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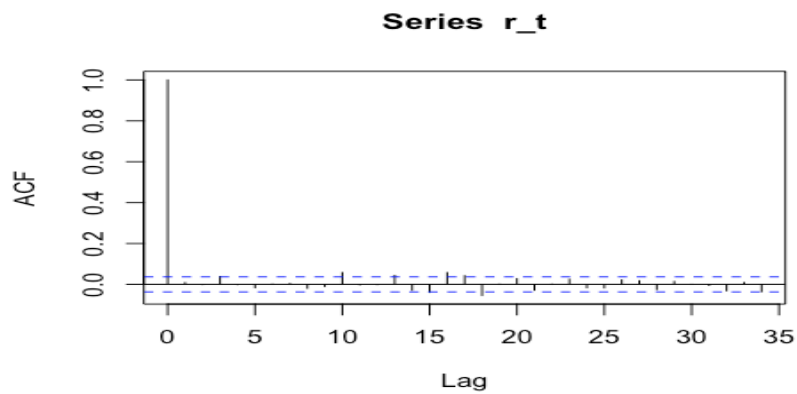
6104640229

Assignment5.R

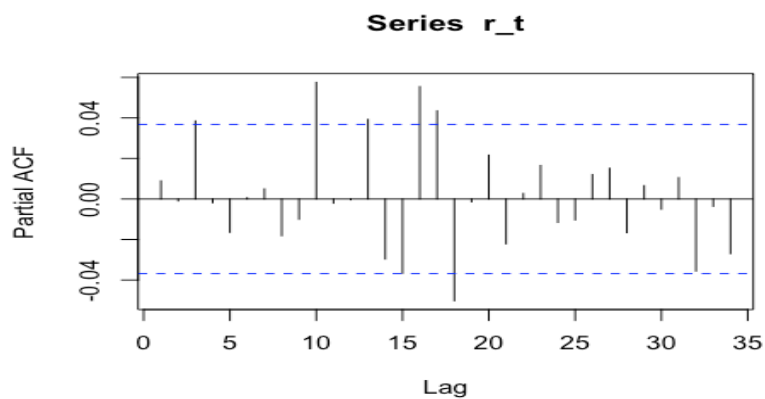
to_on

2021-04-29

```
getSymbols("CAT", from="2006-01-03", to="2017-04-14")  
## [1] "CAT"  
r_t<-diff(log(as.numeric(CAT[,6])))  
#Question1  
#1a  
acf(r_t)
```



```
pacf(r_t)
```



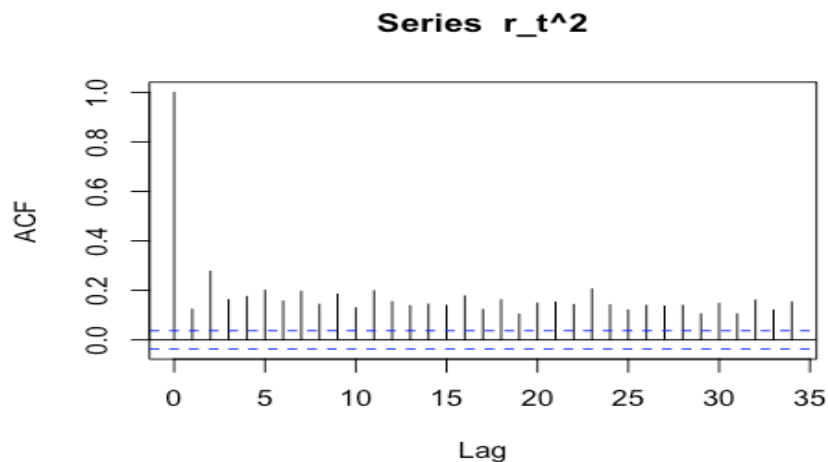
```
Box.test(r_t, lag=10, type='Ljung')
```

```
##  
## Box-Ljung test
```

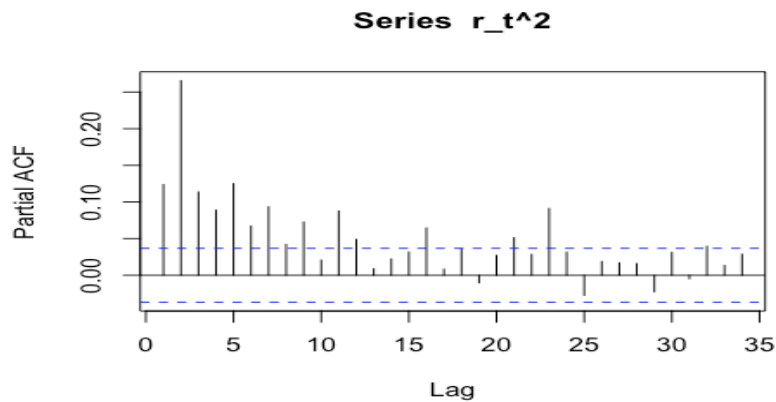
```
##  
## data: r_t  
## X-squared = 16.229, df = 10, p-value = 0.09325
```

Ans1.a) There is no serial correlation in the log return(r_t) with 95% CI. Because the p-value of r_t is greater than 0.05. Then, we can not reject null hypothesis at 0.05 level of significance.

```
#1b  
acf(r_t^2)
```



```
pacf(r_t^2)
```



```
Box.test(r_t^2, lag=10, type='Ljung')
```

```
##  
## Box-Ljung test  
##  
## data: r_t^2  
## X-squared = 917.58, df = 10, p-value < 2.2e-16
```

Ans1b) There exist ARCH effect in log return(r_t) with 95% CI. Because p-value of r_t^2 computed by Ljung-Box test is less than 0.05. Then, we reject the null hypothesis.

#1c

```
m1=garchFit(formula = ~arma(1,0)+garch(1,1),data = r_t,trace=F)
```

```
## Warning: Using formula(x) is deprecated when x is a character vector of length > 1.
```

```
## Consider formula(paste(x, collapse = " ")) instead.
```

```
summary(m1)
```

```
##
```

```
## Title:
```

```
## GARCH Modelling
```

```
##
```

```
## Call:
```

```
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = r_t, trace = F)
```

```
##
```

```
## Mean and Variance Equation:
```

```
## data ~ arma(1, 0) + garch(1, 1)
```

```
## <environment: 0x7fc0d5b78ce8>
```

```
## [data = r_t]
```

```
##
```

```
## Conditional Distribution:
```

```
## norm
```

```
##
```

```
## Coefficient(s):
```

```
##          mu          ar1          omega          alpha1          beta1
```

```
## 4.7490e-04  1.7677e-02  4.4860e-06  4.9755e-02  9.3861e-01
```

```
##
```

```
## Std. Errors:
```

```
## based on Hessian
```

```
##
```

```
## Error Analysis:
```

```
##          Estimate  Std. Error  t value Pr(>|t|)
```

```
## mu      4.749e-04   3.075e-04   1.544 0.122480
```

```
## ar1     1.768e-02   2.004e-02   0.882 0.377774
```

```
## omega   4.486e-06   1.280e-06   3.503 0.000459 ***
```

```
## alpha1  4.976e-02   8.200e-03   6.067 1.3e-09 ***
```

```
## beta1   9.386e-01   1.032e-02  90.934 < 2e-16 ***
```

```
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Log Likelihood:
```

```
## 7381.067 normalized: 2.599883
```

```
##
```

```
## Description:
```

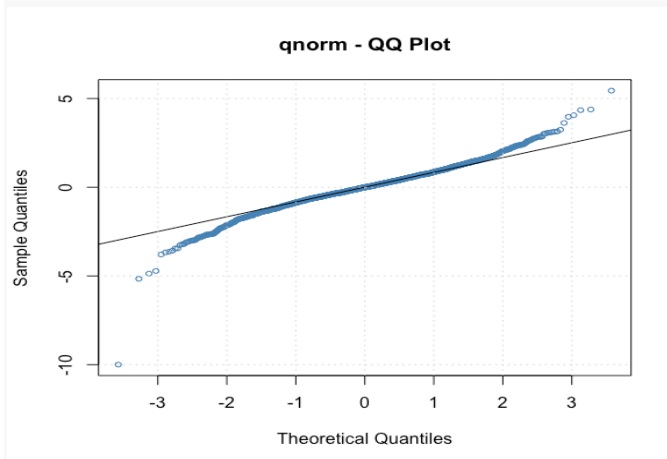
```
## Thu Apr 29 00:29:00 2021 by user:
```

```
##
```

```

##
## Standardised Residuals Tests:
##
##           Statistic p-Value
## Jarque-Bera Test   R      Chi^2 3293.722  0
## Shapiro-Wilk Test  R      W      0.9664334  0
## Ljung-Box Test     R      Q(10) 12.36538  0.2613468
## Ljung-Box Test     R      Q(15) 14.78355  0.4671183
## Ljung-Box Test     R      Q(20) 19.40031  0.4959589
## Ljung-Box Test     R^2    Q(10) 0.9807765 0.9998426
## Ljung-Box Test     R^2    Q(15) 3.684778  0.9986004
## Ljung-Box Test     R^2    Q(20) 6.924134  0.9969265
## LM Arch Test       R      TR^2  2.720581  0.9972158
##
## Information Criterion Statistics:
##           AIC      BIC      SIC      HQIC
## -5.196243 -5.185762 -5.196250 -5.192463

```



```
#plot(m1)
```

The model yield adequate mean & variance equation, but the distribution is not normal.

