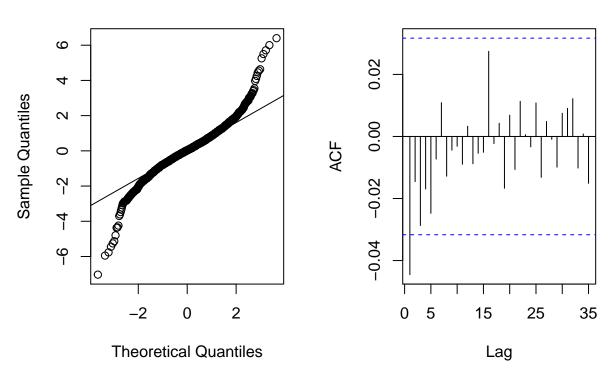


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.9 Dataset 1: Stock TRV

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.9.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
   [1] 3761.931
   [1] 1 2
##
##
## Call:
## garch(x = TRV.return, order = best_params)
##
## Model:
## GARCH(1,2)
##
## Residuals:
        Min
                  1Q
                       Median
                                     3Q
                                             Max
## -7.83779 -0.51256
                      0.04509
                               0.57973
```

```
##
## Coefficient(s):
##
      Estimate Std. Error t value Pr(>|t|)
## a0 8.292e-07 8.897e-08
                             9.320 < 2e-16 ***
## a1 1.012e-01
                1.427e-02
                             7.093 1.31e-12 ***
## a2 2.096e-09 1.602e-02
                              0.000
                                          1
## b1 8.843e-01 7.008e-03 126.187 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
  Jarque Bera Test
##
##
## data: Residuals
## X-squared = 4367.7, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.75537, df = 1, p-value = 0.3848
```

Report the sum of the squared error of the final model.

#### ## [1] 3761.931

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.9.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.95639, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 4367.7, df = 2, p-value < 2.2e-16
## skewness: -0.4956872
## kurtosis: 5.138883</pre>
```

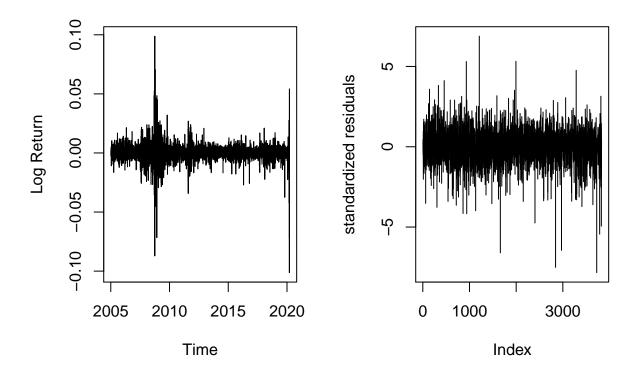


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)

# Series residuals(mpq)^2

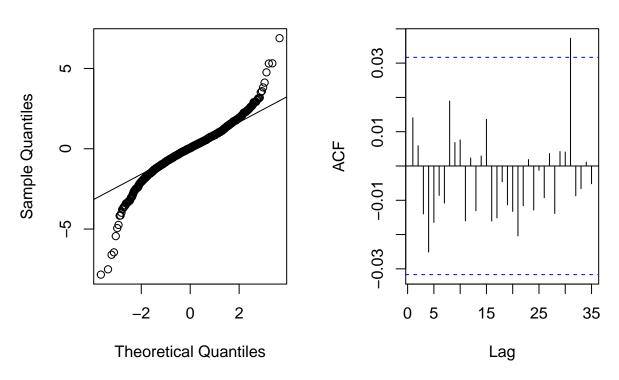


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

### 1.10 Dataset 1: Stock TXT

The original log return series is plotted in Figure on the left panel. The fitted residuals (standardized returns or pseudo-returns) are plotted on the right panel.

### 1.10.1

```
## Warning in garch(x = data, order = c(p, q)): singular information
## Warning in garch(x = data, order = c(p, q)): singular information
   [1] 3816.051
  [1] 1 2
##
##
## Call:
## garch(x = TXT.return, order = best_params)
##
## Model:
## GARCH(1,2)
##
## Residuals:
        Min
                  1Q
                       Median
                                     3Q
                                             Max
## -7.71315 -0.52860
                      0.01008
                               0.57339
                                         8.79217
```

```
##
## Coefficient(s):
##
      Estimate Std. Error t value Pr(>|t|)
                            15.686 < 2e-16 ***
## a0 1.841e-06 1.174e-07
## a1 6.465e-02
                1.286e-02
                             5.029 4.92e-07 ***
## a2 4.805e-02 1.540e-02
                             3.121 0.0018 **
## b1 8.788e-01 5.815e-03 151.124 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
  Jarque Bera Test
##
##
## data: Residuals
## X-squared = 9601.6, df = 2, p-value < 2.2e-16
##
##
##
   Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.14179, df = 1, p-value = 0.7065
```

Report the sum of the squared error of the final model.

