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QUESTION 1

#1a. IMPORT DATA INTO R#

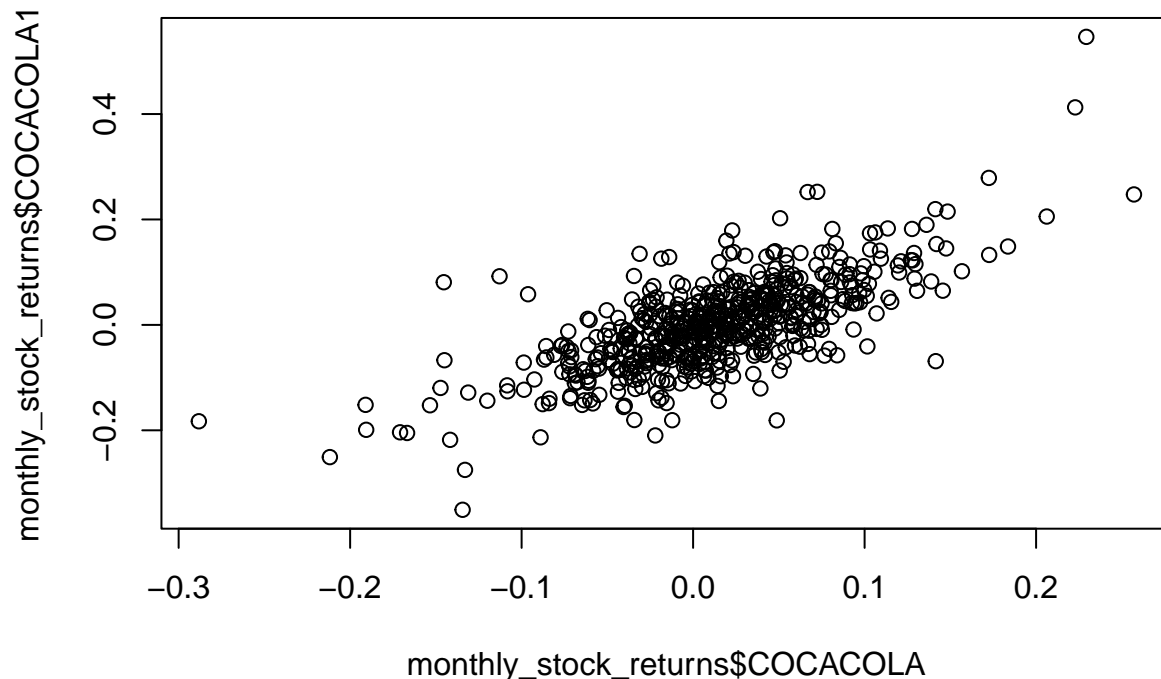
```
library(readxl)
monthly_stock_returns <- read_excel("Documents/4328 - Applied Financial Econometrics/Assignment/Assignment1.xlsx")
```

#1b. TRANSFORM THE SIMPLE RETURNS INTO LOG RETURNS#

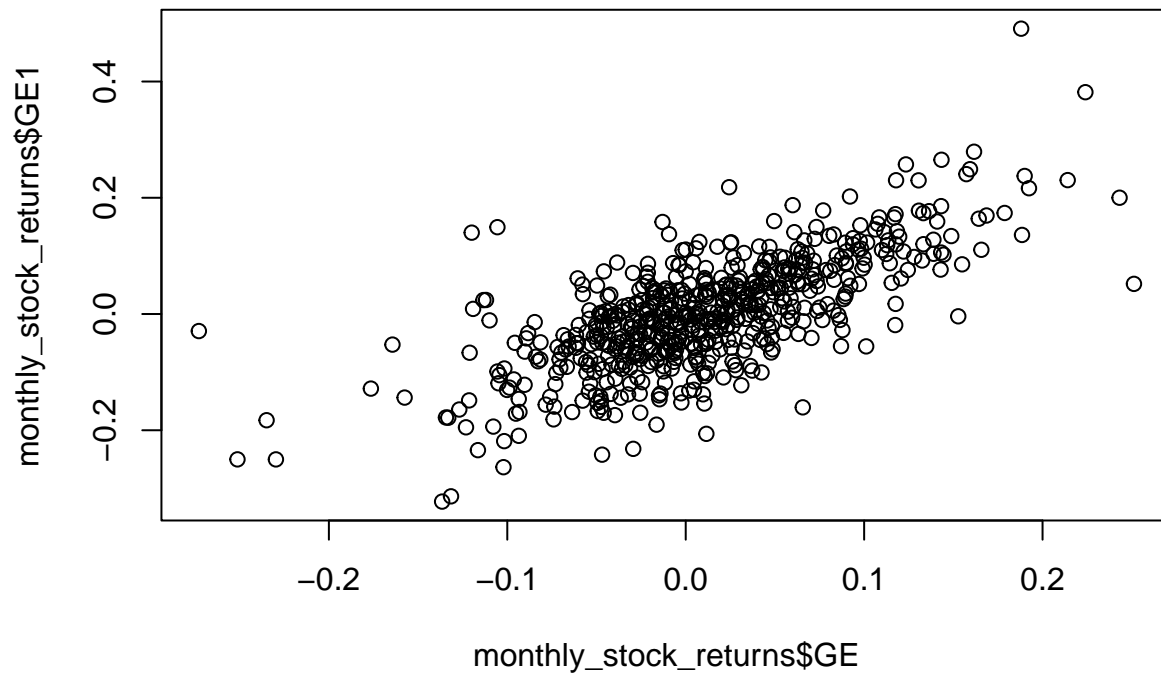
```
monthly_stock_returns$COCACOLA1 <- c(0, diff(log(1+monthly_stock_returns$COCACOLA), lag = 1))
monthly_stock_returns$GE1 <- c(0, diff(log(1+monthly_stock_returns$GE), lag = 1))
monthly_stock_returns$IBM1 <- c(0, diff(log(1+monthly_stock_returns$IBM), lag = 1))
monthly_stock_returns$VWRET1 <- c(0, diff(log(1+monthly_stock_returns$VWRET), lag = 1))
monthly_stock_returns$EWRET1 <- c(0, diff(log(1+monthly_stock_returns$EWRET), lag = 1))
```

#1c. PLOT THE SIMPLE AND LOG RETURNS FOR ALL THE SERIES#

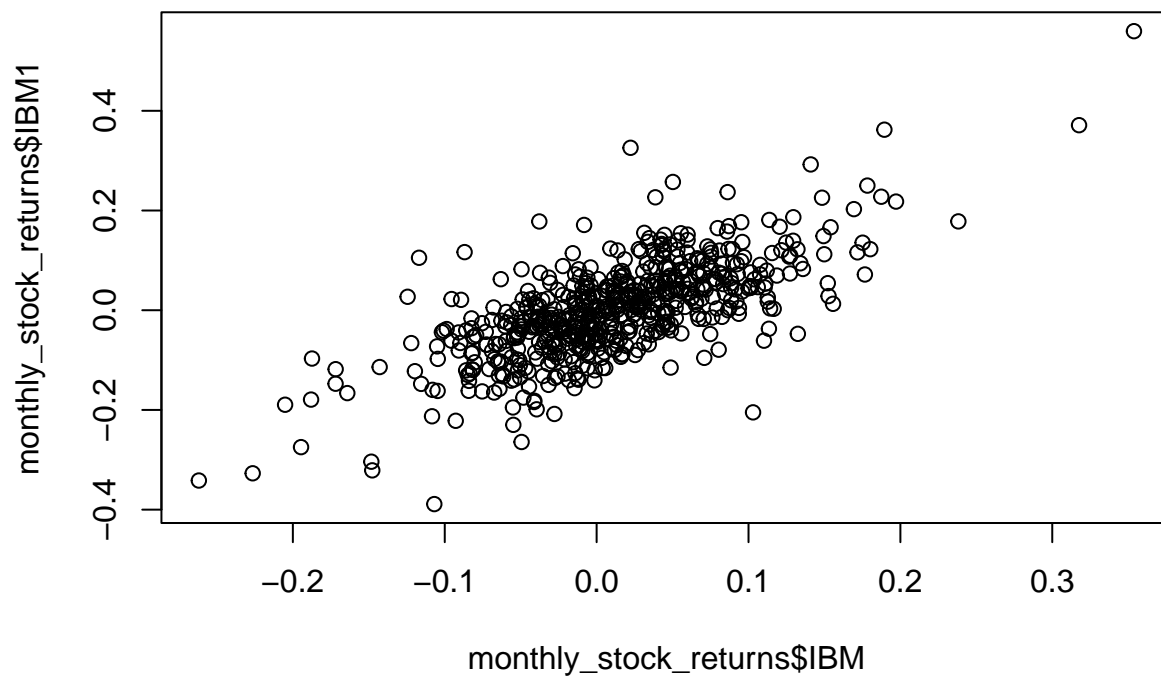
```
plot(monthly_stock_returns$COCACOLA, monthly_stock_returns$COCACOLA1)
```



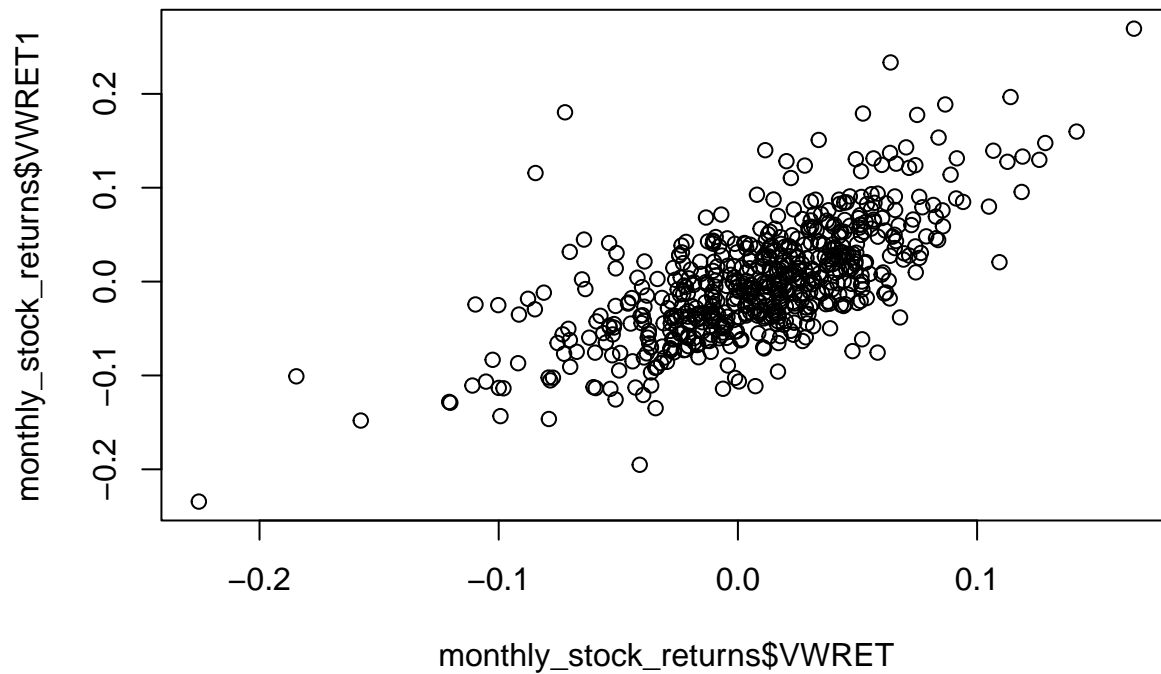
```
plot(monthly_stock_returns$GE, monthly_stock_returns$GE1)
```