The fitted model

$$\text{Mean equation: } \hat{r}_t = 4.870e-04(1-0.01768) + 0.01768 r_{t-1}$$

$$(3.075e-04) \quad (2.006e-02)$$

$$\text{variance equation: } \sigma_t^2 = 4.486e-06 + 0.04476 a_{t-1} + 0.9386 \sigma_{t-1}^2$$

$$(1.28e-06) \quad (0.0082) \quad (0.01032)$$

```

#1d
m2=garchFit(formular = ~garch(1,1),data=r_t,cond.dist="std",trace=F)

## Warning: Using formula(x) is deprecated when x is a character vector of
length > 1.
## Consider formula(paste(x, collapse = " ")) instead.

summary(m2)

##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(data = r_t, cond.dist = "std", trace = F, formular = ~garch(1,
## 1))
##
## Mean and Variance Equation:
## data ~ garch(1, 1)
## <environment: 0x7fc0d5e8ca30>
## [data = r_t]
##
## Conditional Distribution:
## std
##
## Coefficient(s):
##      mu      omega      alpha1      beta1      shape
## 5.9008e-04 4.2138e-06 7.2389e-02 9.2028e-01 5.1049e+00
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      5.901e-04 2.703e-04 2.183 0.02904 *
## omega  4.214e-06 1.574e-06 2.678 0.00741 **
## alpha1 7.239e-02 1.375e-02 5.265 1.4e-07 ***
## beta1  9.203e-01 1.474e-02 62.448 < 2e-16 ***
## shape  5.105e+00 4.839e-01 10.549 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 7509.292 normalized: 2.645048
##
## Description:
## Thu Apr 29 00:29:00 2021 by user:
##
##

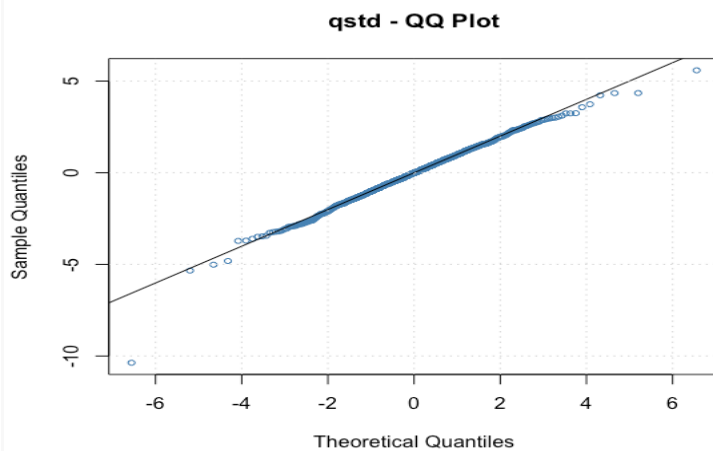
```

```

## Standardised Residuals Tests:
##
##           Statistic p-Value
## Jarque-Bera Test  R    Chi^2  4047.363  0
## Shapiro-Wilk Test R     W      0.9639883  0
## Ljung-Box Test   R     Q(10)  14.87685  0.1366167
## Ljung-Box Test   R     Q(15)  16.83467  0.3288433
## Ljung-Box Test   R     Q(20)  20.67317  0.4165887
## Ljung-Box Test   R^2  Q(10)  2.955969  0.9824413
## Ljung-Box Test   R^2  Q(15)  5.488358  0.9871219
## Ljung-Box Test   R^2  Q(20)  9.459063  0.9769545
## LM Arch Test     R     TR^2   4.275692  0.9779321
##
## Information Criterion Statistics:
##           AIC      BIC      SIC      HQIC
## -5.286574 -5.276093 -5.286580 -5.282793

```

```
#plot(m2)
```



∴ According to the Ljung-box test for both mean and volatility equation, the result of the test show that all of the p-value obtained are greater than 0.05. Then, it can implied that the model is adequate for mean & variance equation as we do not reject any H_0 .

#1e

$$\text{Mean equation: } \hat{r}_t = 0.0007901 \\ (2.7026 \cdot 10^{-4})$$

$$\text{Variance equation: } \hat{\sigma}_t^2 = 4.214 \cdot 10^{-6} + 0.07239 a_{t-1}^2 + 0.9203 r_{t-1}^2 \\ (1.5772 \cdot 10^{-6}) \quad (0.01377) \quad (0.01474)$$

#1f

```
predict(m2,5)
```

```
## meanForecast meanError standardDeviation
## 1 0.0005900796 0.01557857 0.01557857
## 2 0.0005900796 0.01565654 0.01565654
## 3 0.0005900796 0.01573356 0.01573356
## 4 0.0005900796 0.01580965 0.01580965
## 5 0.0005900796 0.01588481 0.01588481
```

#1g

$$1\text{-step: } 0.00059 \pm 1.96(0.01558) = [-0.0299, 0.0311]$$

$$2\text{-step: } 0.00059 \pm 1.96(0.01566) = [-0.0301, 0.0313]$$

$$3\text{-step: } 0.00059 \pm 1.96(0.01573) = [-0.0302, 0.0314]$$

$$4\text{-step: } 0.00059 \pm 1.96(0.01581) = [-0.0304, 0.0316]$$

$$5\text{-step: } 0.00059 \pm 1.96(0.01588) = [-0.0305, 0.0317]$$

Question 2

#Question2

```
m.kovw=read.table("m-kovw-5116.txt",header = T)
```

```
ko=m.kovw[,3]
```

```
logreturnKO=log(ko+1)
```

#2a

```
t.test(logreturnKO)
```

```
##
```

```
## One Sample t-test
```

```
##
```

```
## data: logreturnKO
```

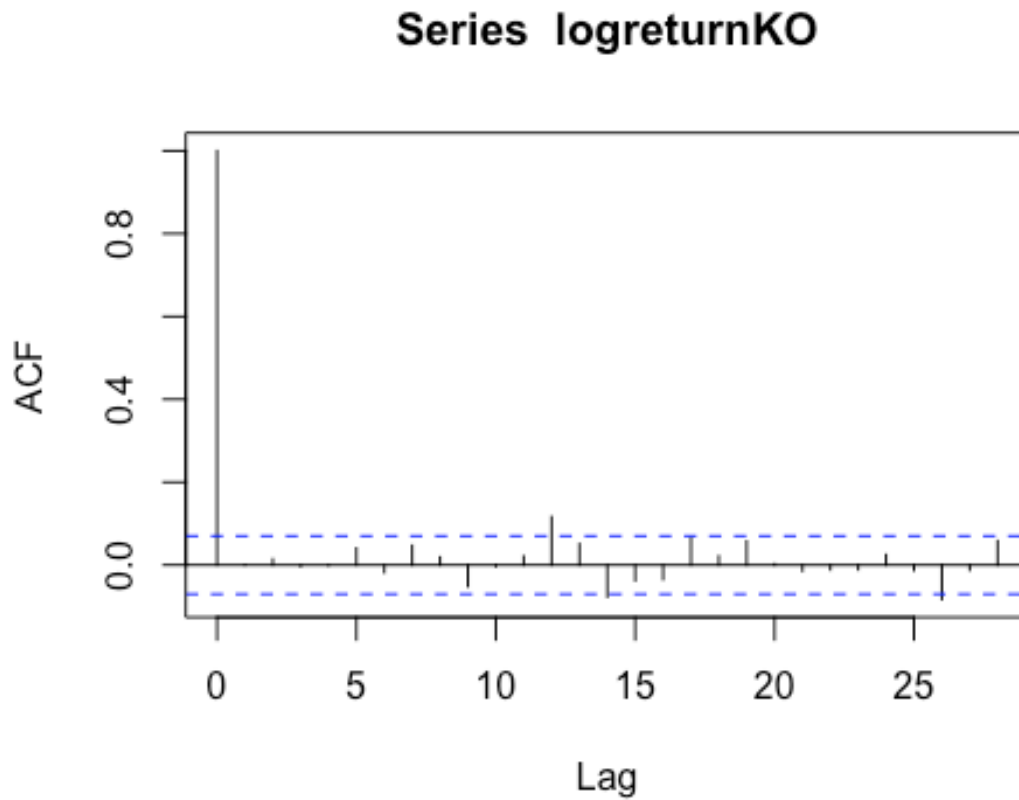
```
## t = 4.9853, df = 779, p-value = 7.628e-07
```

```
## alternative hypothesis: true mean is not equal to 0
```

```
## 95 percent confidence interval:
```