#### ## [1] 3816.051

State the reason why this model is the final model.

GARCH(1,2) gives the smallest sum of squared residuals for this dataset.

### 1.10.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.93782, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 9601.6, df = 2, p-value < 2.2e-16
## skewness: 0.08111025
## kurtosis: 7.758057</pre>
```

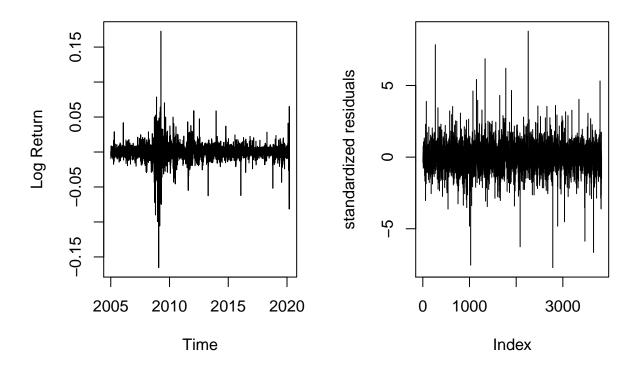
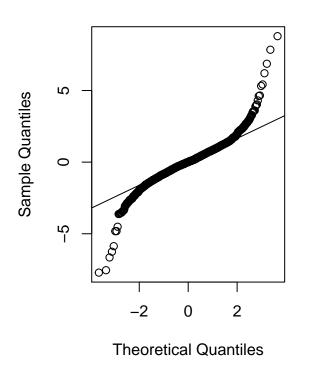


Figure: The left panel is the log return original series from 01/04/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)

# Series residuals(mpq)^2



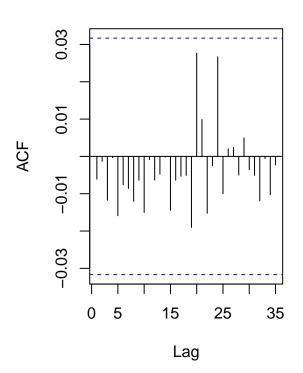


Figure: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

# Step 5

1. Report the sample mean and sample standard deviation  $(c_i = 1/10)$  of  $r_t$ .

## mean: 7.376142e-05

## standard deviation: 0.006107586

2. Report the sample mean and sample standard deviation  $(c_i =?)$  of  $r_t$ .

## Warning: package 'Surrogate' was built under R version 3.5.3

## c\_best: 0.4226344 0.0207333 0.1406372 0.06416519 0.007635655 0.1853899 0.0007749104 0.0756523 0.081

## mean: 0.0001452132

## standard deviation: 0.005593865

3. Report your optimization procedure.

I used Monte Carlo method to simulate the best c vector. I generated a lot of c vectors  $c(c_1, c_2, ..., c_{10})$  that are uniformly distributed in the space where  $c_1 + c_2 + ... + c_{10} = 1$  and  $c_i \ge 0$ . I select the best c vector which generates the largest objective value and satisfy the condition that the std smaller than or equal to the std from c(1/10).

## 3 Step 6

1. Report the sample mean and sample standard deviation  $(c_i = 1/10)$  of  $r_t^{\epsilon}$ .

## mean: 0.01828762

## standard deviation: 0.6505152

2. Report the sample mean and sample standard deviation  $(c_i =?)$  of  $r_t^{\epsilon}$ .

## c\_best: 0.03089828 0.1176029 0.1693533 0.1289134 0.1167062 0.13979 0.09868266 0.07171978 0.07194399

## mean: 0.01835864

## standard deviation: 0.6493505

3. Report your optimization procedure.

I used Monte Carlo method to simulate the best c vector. I generated a lot of c vectors  $c(c_1, c_2, ..., c_{10})$  that are uniformly distributed in the space where  $c_1 + c_2 + ... + c_{10} = 1$  and  $c_i \ge 0$ . I select the best c vector which generates the largest objective value and satisfy the condition that the std smaller than or equal to the std from c(1/10).

## 4 Step 7

1. Report the sample mean and sample standard deviation  $(c_i = 1/10)$  of  $r_t^p$ .

## mean: 0.0001049287

## standard deviation: 0.006093258

2. Report the sample mean and sample standard deviation  $(c_i =?)$  of  $r_t^p$ .

## c best: 0.5358484 0.02768855 0.110723 0.1013489 0.007871983 0.1850356 0.01902716 0.003747687 0.0019

## mean: 0.0001452168

## standard deviation: 0.005444439

3. Report your optimization procedure.

I used Monte Carlo method to simulate the best c vector. I generated a lot of c vectors  $c(c_1, c_2, ..., c_{10})$  that are uniformly distributed in the space where  $c_1 + c_2 + ... + c_{10} = 1$  and  $c_i \ge 0$ . I select the best c vector which generates the largest objective value and satisfy the condition that the std smaller than or equal to the std from c(1/10).

## 5 Step 8

### 5.1 Dataset:

```
r_t^p from c_i=1/10 ## Warning in garch(x = data, order = c(p, q)): singular information ## Warning in garch(x = data, order = c(p, q)): singular information ## [1] 3816.207 ## [1] 2 1
```

#### 5.1.1

```
The best fitted model is GARCH(p,q) with the following estimated parameter values and standard errors.
```

```
## Warning in garch(x = rtp.c1, order = best_params): singular information
##
## Call:
## garch(x = rtp.c1, order = best_params)
##
## Model:
## GARCH(2,1)
##
## Residuals:
                  1Q
                      Median
                                     3Q
## -5.61251 -0.52827 0.07624 0.64961 4.87576
##
## Coefficient(s):
       Estimate Std. Error t value Pr(>|t|)
## a0 5.692e-07
                         NΑ
                                   NA
                                            NΑ
## a1 1.165e-01
                         NA
                                   NA
                                            NA
## b1 8.651e-01
                         NA
                                   NA
                                            NA
## b2 4.384e-09
                         NA
                                   NA
                                            NA
##
## Diagnostic Tests:
    Jarque Bera Test
##
## data: Residuals
## X-squared = 394.43, df = 2, p-value < 2.2e-16
##
##
## Box-Ljung test
##
## data: Squared.Residuals
## X-squared = 0.068531, df = 1, p-value = 0.7935
Report the sum of the squared error of the final model.
## [1] 3816.207
```

State the reason why this model is the final model, instead of GARCH(1,2), GARCH(2,1), or GARCH(2,2).

#### 5.1.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.98631, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 394.43, df = 2, p-value < 2.2e-16
## skewness: -0.3635358</pre>
```

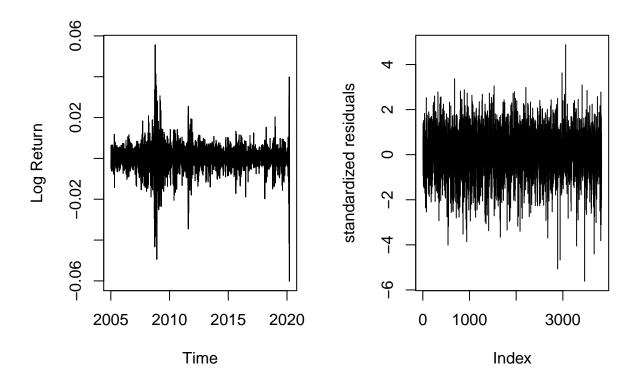


Figure 5.1: The left panel is the log return original series from 01/05/2005 to 03/20/2020. The right panel is the fitted residuals (standardized returns or pseudo- returns.)