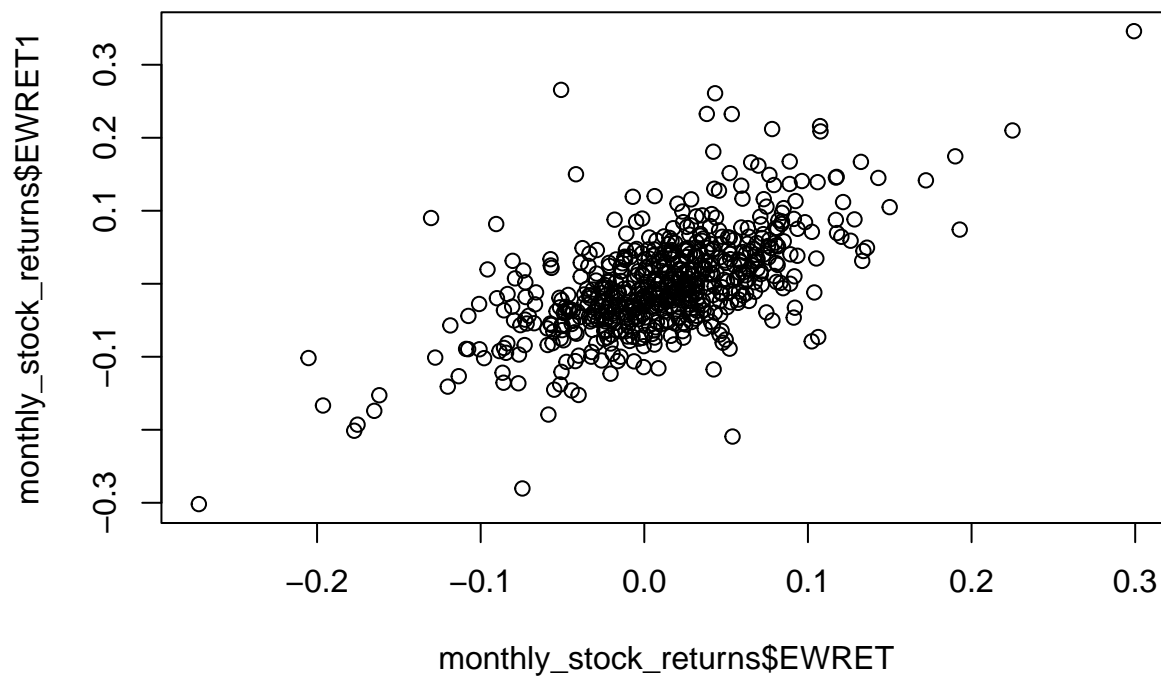
```
plot(monthly_stock_returns$IBM, monthly_stock_returns$IBM1)
```



```
plot(monthly_stock_returns$VWRET, monthly_stock_returns$VWRET1)
```

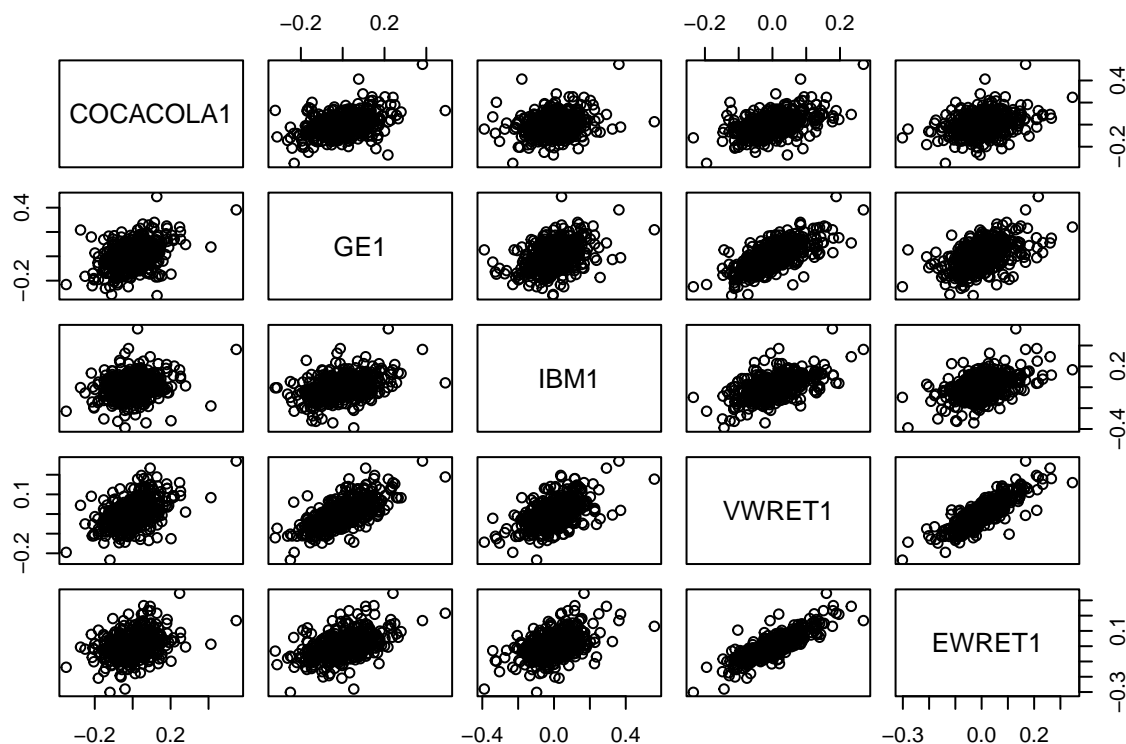


```
plot(monthly_stock_returns$EWRET, monthly_stock_returns$EWRET1)
```



#1d. SCATTERPLOT OF ALL THE LOG RETURNS SERIES#

```
monthly_stock_returns_log <- monthly_stock_returns[,c(7:11)]
plot(monthly_stock_returns_log)
```



#1e. SAMPLE MEAN, VARIANCE, SKEWNESS, EXCESS KURTOSIS, MINIMUM AND MAXIMUM OF LOG RETURNS#

```
library(moments)
COACOLA_log <- c(mean(monthly_stock_returns$COACOLA1), var(monthly_stock_returns$COACOLA1), skewness(monthly_stock_returns$COACOLA1), excess_kurtosis(monthly_stock_returns$COACOLA1), min(monthly_stock_returns$COACOLA1), max(monthly_stock_returns$COACOLA1))
GE_log <- c(mean(monthly_stock_returns$GE1), var(monthly_stock_returns$GE1), skewness(monthly_stock_returns$GE1), excess_kurtosis(monthly_stock_returns$GE1), min(monthly_stock_returns$GE1), max(monthly_stock_returns$GE1))
IBM_log <- c(mean(monthly_stock_returns$IBM1), var(monthly_stock_returns$IBM1), skewness(monthly_stock_returns$IBM1), excess_kurtosis(monthly_stock_returns$IBM1), min(monthly_stock_returns$IBM1), max(monthly_stock_returns$IBM1))
VWRET_log <- c(mean(monthly_stock_returns$VWRET1), var(monthly_stock_returns$VWRET1), skewness(monthly_stock_returns$VWRET1), excess_kurtosis(monthly_stock_returns$VWRET1), min(monthly_stock_returns$VWRET1), max(monthly_stock_returns$VWRET1))
EWRET_log <- c(mean(monthly_stock_returns$EWRET1), var(monthly_stock_returns$EWRET1), skewness(monthly_stock_returns$EWRET1), excess_kurtosis(monthly_stock_returns$EWRET1), min(monthly_stock_returns$EWRET1), max(monthly_stock_returns$EWRET1))

summary_df <- data.frame(COACOLA_log, GE_log, IBM_log, VWRET_log, EWRET_log)
row.names(summary_df) <- c("MEAN", "VARIANCE", "SKEWNESS", "EXCESS KURTOSIS", "MINIMUM", "MAXIMUM")
summary_df
```

	COACOLA_log	GE_log	IBM_log	VWRET_log	EWRET_log
## MEAN	-2.899334e-05	0.0001845287	-0.0001788354	-7.968444e-05	-0.0001513314
## VARIANCE	7.497612e-03	0.0094675802	0.0095485492	3.755511e-03	0.0050531098
## SKEWNESS	5.483142e-01	0.2465962382	0.2450713625	4.192380e-01	0.4110876520
## EXCESS KURTOSIS	6.766125e+00	4.4406753573	5.9861713945	4.352286e+00	5.5793209361
## MINIMUM	-3.507570e-01	-0.3227354940	-0.3888763005	-2.343679e-01	-0.3017915746
## MAXIMUM	5.463954e-01	0.4909426327	0.5596835625	2.694401e-01	0.3459855034

#1f. TEST OF STATISTICAL DIFFERENCE OF SAMPLE MEAN, SKEWNESS AND EXCESS KURTOSIS OF LOG RETURNS FROM ZERO AT 5% SIGNIFICANCE LEVEL #