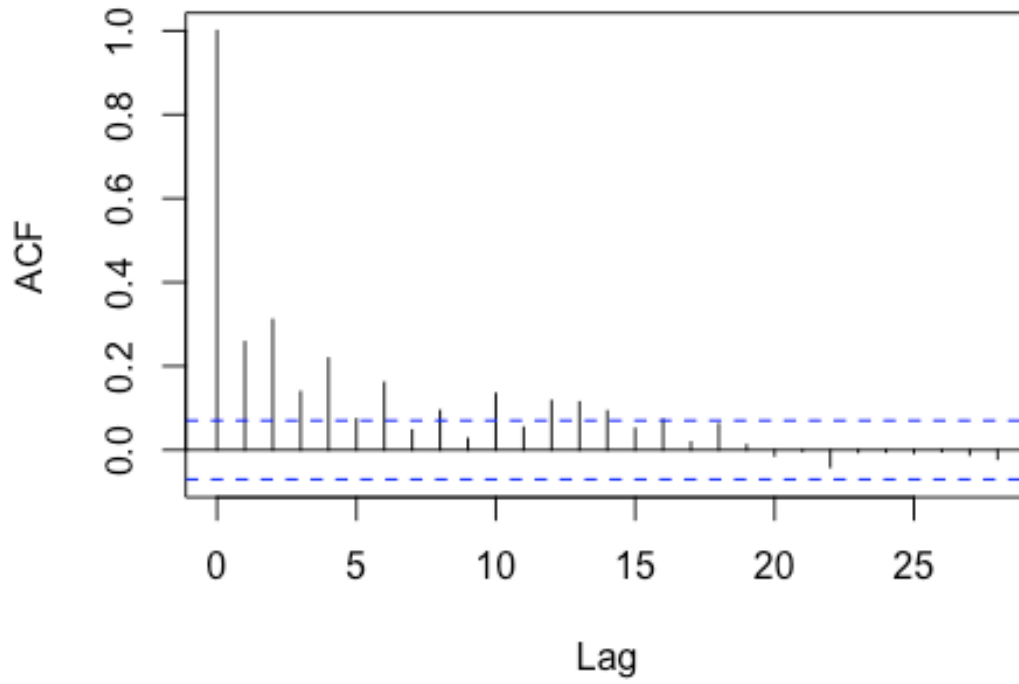
```
## 0.00625636 0.01438347
```

```
## sample estimates:  
## mean of x  
## 0.01031992  
acf(logreturnK0)
```



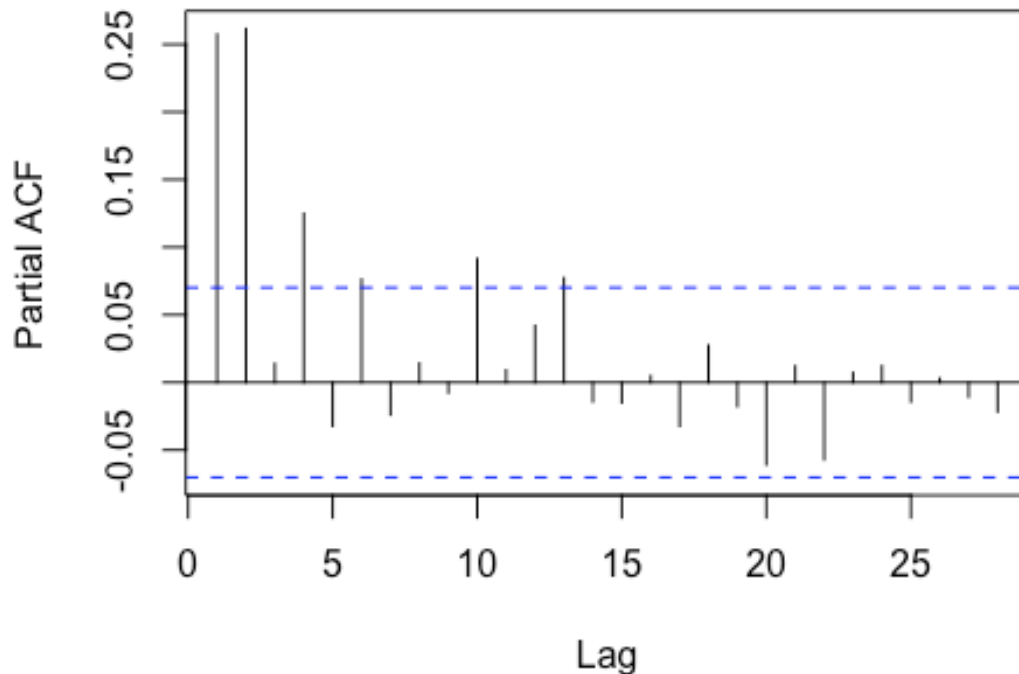
```
Box.test(logreturnK0,lag=10,type = 'Ljung')  
##  
## Box-Ljung test  
##  
## data: logreturnK0  
## X-squared = 5.9201, df = 10, p-value = 0.8219  
acf(logreturnK0^2)
```

Series logreturnKO^2



```
pacf(logreturnKO^2)
```

Series logreturnKO^2



```
Box.test(logreturnKO^2, lag=10, type='Ljung')
```

```
##
## Box-Ljung test
##
## data: logreturnKO^2
## X-squared = 228.23, df = 10, p-value < 2.2e-16
```

Ans. 29

T-test

$$H_0: \mu = 0$$

$$H_1: \mu \neq 0$$

The computed p-value is less than 0.05.

Then, we reject H_0 at 0.05 level of significant.

$\mu \neq 0$ with 95% CI

Ljung-Box test

$$H_0: e_1 = e_2 = e_3 = \dots = e_{10} = 0$$

$$H_1: \exists e_i \neq 0; i = 1, 2, 3, \dots, 10$$

There is no serial correlation in log return KO with 95% CI. because computed p-value from Ljung-Box test is higher than 0.05. Then, we can not reject H_0 at 0.05 level of significant.

Test ARCH effect (logreturnKO²)

Ljung-Box test

$$H_0: \rho_1 = \rho_2 = \rho_3 = \dots = \rho_{10} = 0 / \text{No ARCH effect}$$

$$H_1: \exists \rho_i \neq 0; i = 1, 2, 3, \dots, 10 / \text{then exist ARCH effect}$$

\therefore There is an ARCH effect in log return KO with 95% CI.

Because computed p-value from Ljung-Box test is less than 0.05. Then, we reject H_0 at 0.05 level of significant.

#2b

```
m1=garchFit(formula = ~arma(1,0)+garch(1,1),data = logreturnK0,trace=F)
```

```
## Warning: Using formula(x) is deprecated when x is a character vector of length > 1.
```

```
## Consider formula(paste(x, collapse = " ")) instead.
```

```
summary(m1)
```

```
##
```

```
## Title:
```

```
## GARCH Modelling
```

```
##
```

```
## Call:
```

```
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = logreturnK0,
```

```
## trace = F)
```

```
##
```

```
## Mean and Variance Equation:
```

```
## data ~ arma(1, 0) + garch(1, 1)
```

```
## <environment: 0x7fc0dacff268>
```

```
## [data = logreturnK0]
```

```
##
```

```
## Conditional Distribution:
```

```
## norm
```

```
##
```

```
## Coefficient(s):
```

```
## mu ar1 omega alpha1 beta1
```

```
## 0.01124544 -0.02633742 0.00018112 0.09535029 0.84861593
```

```
##
```

```
## Std. Errors:
```

```
## based on Hessian
```

```
##
```

```
## Error Analysis:
```

```
## Estimate Std. Error t value Pr(>|t|)
```

```
## mu 1.125e-02 1.897e-03 5.929 3.05e-09 ***
```

```
## ar1 -2.634e-02 3.881e-02 -0.679 0.49740
```

```
## omega 1.811e-04 5.852e-05 3.095 0.00197 **
```

```
## alpha1 9.535e-02 1.915e-02 4.978 6.42e-07 ***
```

```
## beta1 8.486e-01 2.766e-02 30.675 < 2e-16 ***
```

```
## ---
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
```

```
## Log Likelihood:
```

```
## 1170.664 normalized: 1.500852
```

```
##
```

```
## Description:
```

```
## Thu Apr 29 00:29:00 2021 by user:
```

```
##
```

```
##
```

```
## Standardised Residuals Tests:
```

```

##                               Statistic p-Value
## Jarque-Bera Test      R      Chi^2  92.91946  0
## Shapiro-Wilk Test    R      W      0.9857081 6.655604e-07
## Ljung-Box Test      R      Q(10)  9.306169  0.5033144
## Ljung-Box Test      R      Q(15)  22.9901   0.0843502
## Ljung-Box Test      R      Q(20)  27.44814  0.1231201
## Ljung-Box Test      R^2    Q(10)  12.63377  0.2448749
## Ljung-Box Test      R^2    Q(15)  13.62088  0.5544545
## Ljung-Box Test      R^2    Q(20)  15.19817  0.7649584
## LM Arch Test        R      TR^2   10.65102  0.5590389
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -2.988883 -2.959016 -2.988965 -2.977396