# Series residuals(mpq)^2

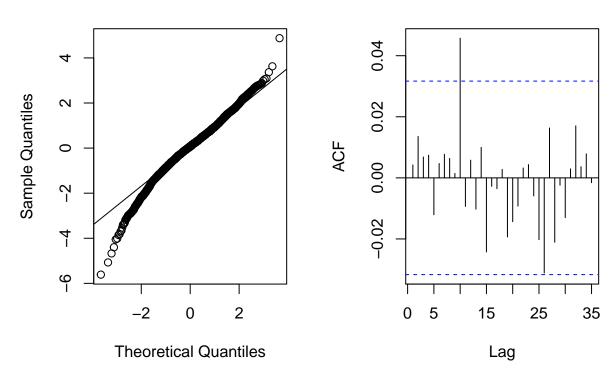


Figure 5.2: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

1. Report the sample mean and sample standard deviation ( $c_i = 1/10$ ) of  $\epsilon_t$ .

## mean: 0.03621546

## standard deviation: 0.9980624

## 6 Step 9

TODO

### 6.1 Dataset:

```
r_t^p from c_i=?

## Warning in garch(x = data, order = c(p, q)): singular information

## Warning in garch(x = data, order = c(p, q)): singular information

## [1] 3744.663

## [1] 2 2

## Warning in garch(x = rtp.cbest, order = best_params): singular information

## ## Call:
```

```
## garch(x = rtp.cbest, order = best_params)
##
## Model:
## GARCH(2,2)
##
## Residuals:
                      Median
       Min
                  1Q
                                    30
                                            Max
## -6.22818 -0.52659 0.06439 0.64123 6.44664
##
## Coefficient(s):
       Estimate Std. Error t value Pr(>|t|)
## a0 1.767e-06
                         NA
                                  NA
## a1 9.870e-02
                         NA
                                  NΑ
                                           NΑ
                         NA
                                  NA
## a2 1.430e-01
                                           NA
## b1 3.215e-01
                         NA
                                  NA
                                           NA
## b2 3.757e-01
                         NA
                                  NA
                                           NA
##
## Diagnostic Tests:
   Jarque Bera Test
##
## data: Residuals
## X-squared = 509.35, df = 2, p-value < 2.2e-16
##
##
##
  Box-Ljung test
## data: Squared.Residuals
## X-squared = 0.07266, df = 1, p-value = 0.7875
```

#### 6.1.1

The best fitted model is GARCH(p,q) with the following estimated parameter values and standard errors. Report the sum of the squared error of the final model.

```
## [1] 3744.663
```

State the reason why this model is the final model, instead of GARCH(1,2), GARCH(2,1), or GARCH(2,2).

### 6.1.2 Some diagnostic results

```
##
## Shapiro-Wilk normality test
##
## data: na.omit(residuals(mpq))
## W = 0.98683, p-value < 2.2e-16
##
## Jarque Bera Test
##
## data: na.omit(residuals(mpq))
## X-squared = 509.35, df = 2, p-value < 2.2e-16
## [1] -0.2150195
## [1] 1.73473</pre>
```

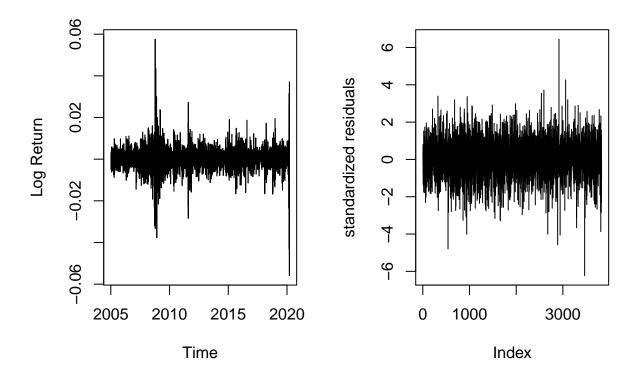


Figure 6.1: The left panel is the log return original series from xx/xx/xxxx to xx/xx/xxxx. The right panel is the fitted residuals (standardized returns or pseudo- returns.)

# Series residuals(mpq)^2

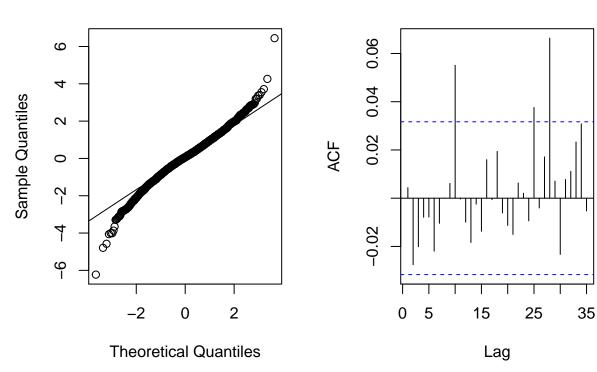


Figure 6.2: The left panel is the residual QQ plot. The right panel is the squared residual ACF plot.

1. Report the sample mean and sample standard deviation ( $c_i = ?$ ) of  $\epsilon_t$ .

## mean: 0.04006906

## standard deviation: 0.9885014

## 7 Step 10

Compare results from Step 5 to Step 9. (Compare the results of the best  $c_i$  and mean values.)

From Step 5 to Step 7, there are 3 distinct best c vectors (Step 8 and 9 are just using the c vector from Step 7). We see that the c\_best vectors from Step 5 and Step 7 are pretty close to each other. The means are also very similar (about 0.00014). This is probably because the formula of r\_t and r\_t^p are very similar and both are computed through similar log returns approach.

However, the c\_best for step 6 is very different from step 5 and 7, and the mean is also much larger than step 5 and 7. This c\_best is actually close to (1/10, ..., 1/10). This is probably because of the fact that the GARCH models did not fit the data so well.