SAMPLE MEAN TEST

```
mean_COCACOLA1 <- t.test(monthly_stock_returns_log$COCACOLA1)
mean_GE1 <- t.test(monthly_stock_returns_log$GE1)
mean_IBM1 <- t.test(monthly_stock_returns_log$IBM1)
mean_VWRET1 <- t.test(monthly_stock_returns_log$VWRET1)
mean_EWRET1 <- t.test(monthly_stock_returns_log$EWRET1)
```

```
mean_COCACOLA1
```

```
##
## One Sample t-test
##
## data: monthly_stock_returns_log$COCACOLA1
## t = -0.0081194, df = 587, p-value = 0.9935
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.007042211 0.006984224
## sample estimates:
## mean of x
## -2.899334e-05
```

```
mean_GE1
```

```
##
## One Sample t-test
##
## data: monthly_stock_returns_log$GE1
## t = 0.045987, df = 587, p-value = 0.9633
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.007696362 0.008065419
## sample estimates:
## mean of x
## 0.0001845287
```

```
mean_IBM1
```

```
##
## One Sample t-test
##
## data: monthly_stock_returns_log$IBM1
## t = -0.044379, df = 587, p-value = 0.9646
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.008093354 0.007735683
## sample estimates:
## mean of x
## -0.0001788354
```

```
mean_VWRET1
```

```
##
## One Sample t-test
##
## data: monthly_stock_returns_log$VWRET1
## t = -0.03153, df = 587, p-value = 0.9749
```

```
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.005043211 0.004883842
## sample estimates:
## mean of x
## -7.968444e-05
```

```
mean_EWRET1
```

```
##
## One Sample t-test
##
## data: monthly_stock_returns_log$EWRET1
## t = -0.051622, df = 587, p-value = 0.9588
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.005908848 0.005606185
## sample estimates:
## mean of x
## -0.0001513314
```

Answer: Since the p-value > 5%, at 5% significance level we do not reject the null hypothesis that the sample means of the log returns are statistically equal to zero. This means that the true means of the log returns of the variables (COCACOLA, IBM, GE, VWRET AND EWRET) are equal to zero.

SKEWNESS TEST

```
skew_COCACOLA1 <- skewness(monthly_stock_returns_log$COCACOLA1)/sqrt(6/588)
skew_GE1 <- skewness(monthly_stock_returns_log$GE1)/sqrt(6/588)
skew_IBM1 <- skewness(monthly_stock_returns_log$IBM1)/sqrt(6/588)
skew_VWRET1 <- skewness(monthly_stock_returns_log$VWRET1)/sqrt(6/588)
skew_EWRET1 <- skewness(monthly_stock_returns_log$EWRET1)/sqrt(6/588)
```

```
skew_COCACOLA1
```

```
## [1] 5.428033
```

```
skew_GE1
```

```
## [1] 2.441178
```

```
skew_IBM1
```

```
## [1] 2.426083
```

```
skew_VWRET1
```

```
## [1] 4.150244
```

```
skew_EWRET1
```

```
## [1] 4.06956
```

Answer: Since the absolute value from the result of the skewness test > 1.96, at 5% significance level we reject the null hypothesis that the skewness of the log returns are equal to zero. This means that the log returns of the variables (COCACOLA, IBM, GE, VWRET AND EWRET) do not follow normal distribution.

EXCESS KURTOSIS TEST

```
kurtosis_COCACOLA1 <- (kurtosis(monthly_stock_returns_log$COCACOLA1)-3)/sqrt(24/588)
kurtosis_GE1 <- (kurtosis(monthly_stock_returns_log$GE1)-3)/sqrt(24/588)
kurtosis_IBM1 <- (kurtosis(monthly_stock_returns_log$IBM1)-3)/sqrt(24/588)
kurtosis_VWRET1 <- (kurtosis(monthly_stock_returns_log$VWRET1)-3)/sqrt(24/588)
kurtosis_EWRET1 <- (kurtosis(monthly_stock_returns_log$EWRET1)-3)/sqrt(24/588)
```

```
kurtosis_COCACOLA1
```

```
## [1] 18.64137
```

```
kurtosis_GE1
```

```
## [1] 7.130979
```

```
kurtosis_IBM1
```

```
## [1] 14.78079
```

```
kurtosis_VWRET1
```

```
## [1] 6.693475
```

```
kurtosis_EWRET1
```

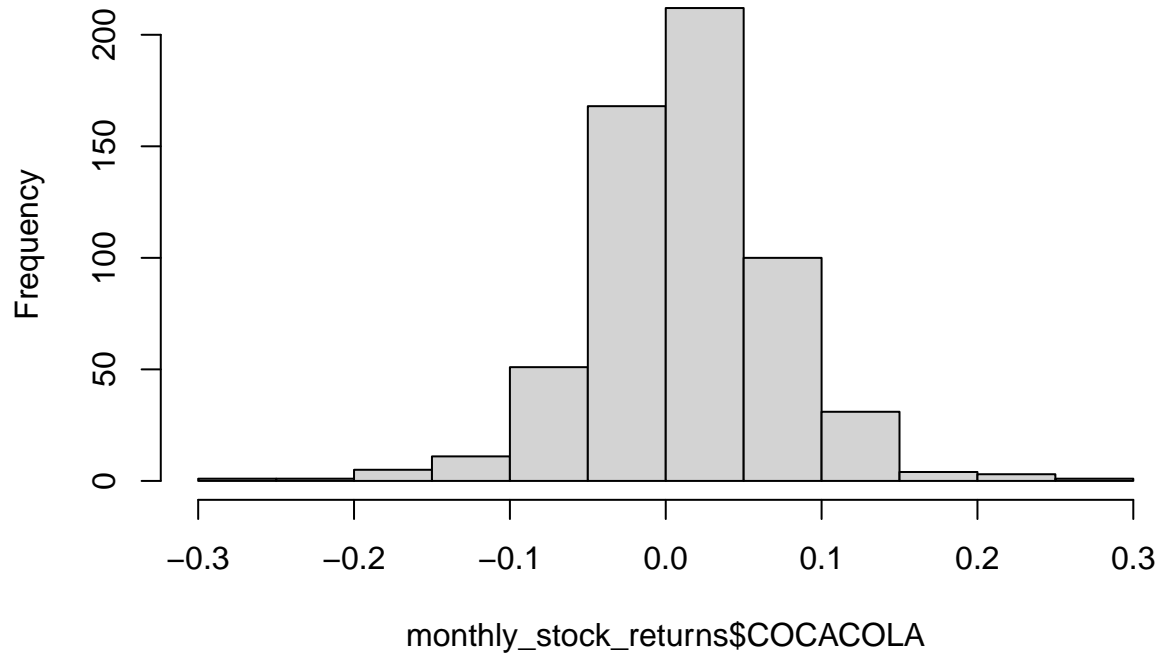
```
## [1] 12.76699
```

Answer: Since the value from the result of the kurtosis test > 1.96 , at 5% significance level we reject the null hypothesis that the kurtosis of the log returns are equal to zero. This means that the log returns of the variables (COCACOLA, IBM, GE, VWRET AND EWRET) do not follow normal distribution.

#1g. HISTOGRAM OF RETURNS IN COMPARISON TO NORMAL AND STUDENT DISTRIBUTIONS
#

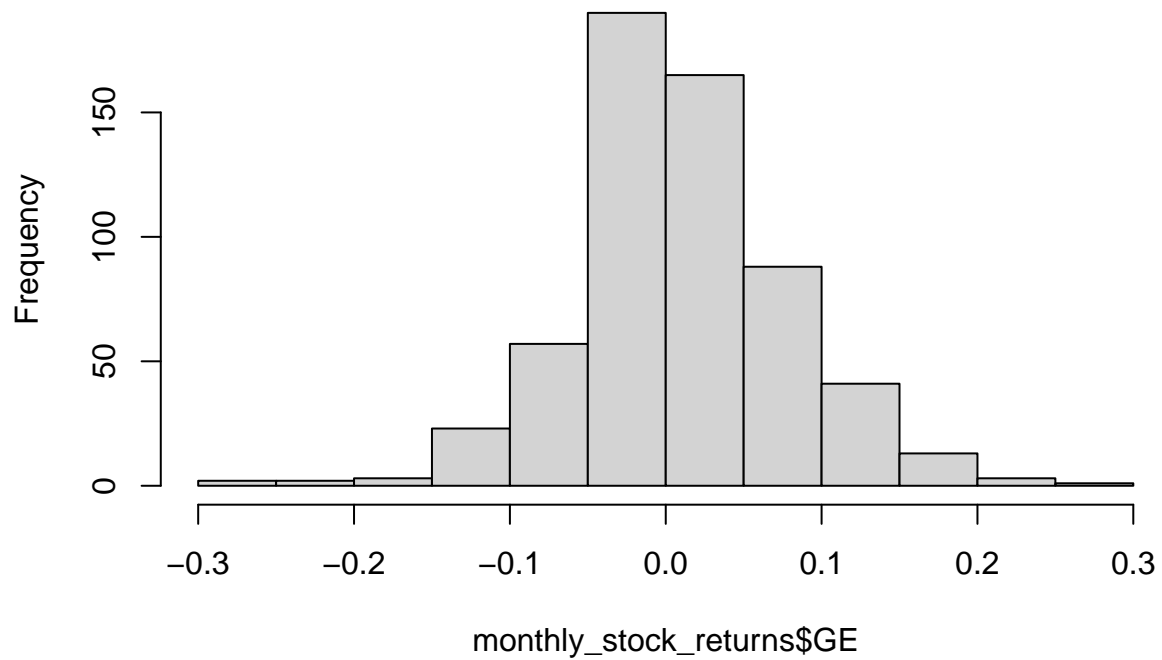
```
hist(monthly_stock_returns$COCACOLA)
```