```

\therefore Mean & variance equation is adequate but the distribution is not normal.

$$\text{Mean equation: } \hat{r}_t = 0.0125(1 + 0.02634) - 0.02634 r_{t-1}$$

(0.0019) (0.0391)

$$\text{Variance equation: } \hat{\sigma}_t^2 = 0.0001811 + 0.09535 a_{t-1}^2 + 0.8486 \hat{\sigma}_{t-1}^2$$

(5.9526-05) (0.01915) (0.02766)

```

#2c
m2=garchFit(formula =
~arma(1,0)+garch(1,1),data=logreturnK0,cond.dist="std",trace=F)

## Warning: Using formula(x) is deprecated when x is a character vector of
length > 1.
## Consider formula(paste(x, collapse = " ")) instead.

summary(m2)

##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = logreturnK0,
## cond.dist = "std", trace = F)
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + garch(1, 1)
## <environment: 0x7fc0d8ca8228>
## [data = logreturnK0]
##
## Conditional Distribution:
## std
##
## Coefficient(s):
##          mu          ar1          omega          alpha1          beta1
shape
## 0.01124020 -0.01887601  0.00017395  0.09642928  0.85044150
7.47877780
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate  Std. Error  t value Pr(>|t|)
## mu      1.124e-02  1.810e-03   6.211 5.27e-10 ***
## ar1     -1.888e-02  3.691e-02  -0.511 0.60904
## omega   1.739e-04  6.596e-05   2.637 0.00836 **
## alpha1  9.643e-02  2.338e-02   4.124 3.72e-05 ***
## beta1   8.504e-01  3.267e-02  26.028 < 2e-16 ***
## shape   7.479e+00  1.840e+00   4.066 4.79e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 1184.863 normalized: 1.519055
##
## Description:

```

```

## Thu Apr 29 00:29:00 2021 by user:
##
##
## Standardised Residuals Tests:
##
##           Statistic p-Value
## Jarque-Bera Test   R    Chi^2  93.6433  0
## Shapiro-Wilk Test  R     W      0.9857385 6.832848e-07
## Ljung-Box Test     R    Q(10)  8.966733 0.5352637
## Ljung-Box Test     R    Q(15) 22.44818 0.09657967
## Ljung-Box Test     R    Q(20) 26.86769 0.1390276
## Ljung-Box Test     R^2  Q(10) 12.48941 0.2536355
## Ljung-Box Test     R^2  Q(15) 13.37442 0.5734021
## Ljung-Box Test     R^2  Q(20) 14.90709 0.7816988
## LM Arch Test       R     TR^2 10.48089 0.5738501
##
## Information Criterion Statistics:
##           AIC      BIC      SIC      HQIC
## -3.022725 -2.986885 -3.022843 -3.008941

```

∴ According to the Ljung-box test for b-th mean and volatility equation, the result of the test show that all of the p-value obtained are greater than 0.05. Then, it can implied that the model is adequate for mean & variance equation as we do not reject any H_0 .

$$\text{Mean equation: } \hat{r}_t = 0.0124(1 + 0.01888) - 0.01888 r_{t-1} \\ (0.00181)$$

$$\text{variance equation: } \hat{\sigma}_t^2 = 1.739 \times 10^{-09} + 0.09643 a_{t-1}^2 + 0.08807 \hat{\sigma}_{t-1}^2 \\ (6.196 \times 10^{-05}) \quad (0.02378) \quad (0.03267)$$

#2d

```
m3=garchFit(formula = ~garch(1,1),data = logreturnKO,trace=F)
```

```
## Warning: Using formula(x) is deprecated when x is a character vector of length > 1.
```

```
## Consider formula(paste(x, collapse = " ")) instead.
```

```
summary(m3)
```

```
##
```

```
## Title:
```

```
## GARCH Modelling
```

```

##
## Call:
## garchFit(formula = ~garch(1, 1), data = logreturnK0, trace = F)
##
## Mean and Variance Equation:
## data ~ garch(1, 1)
## <environment: 0x7fc0da964ba0>
## [data = logreturnK0]
##
## Conditional Distribution:
## norm
##
## Coefficient(s):
##      mu      omega      alpha1      beta1
## 0.01098417 0.00018497 0.09479925 0.84780406
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      1.098e-02  1.846e-03   5.950 2.68e-09 ***
## omega  1.850e-04  5.899e-05   3.135 0.00172 **
## alpha1 9.480e-02  1.912e-02   4.958 7.11e-07 ***
## beta1  8.478e-01  2.787e-02  30.416 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 1170.393    normalized: 1.500504
##
## Description:
## Thu Apr 29 00:29:00 2021 by user:
##
##
## Standardised Residuals Tests:
##
##      Statistic p-Value
## Jarque-Bera Test R Chi^2 95.07163 0
## Shapiro-Wilk Test R W 0.9856773 6.481596e-07
## Ljung-Box Test R Q(10) 8.125181 0.6166108
## Ljung-Box Test R Q(15) 21.27199 0.128362
## Ljung-Box Test R Q(20) 25.62765 0.1784646
## Ljung-Box Test R^2 Q(10) 12.90586 0.228983
## Ljung-Box Test R^2 Q(15) 13.87463 0.5350581
## Ljung-Box Test R^2 Q(20) 15.35522 0.755734
## LM Arch Test R TR^2 10.96004 0.532346
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -2.990752 -2.966858 -2.990804 -2.981562

```

∴ Mean & Variance is adequate but the distribution is not normal.

Mean equation: $\hat{r}_t = 0.01098 + 0.00018497 A_{t-1}^2 + 0.09479925 r_{t-1} + 0.84780406 r_{t-1}$

variance equation: $\hat{\sigma}_t^2 = 1.850e-04 + 0.09479925 A_{t-1}^2 + 0.84780406 \sigma_{t-1}^2$

```

#2e
m4=garchFit(formula = ~garch(1,1),data=logreturnK0,cond.dist="std",trace=F)

## Warning: Using formula(x) is deprecated when x is a character vector of
length > 1.
## Consider formula(paste(x, collapse = " ")) instead.

summary(m4)

##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~garch(1, 1), data = logreturnK0, cond.dist = "std",
## trace = F)
##
## Mean and Variance Equation:
## data ~ garch(1, 1)
## <environment: 0x7fc0d2e03db0>
## [data = logreturnK0]
##
## Conditional Distribution:
## std
##
## Coefficient(s):
##      mu      omega      alpha1      beta1      shape
## 0.01105016 0.00017528 0.09632874 0.85006800 7.48604505
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      1.105e-02 1.757e-03 6.291 3.16e-10 ***
## omega  1.753e-04 6.627e-05 2.645 0.00817 **
## alpha1 9.633e-02 2.337e-02 4.123 3.75e-05 ***
## beta1  8.501e-01 3.277e-02 25.941 < 2e-16 ***
## shape  7.486e+00 1.840e+00 4.069 4.72e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 1184.68 normalized: 1.518821
##
## Description:
## Thu Apr 29 00:29:00 2021 by user:
##
##

```

```

## Standardised Residuals Tests:
##
## Jarque-Bera Test   R      Chi^2  95.31715  0
## Shapiro-Wilk Test  R      W      0.9857263 6.761141e-07
## Ljung-Box Test     R      Q(10)  8.228765  0.6065024
## Ljung-Box Test     R      Q(15)  21.34759  0.1260864
## Ljung-Box Test     R      Q(20)  25.67699  0.1767469
## Ljung-Box Test     R^2    Q(10)  12.61146  0.2462139
## Ljung-Box Test     R^2    Q(15)  13.4693   0.5660982
## Ljung-Box Test     R^2    Q(20)  14.93694  0.7800047
## LM Arch Test       R      TR^2   10.62989  0.560875
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -3.024822 -2.994954 -3.024903 -3.013334

```

$$\text{Mean equation: } \hat{r}_t = 0.0105 \\ (0.001757)$$

$$\text{Variance equation: } \hat{\sigma}_t^2 = 1.753e-04 + 0.09673 a_{t-1}^2 \\ (6.627e-05) \quad (0.02337) \\ + 0.8501 b_{t-1}^2 \\ (0.0327)$$

∴ According to the Ljung-box test for both mean and volatility equation, the result of the test show that all of the p-value obtained are greater than 0.05. Then, it can implied that the model is adequate for mean & variance equation as we do not reject any H_0 .

#2f

Model m_4 (GARCH(1,1) with t -dist²) is to be selected from its 1) Model adequacy for both mean & variance 2) significant coefficient estimates 3) lowest AIC & BIC among m_1, m_2, m_3, m_4 .

```
#Question3
getSymbols("^GSPC",from="2005-01-02",to="2021-03-31")

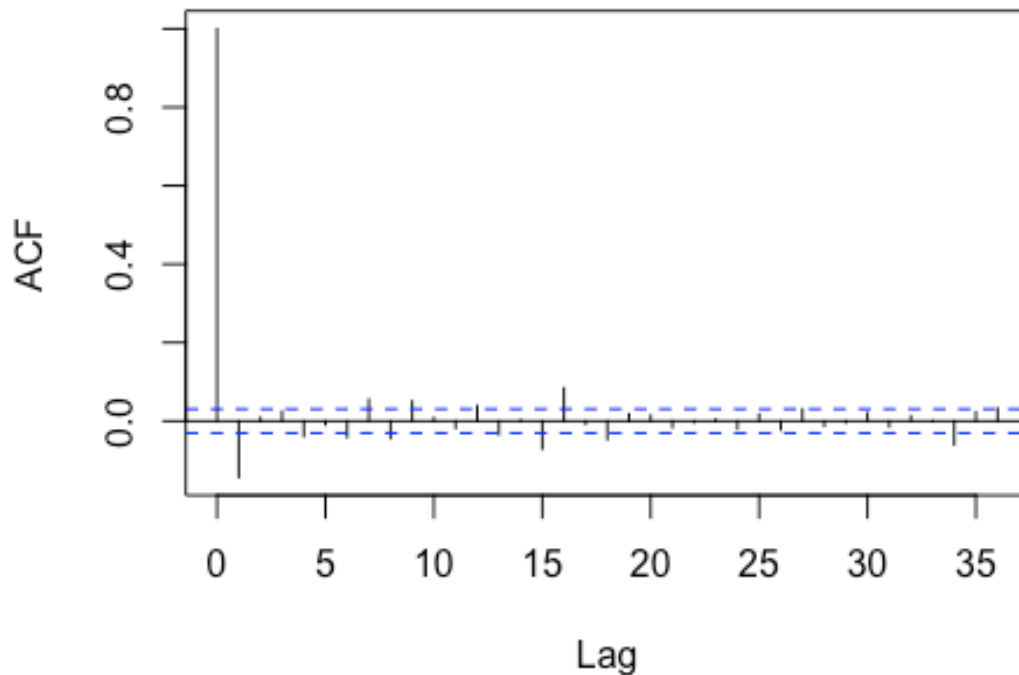
## [1] "^GSPC"

rt<-diff(log(as.numeric(GSPC[,6])))
#3a
t.test(rt)

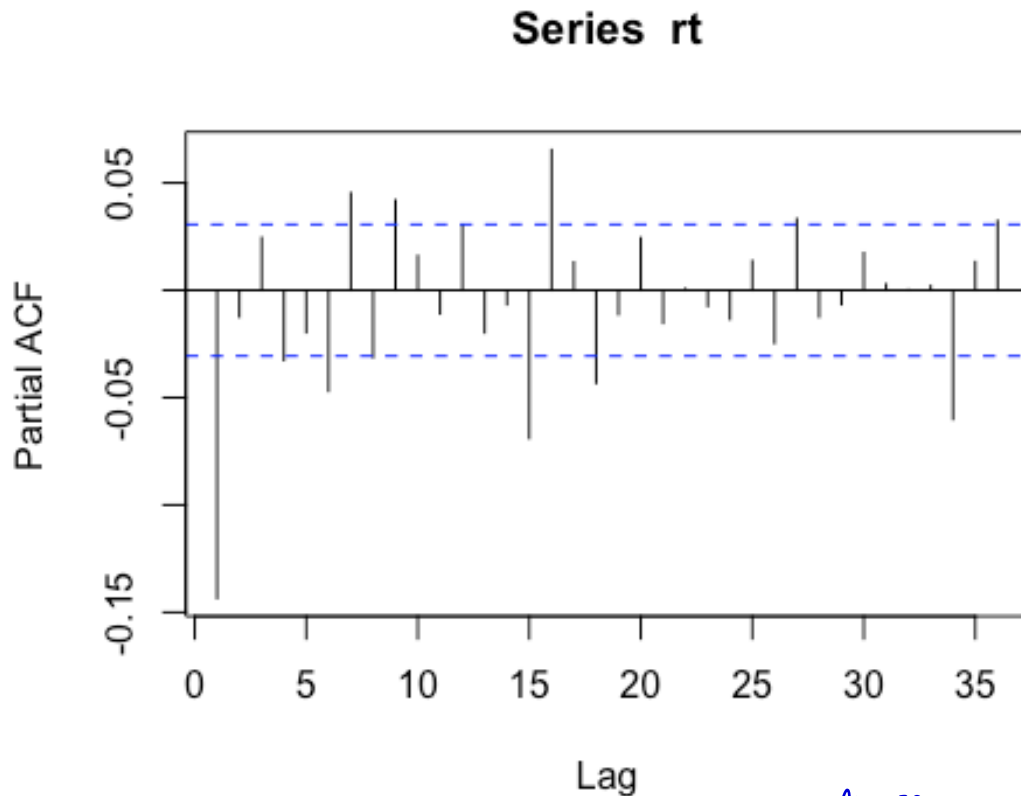
##
## One Sample t-test
##
## data:  rt
## t = 1.4961, df = 4086, p-value = 0.1347
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -9.051939e-05  6.737464e-04
## sample estimates:
## mean of x
## 0.0002916135

acf(rt)
```

Series rt



pacf(rt)



```
Box.test(rt,lag=10,type='Ljung')  
##  
## Box-Ljung test  
##  
## data: rt  
## X-squared = 131.85, df = 10, p-value < 2.2e-16
```

#3b

```
auto.arima(rt)  
## Series: rt  
## ARIMA(1,0,0) with non-zero mean  
##  
## Coefficients:  
##          ar1    mean  
##      -0.1435  3e-04  
## s.e.   0.0155  2e-04  
##  
## sigma^2 estimated as 0.0001521: log likelihood=12166.05  
## AIC=-24326.09 AICc=-24326.09 BIC=-24307.15
```

Ans. 3a

T-test

$$H_0: \mu = 0$$

$$H_1: \mu \neq 0$$

The computed p-value is higher than 0.05.
Then, we cannot reject H_0 at 0.05 level of significant. $\mu = 0$ with 95% CI