Histogram of monthly_stock_returns\$COACOLA



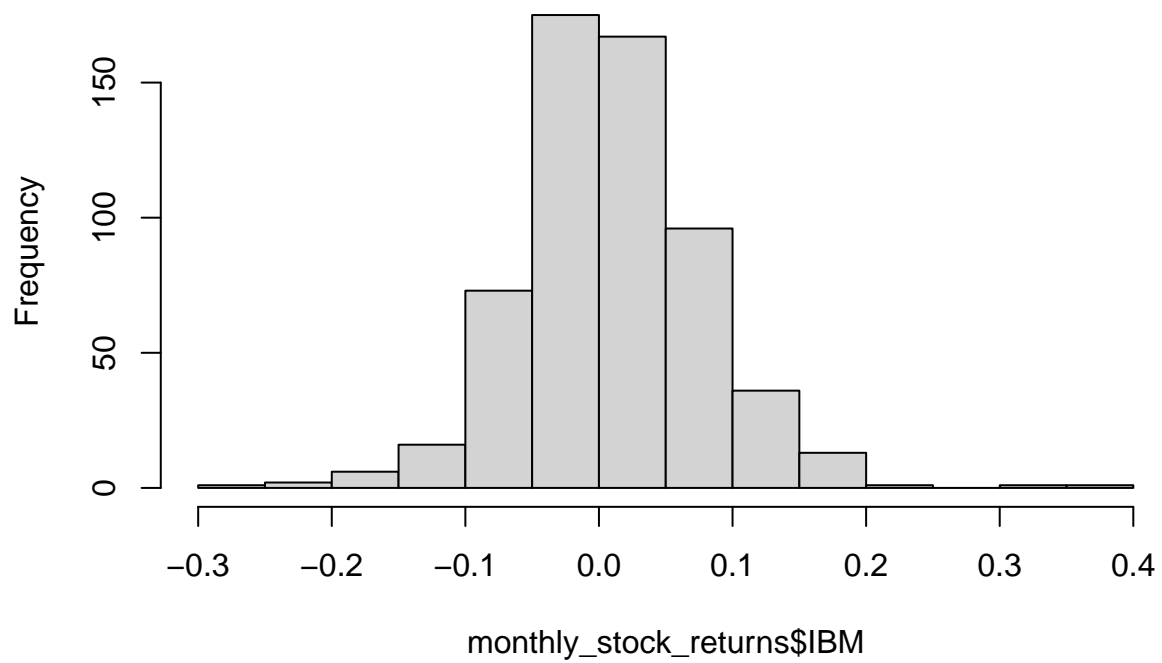
```
hist(monthly_stock_returns$GE)
```

Histogram of monthly_stock_returns\$GE



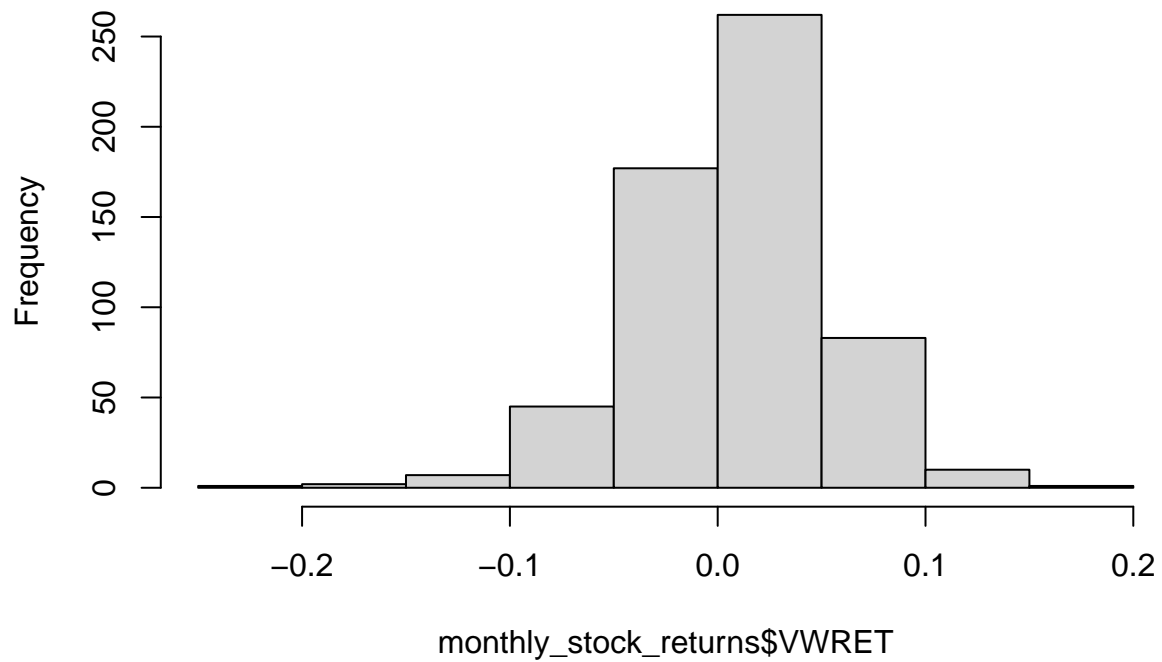
```
hist(monthly_stock_returns$IBM)
```


Histogram of monthly_stock_returns\$IBM



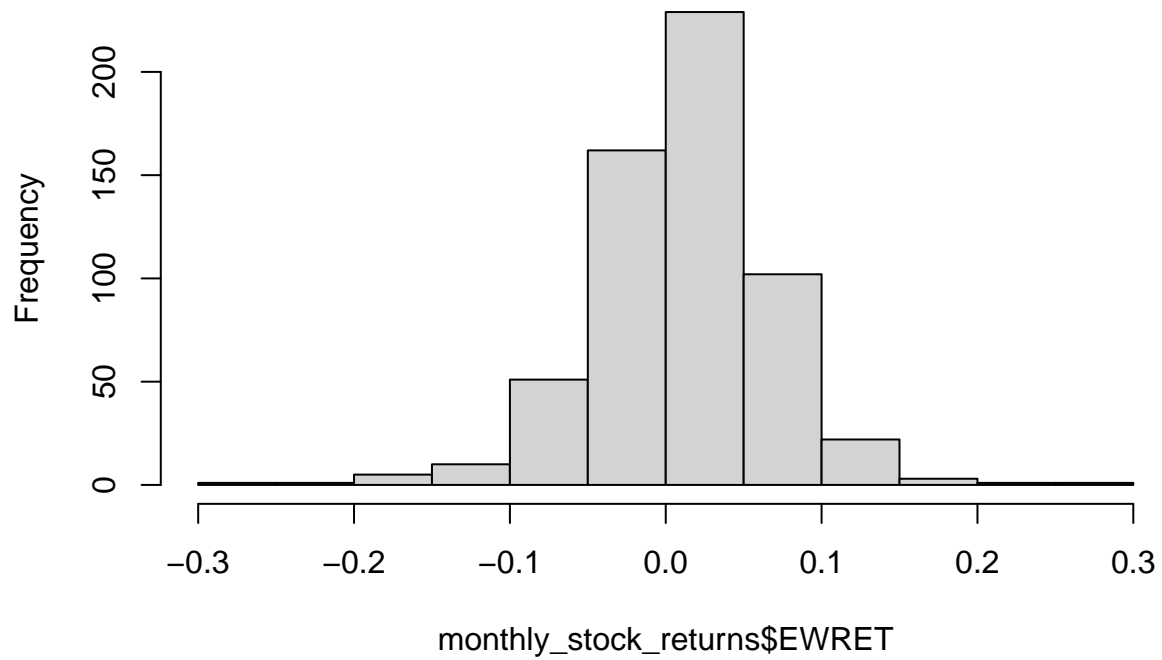
```
hist(monthly_stock_returns$VWRET)
```

Histogram of monthly_stock_returns\$VWRET



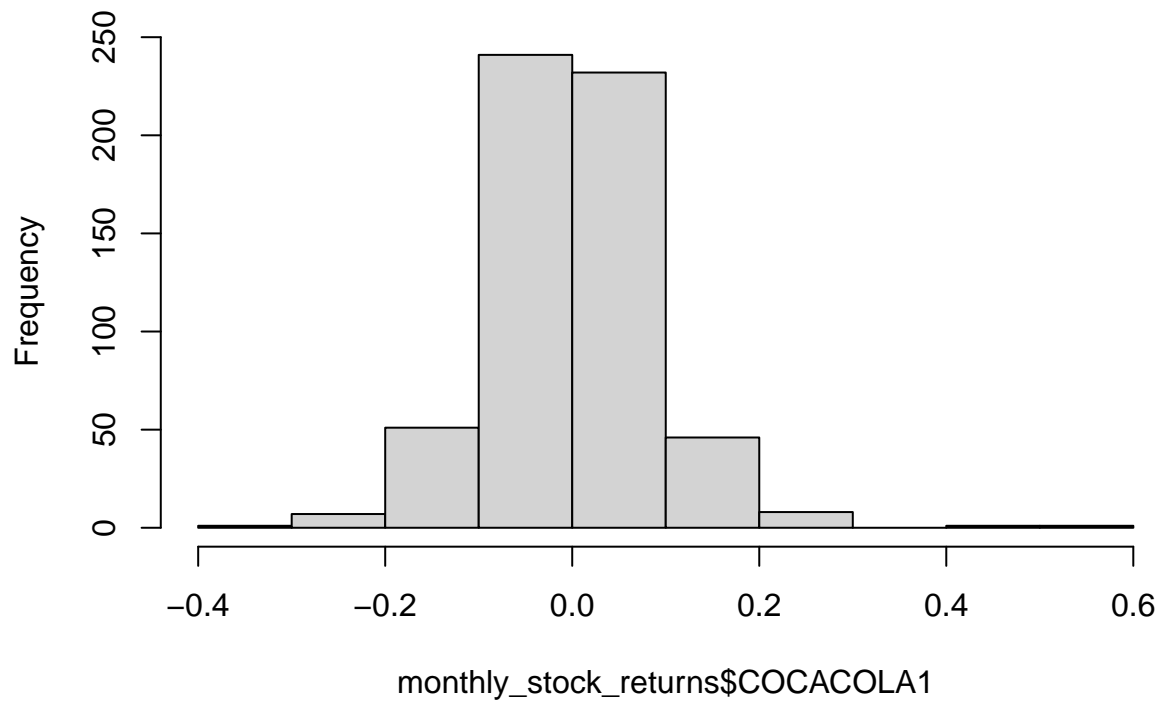
```
hist(monthly_stock_returns$EWRET)
```

Histogram of monthly_stock_returns\$EWRET



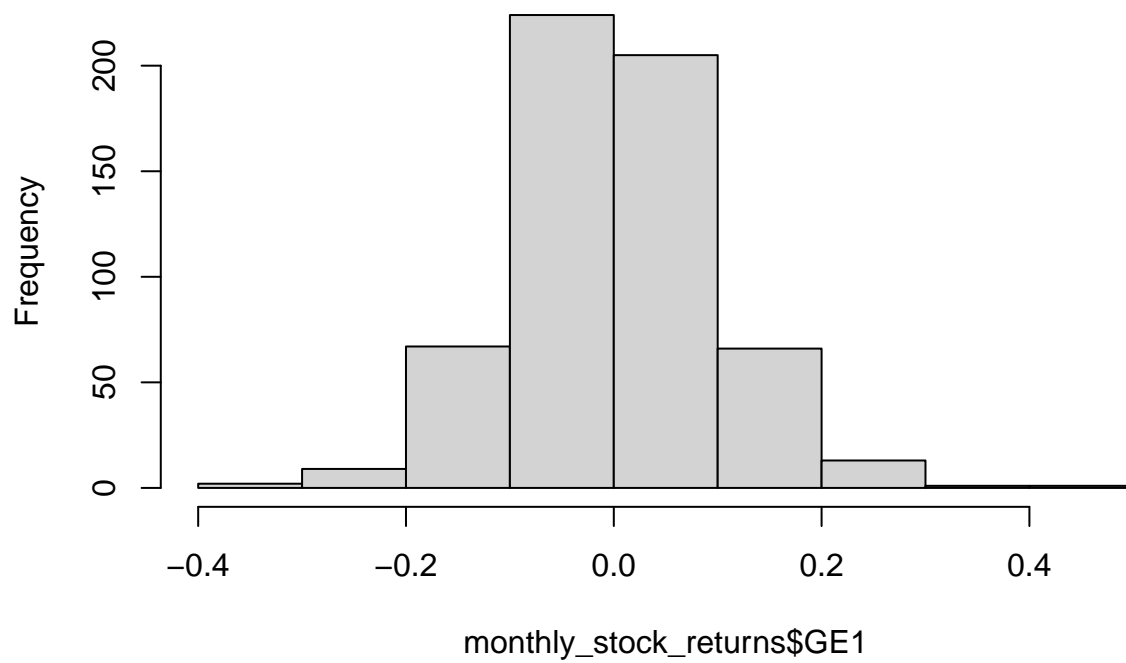
```
hist(monthly_stock_returns$COCACOLA1)
```

Histogram of monthly_stock_returns\$COCACOLA1



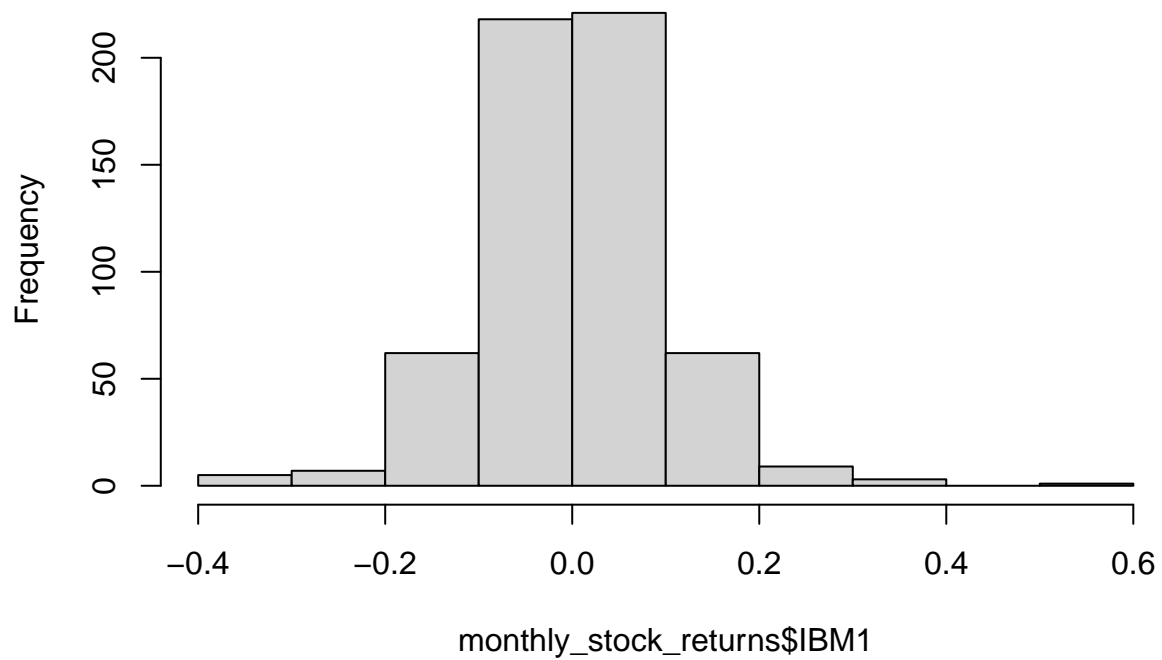
```
hist(monthly_stock_returns$GE1)
```

Histogram of monthly_stock_returns\$GE1



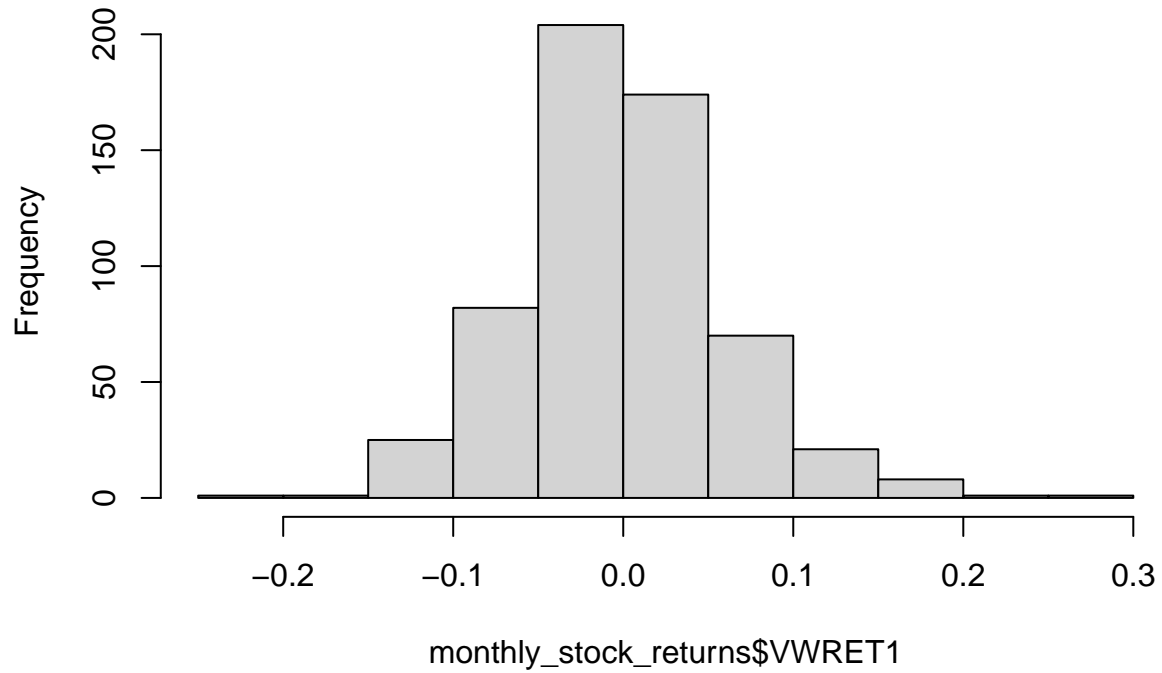
```
hist(monthly_stock_returns$IBM1)
```

Histogram of monthly_stock_returns\$IBM1



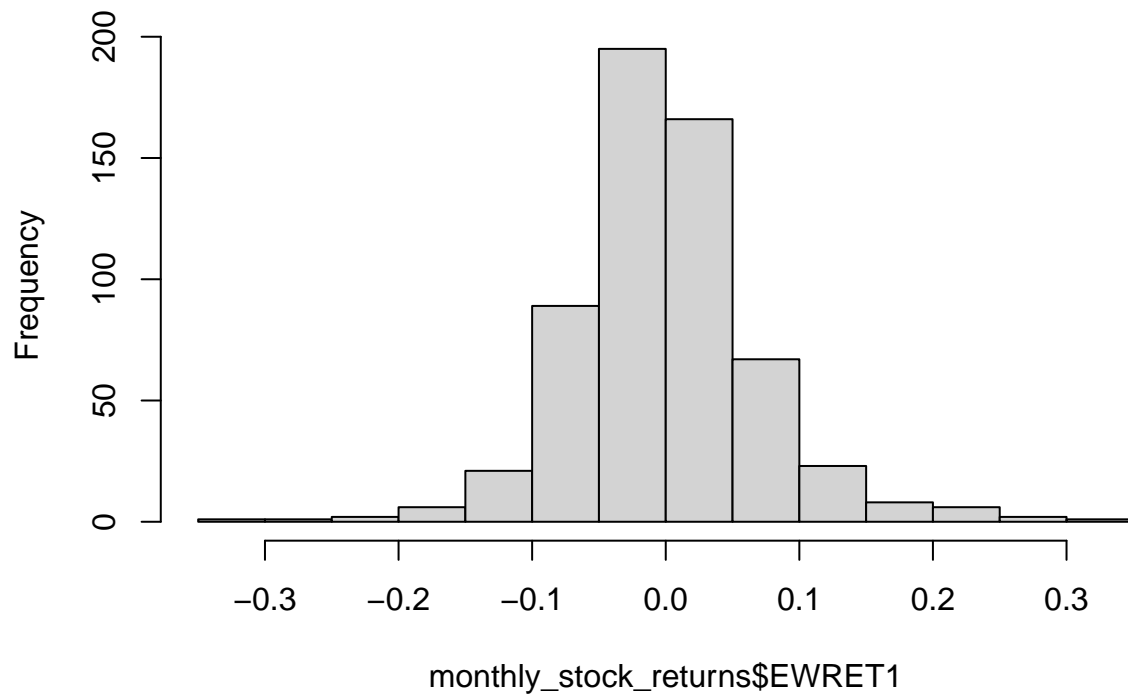
```
hist(monthly_stock_returns$VWRET1)
```