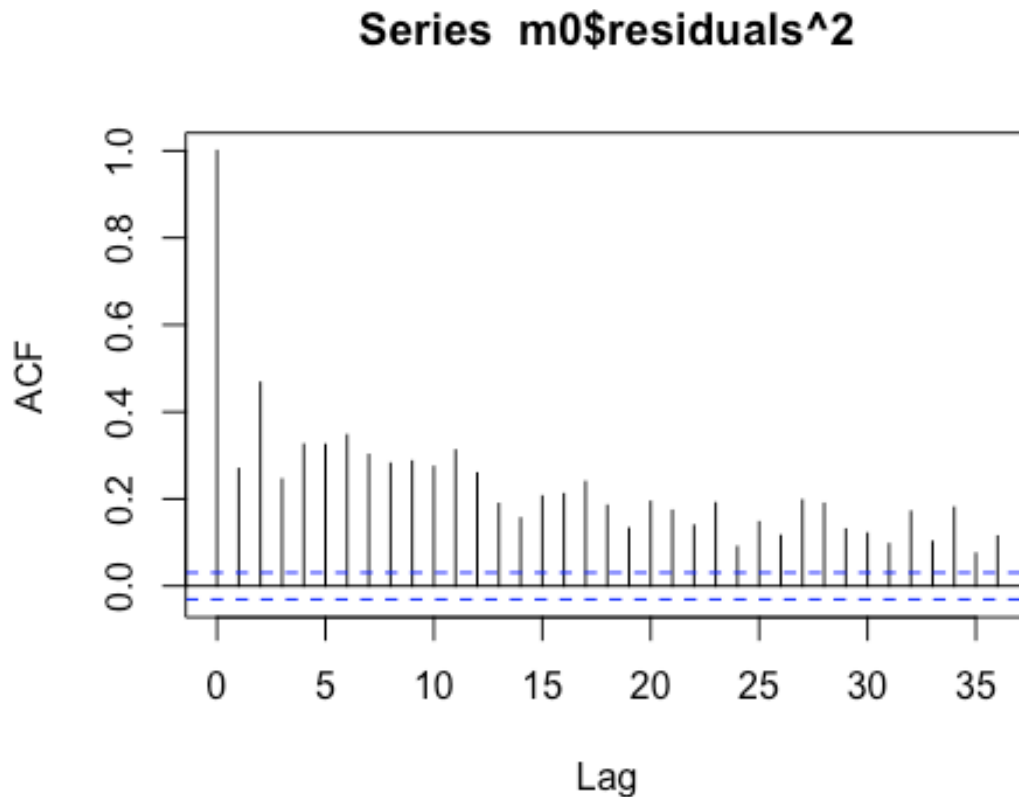
Ljung-Box test

$$H_0: e_1 = e_2 = e_3 = \dots = e_{10} = 0$$

$$H_1: \exists e_i \neq 0; i = 1, 2, 3, \dots, 10$$

There is serial correlation in log return (rt) with 95% CI, because computed p-value from Ljung-Box test is lower than 0.05. Then, we reject H_0 at 0.05 level of significant.

```
m0=auto.arima(rt)
acf(m0$residuals^2)
```



```
Box.test(m0$residuals^2,lag=10,type='Ljung')

##
## Box-Ljung test
##
## data: m0$residuals^2
## X-squared = 4132.1, df = 10, p-value < 2.2e-16

m1=garchFit(formula = ~arma(1,0)+garch(1,1),data = rt,trace=F)

## Warning: Using formula(x) is deprecated when x is a character vector of
length > 1.
## Consider formula(paste(x, collapse = " ")) instead.

summary(m1)

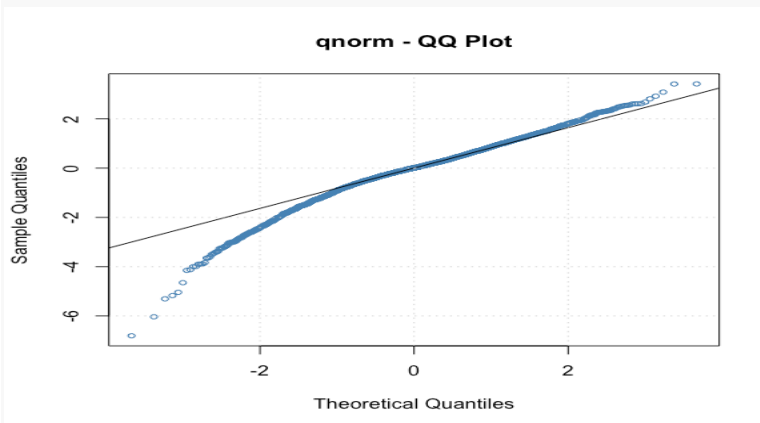
##
## Title:
## GARCH Modelling
##
## Call:
```

```

## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = rt, trace = F)
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + garch(1, 1)
## <environment: 0x7fc0d2e259f8>
## [data = rt]
##
## Conditional Distribution:
## norm
##
## Coefficient(s):
##          mu          ar1          omega          alpha1          beta1
## 7.4132e-04 -7.5408e-02  2.7169e-06  1.4182e-01  8.3740e-01
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate  Std. Error  t value Pr(>|t|)
## mu      7.413e-04  1.178e-04   6.292 3.14e-10 ***
## ar1    -7.541e-02  1.730e-02  -4.358 1.31e-05 ***
## omega   2.717e-06  3.504e-07   7.753 8.88e-15 ***
## alpha1  1.418e-01  1.202e-02  11.800 < 2e-16 ***
## beta1   8.374e-01  1.196e-02  70.032 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 13428.16    normalized:  3.285579
##
## Description:
## Thu Apr 29 00:29:01 2021 by user:
##
## Standardised Residuals Tests:
##                                     Statistic p-Value
## Jarque-Bera Test    R    Chi^2  1128.594  0
## Shapiro-Wilk Test   R    W      0.9719221  0
## Ljung-Box Test      R    Q(10)  17.65633  0.061045
## Ljung-Box Test      R    Q(15)  25.9899   0.03812896
## Ljung-Box Test      R    Q(20)  31.68968  0.04671932
## Ljung-Box Test      R^2  Q(10)  15.9892   0.09994212
## Ljung-Box Test      R^2  Q(15)  18.25272  0.2496126
## Ljung-Box Test      R^2  Q(20)  19.62331  0.4817041
## LM Arch Test        R    TR^2   16.7807   0.1580346
##
## Information Criterion Statistics:
##          AIC          BIC          SIC          HQIC
## -6.568712 -6.560985 -6.568715 -6.565976

```

```
#plot(m1)
```



3b.

$$\text{Mean equation: } \hat{r}_t = 7.413e-04 + (-7.541e-02) r_{t-1}$$

$(1.173e-04) \quad (1.730e-02)$

$$\text{variance equation: } \hat{\sigma}_t^2 = 2.717e-06 + (1.418e-01) a_{t-1}^2 + (9.374e-01) \hat{\sigma}_{t-1}^2$$

$(3.504e-07) \quad (1.202e-02) \quad (1.19e-02)$

ARMA(1,0)+GARCH(1,1) is the most appropriate because it provide the lowest AIC, BIC.
The model is adequate except for the mean equation with 15 and 20 lags.

#3c

```
m2=garchFit(formula = ~arma(1,0)+garch(1,1),data = rt,cond.dist =  
"std",trace=F)
```

```
## Warning: Using formula(x) is deprecated when x is a character vector of  
length > 1.
```

```
## Consider formula(paste(x, collapse = " ")) instead.
```

```
summary(m2)
```

```
##
```

```
## Title:
```

```
## GARCH Modelling
```

```
##
```

```
## Call:
```

```
## garchFit(formula = ~arma(1, 0) + garch(1, 1), data = rt, cond.dist =  
"std",
```

```
## trace = F)
```

```
##
```

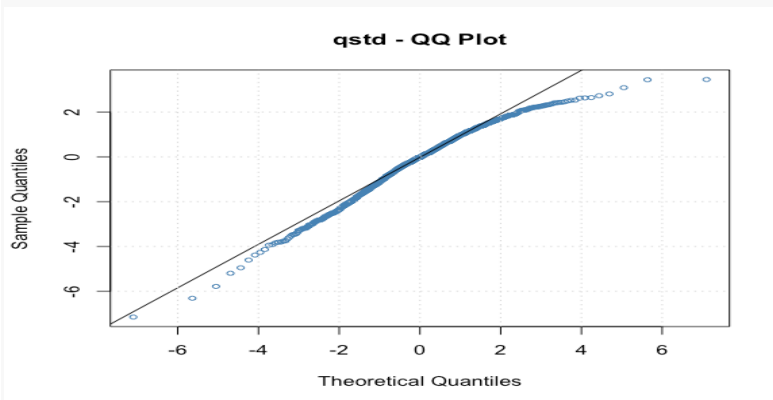
```
## Mean and Variance Equation:
```

```

## data ~ arma(1, 0) + garch(1, 1)
## <environment: 0x7fc0d7826708>
## [data = rt]
##
## Conditional Distribution:
## std
##
## Coefficient(s):
##          mu          ar1          omega          alpha1          beta1
shape
## 9.1359e-04 -7.0189e-02 1.6949e-06 1.4129e-01 8.5658e-01
5.0787e+00
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      9.136e-04 1.061e-04 8.612 < 2e-16 ***
## ar1     -7.019e-02 1.570e-02 -4.469 7.85e-06 ***
## omega   1.695e-06 3.625e-07 4.676 2.93e-06 ***
## alpha1  1.413e-01 1.460e-02 9.675 < 2e-16 ***
## beta1   8.566e-01 1.286e-02 66.618 < 2e-16 ***
## shape   5.079e+00 4.283e-01 11.857 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 13563.45 normalized: 3.318682
##
## Description:
## Thu Apr 29 00:29:02 2021 by user:
##
## Standardised Residuals Tests:
##
##          Statistic p-Value
## Jarque-Bera Test R Chi^2 1426.546 0
## Shapiro-Wilk Test R W 0.9697692 0
## Ljung-Box Test R Q(10) 17.75546 0.05923155
## Ljung-Box Test R Q(15) 26.09727 0.03701267
## Ljung-Box Test R Q(20) 31.5407 0.04844502
## Ljung-Box Test R^2 Q(10) 13.10432 0.2178975
## Ljung-Box Test R^2 Q(15) 17.76262 0.275351
## Ljung-Box Test R^2 Q(20) 20.49609 0.4273063
## LM Arch Test R TR^2 14.77776 0.2538157
##
## Information Criterion Statistics:
##          AIC          BIC          SIC          HQIC
## -6.634429 -6.625157 -6.634433 -6.631146