Histogram of monthly_stock_returns\$VWRET1



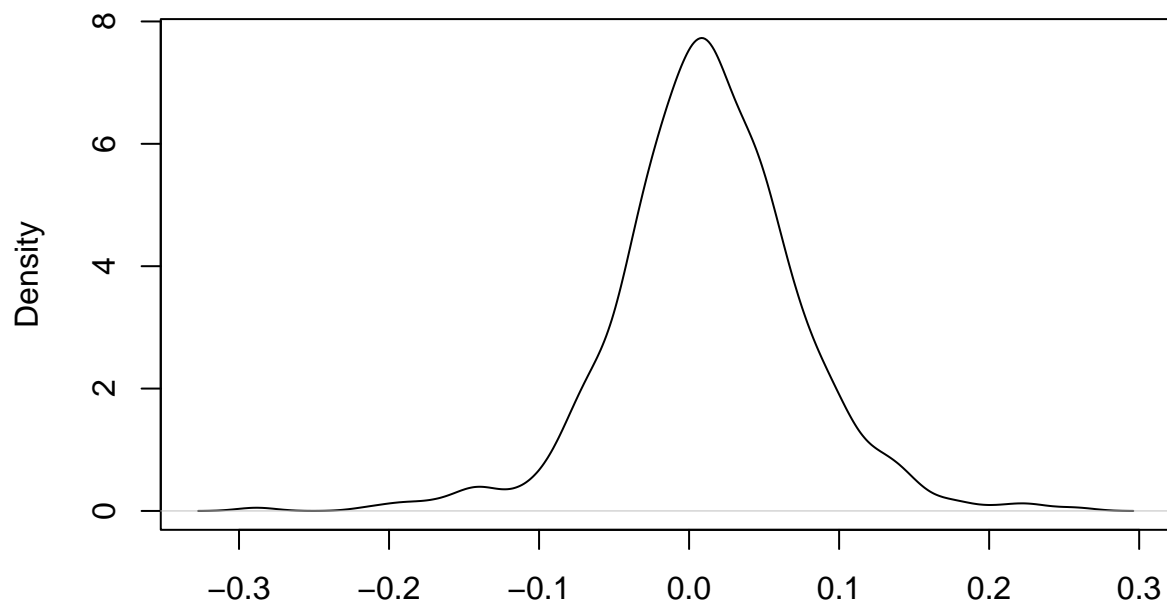
```
hist(monthly_stock_returns$EWRET1)
```

Histogram of monthly_stock_returns\$EWRET1



```
plot(density(monthly_stock_returns$COCACOLA))
```

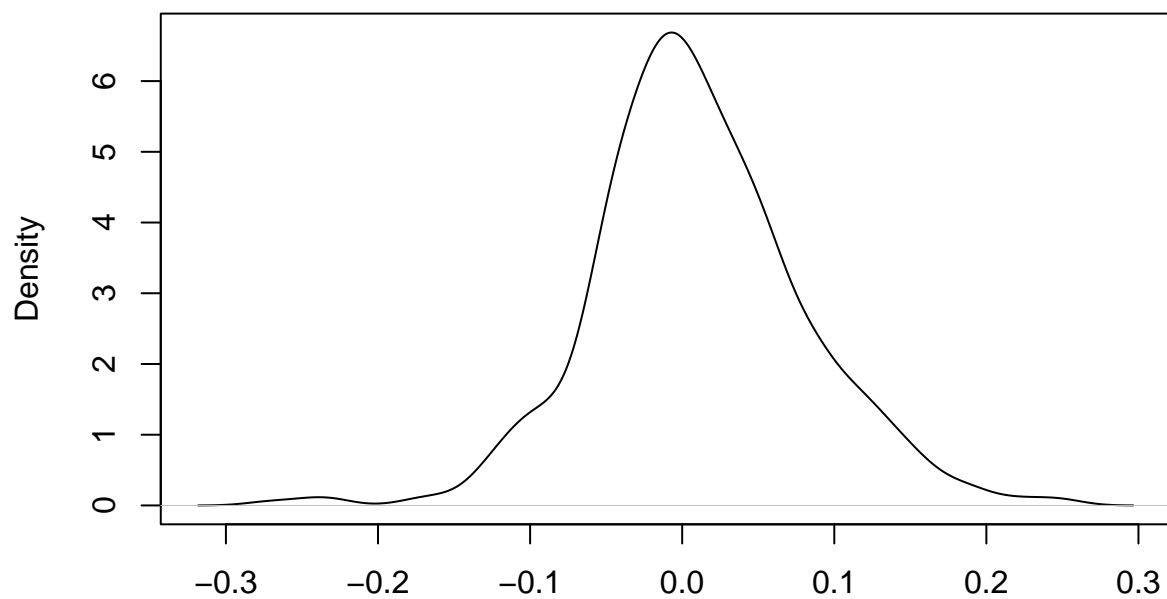
density.default(x = monthly_stock_returns\$COACOLA)



N = 588 Bandwidth = 0.013

```
plot(density(monthly_stock_returns$GE))
```

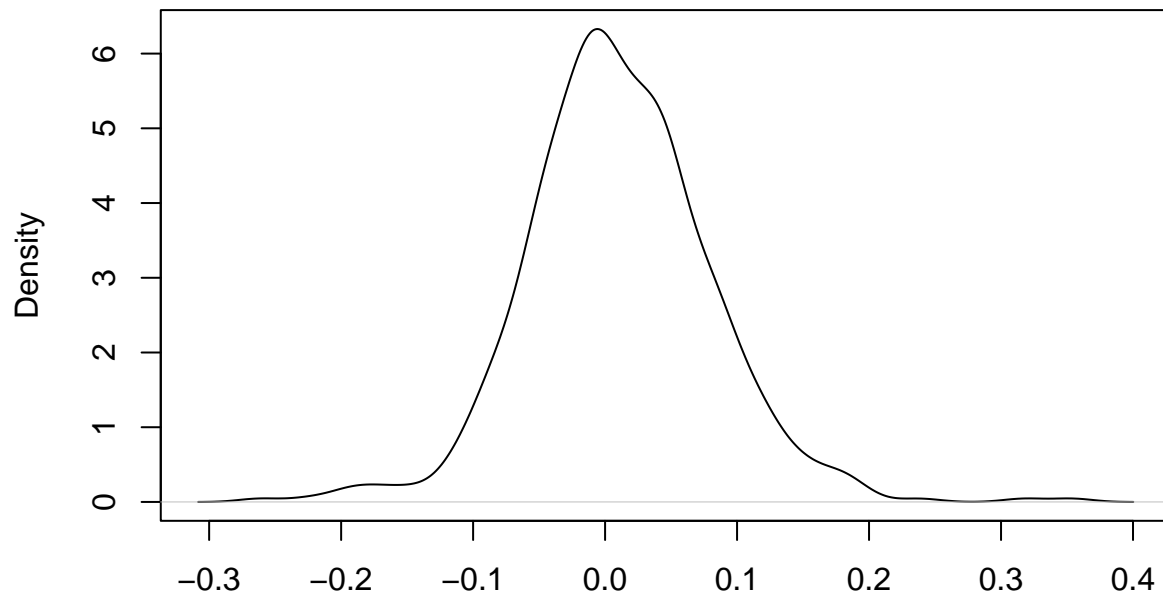
density.default(x = monthly_stock_returns\$GE)



N = 588 Bandwidth = 0.01512

```
plot(density(monthly_stock_returns$IBM))
```

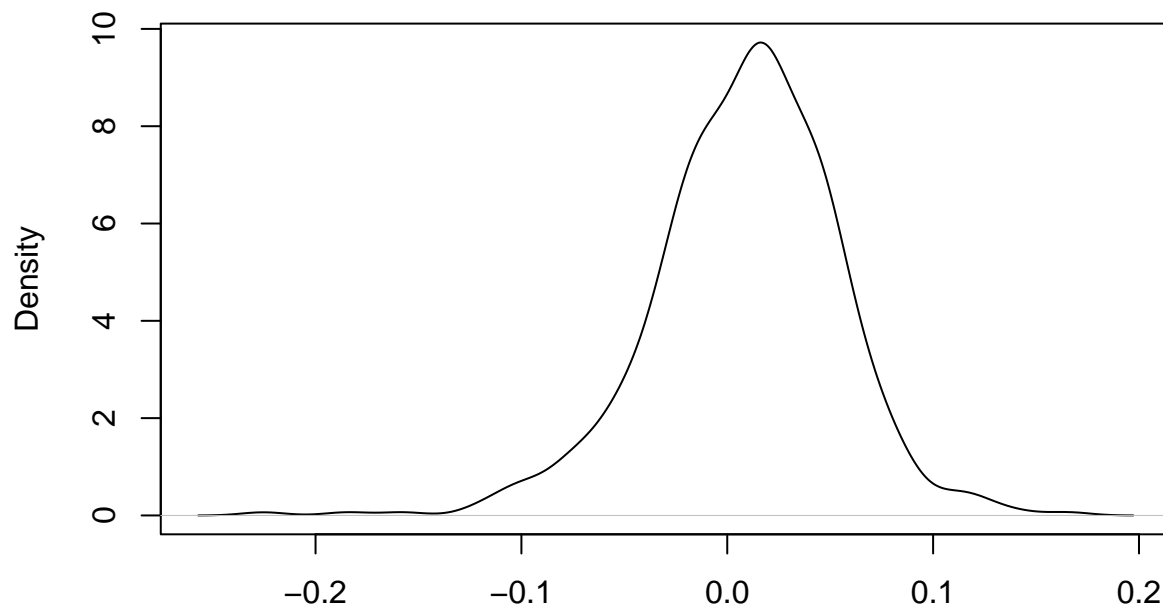
density.default(x = monthly_stock_returns\$IBM)



N = 588 Bandwidth = 0.01547

```
plot(density(monthly_stock_returns$VWRET))
```

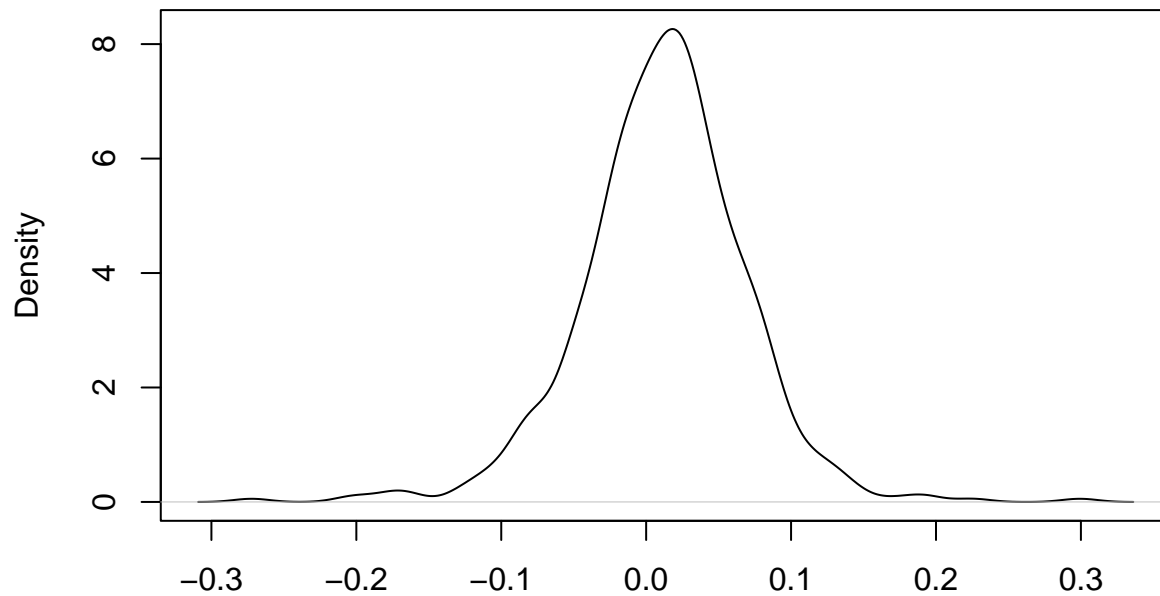
density.default(x = monthly_stock_returns\$VWRET)



N = 588 Bandwidth = 0.01055

```
plot(density(monthly_stock_returns$EWRET))
```

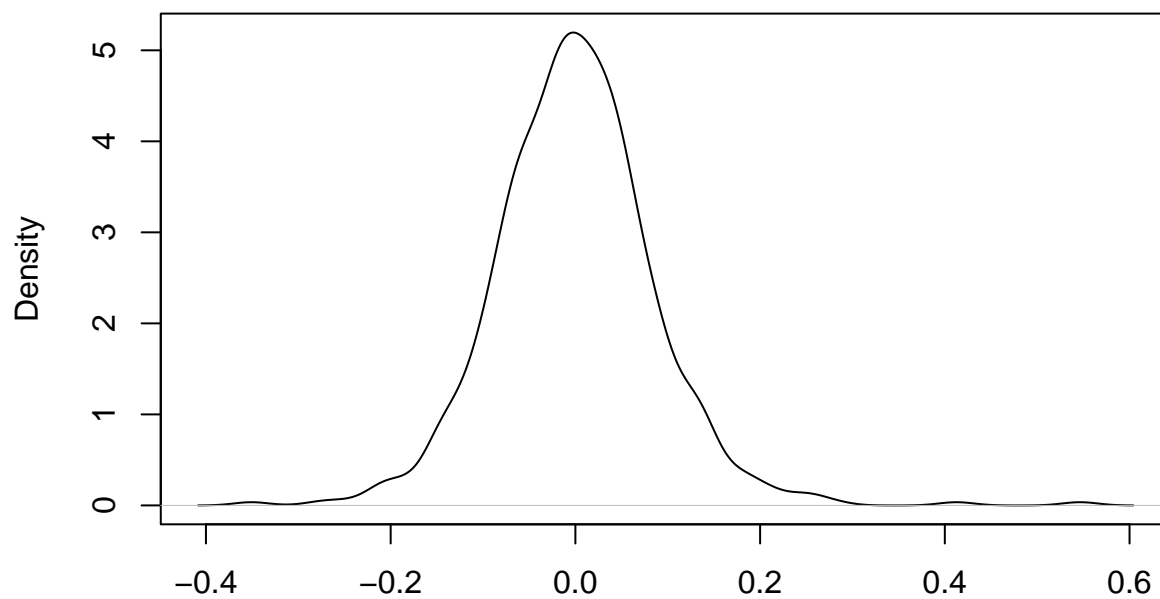
density.default(x = monthly_stock_returns\$EWRET)



N = 588 Bandwidth = 0.0123

```
plot(density(monthly_stock_returns$COCACOLA1))
```

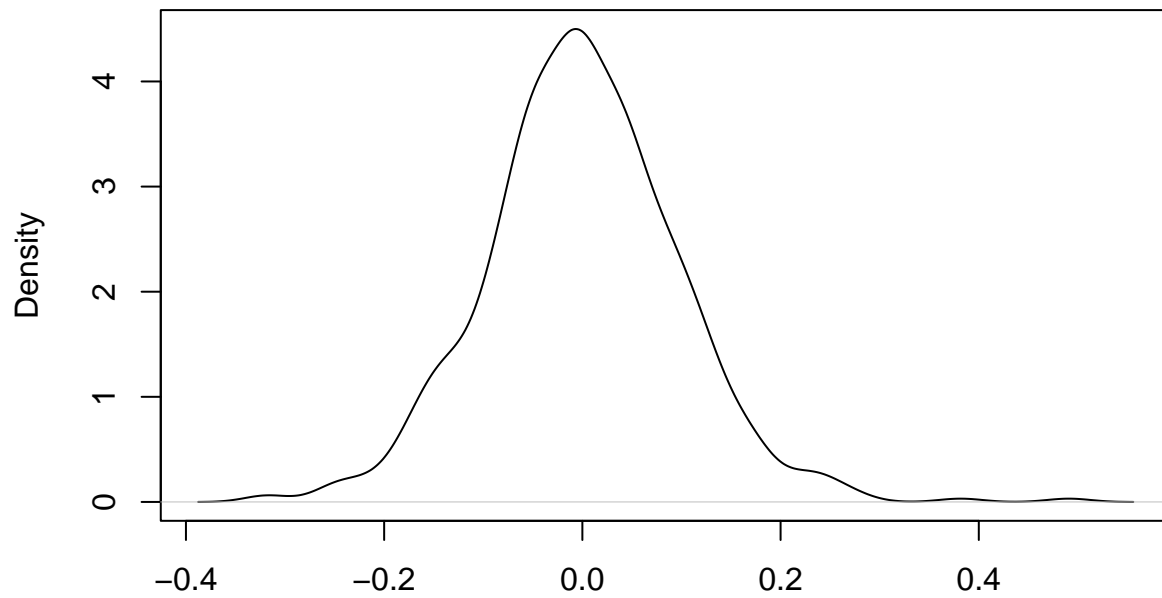
density.default(x = monthly_stock_returns\$COCACOLA1)



N = 588 Bandwidth = 0.01922

```
plot(density(monthly_stock_returns$GE1))
```

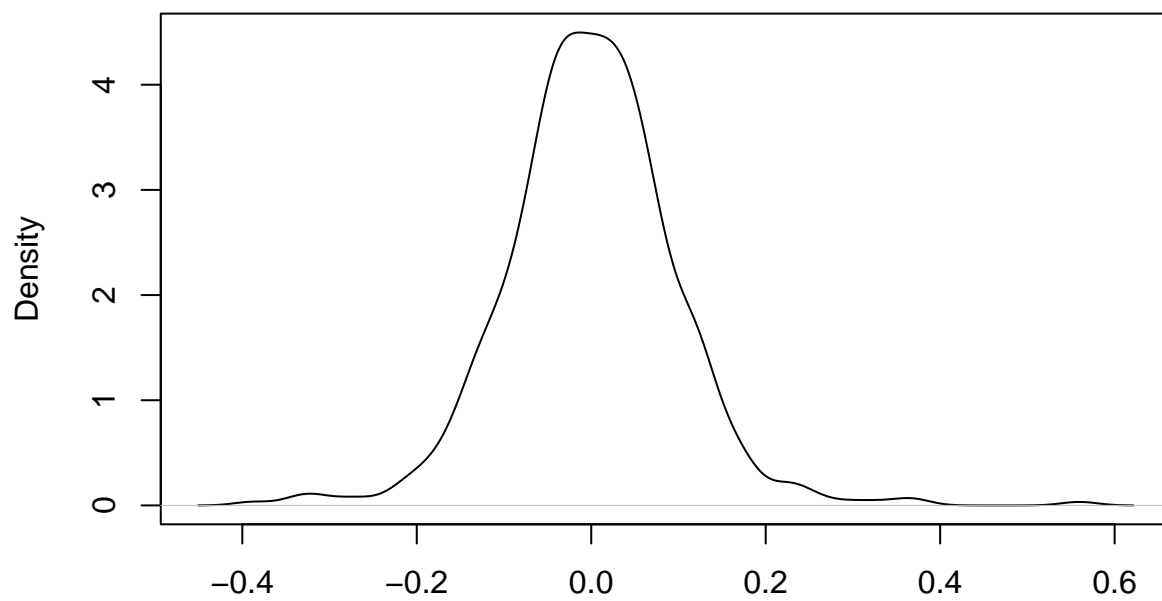
density.default(x = monthly_stock_returns\$GE1)



N = 588 Bandwidth = 0.02164

```
plot(density(monthly_stock_returns$IBM1))
```

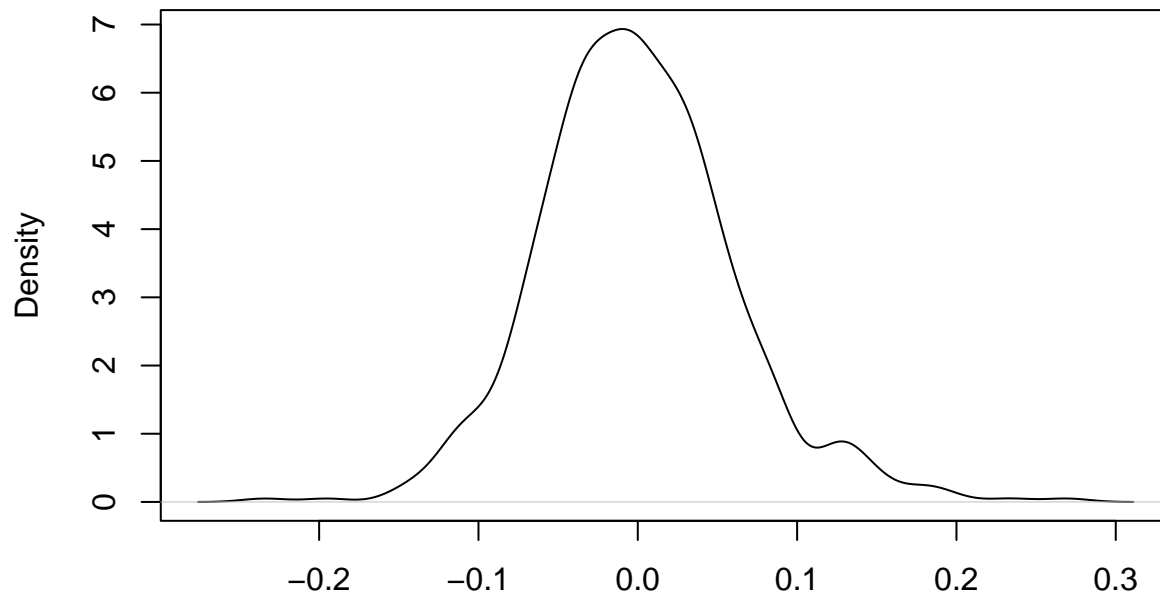
density.default(x = monthly_stock_returns\$IBM1)