```

#plot(m2)



Fitted model

$$\text{Mean equation: } \hat{r}_t = 9.176e-04 + (-7.019e-02) r_{t-1}$$

$(1.061e-04) \quad (1.570e-02)$

$$\text{variance equation: } \hat{\sigma}_t^2 = 1.695e-06 + (1.413e-01) \hat{\sigma}_{t-1}^2 + (8.566e-01) \delta_{t-1}^2$$

$(3.615e-07) \quad (1.460e-02) \quad (1.286e-02)$

#3d

predict(m2,5)

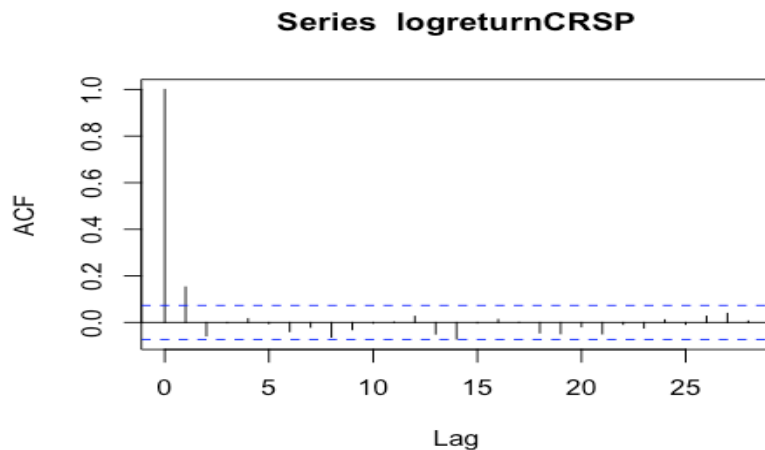
```
## meanForecast meanError standardDeviation
## 1 0.0011355887 0.009054724 0.009054724
## 2 0.0008338888 0.009160372 0.009138299
## 3 0.0008550647 0.009243329 0.009220941
## 4 0.0008535784 0.009325271 0.009302675
## 5 0.0008536827 0.009406326 0.009383526
```

#Question4

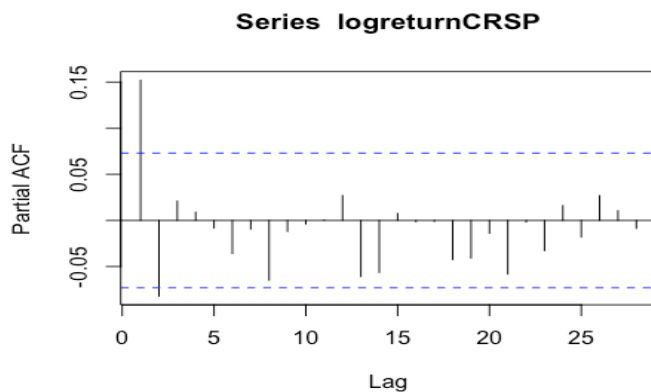
```
CRSP=read.table("m-deciles.txt",header = T)
simplereturnCRSP=CRSP[,10]
logreturnCRSP=log(simplereturnCRSP+1)
#4a
t.test(logreturnCRSP)

##
## One Sample t-test
##
## data: logreturnCRSP
## t = 5.1808, df = 719, p-value = 2.873e-07
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
##  0.005946562 0.013203545
## sample estimates:
## mean of x
## 0.009575054

acf(logreturnCRSP)
```



```
pacf(logreturnCRSP)
```



```

Box.test(logreturnCRSP,lag=10,type='Ljung')

##
## Box-Ljung test
##
## data: logreturnCRSP
## X-squared = 24.257, df = 10, p-value = 0.006946

auto.arima(logreturnCRSP)

## Series: logreturnCRSP
## ARIMA(0,0,2) with non-zero mean
##
## Coefficients:
##          ma1          ma2          mean
##          0.1660 -0.0558  0.0096
## s.e.      0.0373  0.0370  0.0020
##
## sigma^2 estimated as 0.002392: log likelihood=1152.66
## AIC=-2297.32 AICc=-2297.26 BIC=-2279

m1<-arima(logreturnCRSP,order=c(0,0,2))
m1

##
## Call:
## arima(x = logreturnCRSP, order = c(0, 0, 2))
##
## Coefficients:
##          ma1          ma2  intercept
##          0.1660 -0.0558   0.0096
## s.e.      0.0373  0.0370   0.0020
##
## sigma^2 estimated as 0.002382: log likelihood = 1152.66, aic = -2297.32

```

```
Box.test(m1$residuals,lag=10,type='Ljung')
```

```

##
## Box-Ljung test
##
## data: m1$residuals
## X-squared = 4.3881, df = 10, p-value = 0.9281

```

$\hat{\rho}_k = 0.0096 + 0.166a_{k-1} - 0.0558a_{k-2}$
 Checking the serial correlation in residual form (a_t)
 Ljung-Box test: $H_0: a_1 + a_2 + \dots + a_{10} = 0$
 $H_1: \exists a_i \neq 0$
 The computed p-value = 0.92 > 0.05. Thus, we can not reject H_0 at 0.05 level of significance. This model is adequate (no serial correlation in a_t)

Ans. 90

T-test

$H_0: \mu = 0$
 $H_1: \mu \neq 0$

The computed p-value is higher than 0.05.
 Thus, we can not reject H_0 at 0.05 level of significance. $\mu = 0$ with 95% CI. Thus, expected value of log return of CRSP is equal to 0.

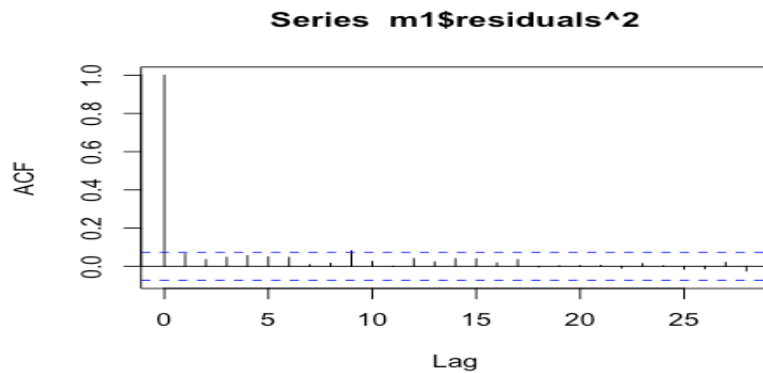
Ljung-Box test

$H_0: \rho_1 = \rho_2 = \rho_3 = \dots = \rho_{10} = 0$
 $H_1: \exists \rho_i \neq 0; i = 1, 2, 3, \dots, 10$

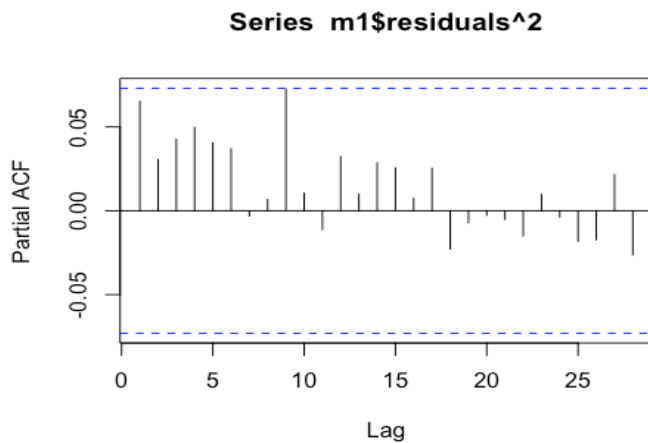
There is serial correlation in log return of CRSP with 95% CI, because computed p-value from Ljung-Box test is lower than 0.05. Thus, we reject H_0 at 0.05 level of significance.

#4b

```
acf(m1$residuals^2)
```



```
pacf(m1$residuals^2)
```



```
Box.test(m1$residuals^2, lag = 10, type = 'Ljung')
```

```
##  
## Box-Ljung test  
##  
## data: m1$residuals^2  
## X-squared = 16.677, df = 10, p-value = 0.08183
```

4b. Test ARCH effect (log return of CRSP)

Ljung-Box test

$H_0: \rho_1 = \rho_2 = \rho_3 = \dots = \rho_{10} = 0$ / No ARCH effect

$H_1: \exists \rho_i \neq 0 ; i = 1, 2, 3, \dots, 10$ / there exist ARCH effect

\therefore There is no ARCH effect in log return of CRSP with 95% CI.
Because computed p-value from Ljung-Box test is higher than 0.05. Thus, we cannot reject H_0 at 0.05 level of significance.


```

## Jarque-Bera Test R Chi^2 647.6357 0
## Shapiro-Wilk Test R W 0.962804 1.502752e-12
## Ljung-Box Test R Q(10) 8.582995 0.572082
## Ljung-Box Test R Q(15) 12.95811 0.6055337
## Ljung-Box Test R Q(20) 15.77362 0.7305655
## Ljung-Box Test R^2 Q(10) 8.153476 0.6138486
## Ljung-Box Test R^2 Q(15) 12.29206 0.6568012
## Ljung-Box Test R^2 Q(20) 13.57787 0.851238
## LM Arch Test R TR^2 8.24593 0.7656286
##
## Information Criterion Statistics:
## AIC BIC SIC HQIC
## -3.207301 -3.181861 -3.207363 -3.197480

```

Mean & variance equation is adequate but the distribution is not normal.

Fitted equation

$$\text{Mean: } \hat{\mu}_t = 0.01053 + 0.197 \mu_{t-1}$$

(0.00194) (0.082)

$$\text{Variance: } \hat{\sigma}_t^2 = 0.002 + 0.1815 \sigma_{t-1}^2$$

(0.0001) (0.015)

#4d

```
m3=garchFit(~arma(1,0)+garch(1,0),data=logreturnCRSP,cond.dist="std",trace=F)
```

```
## Warning: Using formula(x) is deprecated when x is a character vector of
length > 1.
```

```
## Consider formula(paste(x, collapse = " ")) instead.
```

```
summary(m3)
```

```
##
```

```
## Title:
```

```
## GARCH Modelling
```

```

##
## Call:
## garchFit(formula = ~arma(1, 0) + garch(1, 0), data = logreturnCRSP,
##   cond.dist = "std", trace = F)
##
## Mean and Variance Equation:
## data ~ arma(1, 0) + garch(1, 0)
## <environment: 0x7fc0d8e3b8b0>
## [data = logreturnCRSP]
##
## Conditional Distribution:
## std
##
## Coefficient(s):
##      mu      ar1      omega      alpha1      shape
## 0.0116189 0.1076982 0.0019203 0.1900830 6.4225253
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      0.0116189 0.0017710 6.561 5.35e-11 ***
## ar1     0.1076982 0.0403169 2.671 0.00756 **
## omega  0.0019203 0.0001818 10.564 < 2e-16 ***
## alpha1 0.1900830 0.0713692 2.663 0.00774 **
## shape  6.4225253 1.3115897 4.897 9.74e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 1187.516 normalized: 1.649328
##
## Description:
## Thu Apr 29 00:29:02 2021 by user:
##
##
## Standardised Residuals Tests:
##
##      Statistic p-Value
## Jarque-Bera Test R Chi^2 680.4834 0
## Shapiro-Wilk Test R W 0.9612945 7.49198e-13
## Ljung-Box Test R Q(10) 9.486455 0.4866411
## Ljung-Box Test R Q(15) 13.8214 0.5391158
## Ljung-Box Test R Q(20) 17.0087 0.6524086
## Ljung-Box Test R^2 Q(10) 7.567444 0.671006
## Ljung-Box Test R^2 Q(15) 11.42176 0.7221637
## Ljung-Box Test R^2 Q(20) 12.79422 0.8860373
## LM Arch Test R TR^2 7.723697 0.8063327
##
## Information Criterion Statistics:

```



```

##      mu      omega      alpha1
## 0.012346 0.002016 0.194126
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate  Std. Error  t value Pr(>|t|)
## mu      0.012346    0.001923    6.421 1.35e-10 ***
## omega   0.002016    0.000145   13.900 < 2e-16 ***
## alpha1  0.194126    0.062209    3.121 0.00181 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 1152.289    normalized: 1.600401
##
## Description:
## Thu Apr 29 00:29:02 2021 by user:
##
##
## Standardised Residuals Tests:
##
##      Jarque-Bera Test  R      Chi^2  746.8024  0
##      Shapiro-Wilk Test  R      W      0.9570702 1.171061e-13
##      Ljung-Box Test    R      Q(10)  18.72223  0.04393591
##      Ljung-Box Test    R      Q(15)  23.27998  0.07837365
##      Ljung-Box Test    R      Q(20)  27.61004  0.1189566
##      Ljung-Box Test    R^2    Q(10)  7.012055  0.7243064
##      Ljung-Box Test    R^2    Q(15)  10.3604   0.7964784
##      Ljung-Box Test    R^2    Q(20)  11.97825  0.9168225
##      LM Arch Test      R      TR^2   7.047101  0.8544853
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -3.192468 -3.173388 -3.192503 -3.185102

```

4e.

∴ Variance equation is adequate while mean equation is not fully adequate, and not normally distributed.

Fitted equation Mean: $\hat{\mu}_t = 0.01235$
(0.0019)

Variance: $\hat{\sigma}_t^2 = 0.002016 + 0.194126$
(0.000145) (0.0622)

```

#4f
m5=garchFit(formula = ~garch(1,0),data = logreturnCRSP,cond.dist =
"std",trace=F)

## Warning: Using formula(x) is deprecated when x is a character vector of
length > 1.
## Consider formula(paste(x, collapse = " ")) instead.

summary(m5)

##
## Title:
## GARCH Modelling
##
## Call:
## garchFit(formula = ~garch(1, 0), data = logreturnCRSP, cond.dist = "std",
## trace = F)
##
## Mean and Variance Equation:
## data ~ garch(1, 0)
## <environment: 0x7fc0d71e5d68>
## [data = logreturnCRSP]
##
## Conditional Distribution:
## std
##
## Coefficient(s):
##      mu      omega      alpha1      shape
## 0.013356 0.001928 0.204163 6.220222
##
## Std. Errors:
## based on Hessian
##
## Error Analysis:
##      Estimate Std. Error t value Pr(>|t|)
## mu      0.013356 0.001685 7.927 2.22e-15 ***
## omega  0.001928 0.000185 10.421 < 2e-16 ***
## alpha1 0.204163 0.072375 2.821 0.00479 **
## shape  6.220223 1.236608 5.030 4.90e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Log Likelihood:
## 1183.349 normalized: 1.64354
##
## Description:
## Thu Apr 29 00:29:02 2021 by user:
##
##
## Standardised Residuals Tests:

```

```

##                               Statistic p-Value
## Jarque-Bera Test      R      Chi^2    746.3878    0
## Shapiro-Wilk Test     R      W         0.9574271  1.363165e-13
## Ljung-Box Test       R      Q(10)    18.51406   0.04688705
## Ljung-Box Test       R      Q(15)    22.99926   0.08415548
## Ljung-Box Test       R      Q(20)    27.3947    0.1245202
## Ljung-Box Test       R^2    Q(10)    6.667871   0.7563837
## Ljung-Box Test       R^2    Q(15)    10.0157    0.8187514
## Ljung-Box Test       R^2    Q(20)    11.63085   0.928196
## LM Arch Test         R      TR^2    6.819315   0.8693193
##
## Information Criterion Statistics:
##      AIC      BIC      SIC      HQIC
## -3.275969 -3.250529 -3.276031 -3.266148

```

#4g

4f. Variance equation is adequate while mean equation is not fully adequate.

Fitted eq

$$\text{Mean: } \hat{\mu}_t = 0.013356$$

(0.0016881)

$$\text{variance: } \hat{\sigma}_t^2 = 0.001918 + 0.209167 a_{t-1}^2$$

(0.00185) (0.0720)

4g The most appropriate model is m3: AR(1) - ARCH(1) with t-distribution because

1. adequate mean & variance equation
2. lowest AIC & BIC
3. All coefficient estimate are significant.