N = 588 Bandwidth = 0.02057

```
plot(density(monthly_stock_returns$VWRET1))
```

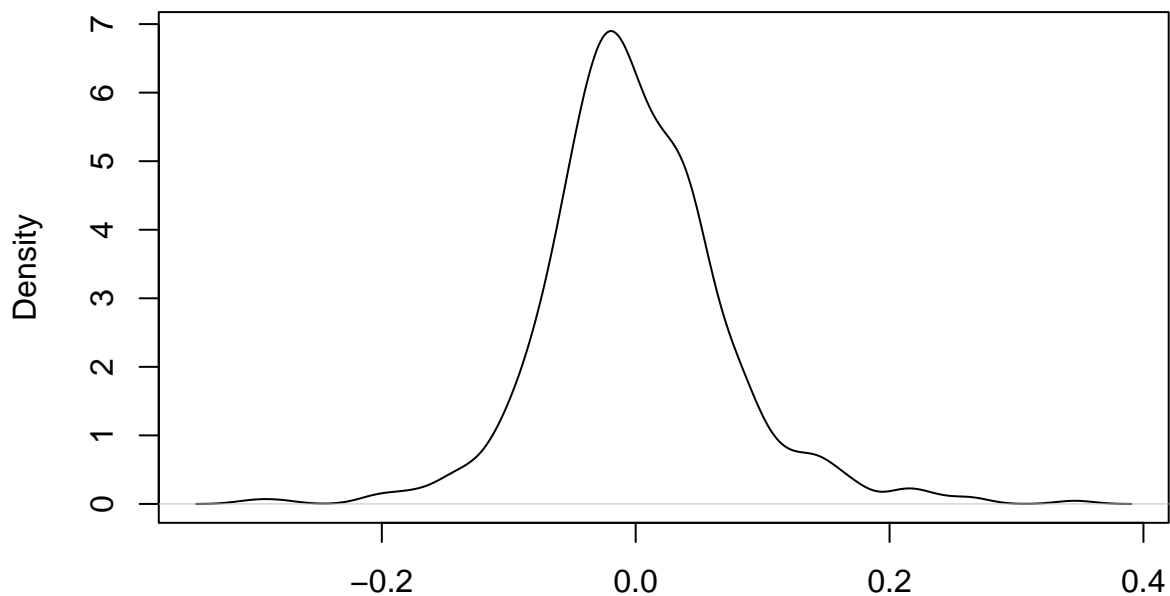

density.default(x = monthly_stock_returns\$VWRET1)



N = 588 Bandwidth = 0.01385

```
plot(density(monthly_stock_returns$EWRET1))  
  
#JARQUE - BERA TEST AT THE 5% SIGNIFICANCE LEVEL #  
  
library(tseries)  
  
## Registered S3 method overwritten by 'quantmod':  
##   method      from  
## as.zoo.data.frame zoo
```

density.default(x = monthly_stock_returns\$EWRET1)



N = 588 Bandwidth = 0.01481

```
jarque.bera.test(monthly_stock_returns$COCACOLA)
```

```
##  
## Jarque Bera Test  
##  
## data: monthly_stock_returns$COCACOLA  
## X-squared = 115.67, df = 2, p-value < 2.2e-16
```

```
jarque.bera.test(monthly_stock_returns$IBM)
```

```
##  
## Jarque Bera Test  
##  
## data: monthly_stock_returns$IBM  
## X-squared = 102.37, df = 2, p-value < 2.2e-16
```

```
jarque.bera.test(monthly_stock_returns$GE)
```

```
##  
## Jarque Bera Test  
##  
## data: monthly_stock_returns$GE  
## X-squared = 37.488, df = 2, p-value = 7.238e-09
```

```
jarque.bera.test(monthly_stock_returns$EWRET)
```

```
##  
## Jarque Bera Test  
##  
## data: monthly_stock_returns$EWRET  
## X-squared = 202.95, df = 2, p-value < 2.2e-16
```

```
jarque.bera.test(monthly_stock_returns$VWRET)

##
## Jarque Bera Test
##
## data: monthly_stock_returns$VWRET
## X-squared = 124.2, df = 2, p-value < 2.2e-16
jarque.bera.test(monthly_stock_returns$COCACOLA1)
```

```
##
## Jarque Bera Test
##
## data: monthly_stock_returns$COCACOLA1
## X-squared = 376.96, df = 2, p-value < 2.2e-16
jarque.bera.test(monthly_stock_returns$IBM1)
```

```
##
## Jarque Bera Test
##
## data: monthly_stock_returns$IBM1
## X-squared = 224.36, df = 2, p-value < 2.2e-16
jarque.bera.test(monthly_stock_returns$GE1)
```

```
##
## Jarque Bera Test
##
## data: monthly_stock_returns$GE1
## X-squared = 56.81, df = 2, p-value = 4.611e-13
jarque.bera.test(monthly_stock_returns$EWRET1)
```

```
##
## Jarque Bera Test
##
## data: monthly_stock_returns$EWRET1
## X-squared = 179.56, df = 2, p-value < 2.2e-16
jarque.bera.test(monthly_stock_returns$VWRET1)
```

```
##
## Jarque Bera Test
##
## data: monthly_stock_returns$VWRET1
## X-squared = 62.027, df = 2, p-value = 3.397e-14
```

Answer: Since the p-values from the result of the jarque-bera tests for the returns and log returns of COCACOLA, GE, IBM, VWRET and EWRET are less than 5%, then we reject the null hypothesis and conclude that the returns and log returns do not follow a normal distribution.

#1H. CAPM REGRESSION FOR COCACOLA, GE AND IBM #

```
COCACOLA_FIT <- lm(COCACOLA ~ VWRET, data = monthly_stock_returns)
GE_FIT <- lm(GE ~ VWRET, data = monthly_stock_returns)
IBM_FIT <- lm(IBM ~ VWRET, data = monthly_stock_returns)
summary(COCACOLA_FIT)
```

```
##
## Call:
## lm(formula = COCACOLA ~ VWRET, data = monthly_stock_returns)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.216102 -0.029532 -0.001404  0.030982  0.171847
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.006496   0.002206   2.944  0.00336 **
## VWRET        0.716446   0.048027  14.918 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.05251 on 586 degrees of freedom
## Multiple R-squared:  0.2752, Adjusted R-squared:  0.274
## F-statistic: 222.5 on 1 and 586 DF, p-value: < 2.2e-16
```

```
summary(GE_FIT)
```

```
##
## Call:
## lm(formula = GE ~ VWRET, data = monthly_stock_returns)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.174175 -0.030702 -0.003161  0.029411  0.198822
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.0004826  0.0020217   0.239   0.811
## VWRET        1.1199868  0.0440096  25.449 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.04812 on 586 degrees of freedom
## Multiple R-squared:  0.525, Adjusted R-squared:  0.5242
## F-statistic: 647.6 on 1 and 586 DF, p-value: < 2.2e-16
```

```
summary(IBM_FIT)
```

```
##
## Call:
## lm(formula = IBM ~ VWRET, data = monthly_stock_returns)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.280296 -0.031656 -0.001447  0.028898  0.285129
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.002762   0.002421   1.141   0.254
## VWRET        0.879652   0.052696  16.693 <2e-16 ***
## ---
```

```
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.05762 on 586 degrees of freedom
## Multiple R-squared:  0.3223, Adjusted R-squared:  0.3211
## F-statistic: 278.7 on 1 and 586 DF,  p-value: < 2.2e-16
```

Answer: Please note that the VWRET was selected as the proxy for market returns because the market portfolio is a value-weighted portfolio of all securities traded in the market. The alphas for COCACOLA, GE and IBM are 0.006496, 0.0004826 and 0.002762 respectively while the betas for COCACOLA, GE and IBM are 0.716446, 1.1199868 and 0.879652 respectively. The p-value of the betas are less than 1.96, hence at 5% significance levels we do not reject the null hypothesis and conclude that are statistically significant.

QUESTION 2

#2a. IMPORT DATA INTO R #

```
library("haven")
```

```
pension_df <- read_dta("Documents/4328 - Applied Financial Econometrics/Assignment/Assignment 1/HW1data/
```

#2b. HOW MANY SINGLE PERSON HOUSEHOLDS ARE THERE IN THE DATASET #

```
library("dplyr")
```

```
## Warning: package 'dplyr' was built under R version 4.1.2
```

```
##
```

```
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      filter, lag
```

```
## The following objects are masked from 'package:base':
```

```
##
```

```
##      intersect, setdiff, setequal, union
```

```
no_of_single_person_households <- length(which(pension_df$marr == 0))
```

```
no_of_single_person_households
```

```
## [1] 3445
```

#2c. OLS ESTIMATION OF THE MODEL TO REFLECT RESULTS OF SINGLE-PERSON HOUSEHOLDS #

```
OLS_estimetea <- lm(nettfa ~ inc + age, data = pension_df)
```

```
summary(OLS_estimetea)
```

```
##
```

```
## Call:
```

```
## lm(formula = nettfa ~ inc + age, data = pension_df)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -509.27  -18.71   -4.09   10.02  1464.74
```

```
##
```

```
## Coefficients:
```

```
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -60.69654    2.59633  -23.38  <2e-16 ***
## inc         0.95336     0.02528   37.72  <2e-16 ***
## age         1.03078     0.05912   17.43  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 58.31 on 9272 degrees of freedom
## Multiple R-squared:  0.1691, Adjusted R-squared:  0.1689
## F-statistic: 943.2 on 2 and 9272 DF,  p-value: < 2.2e-16

OLS_estimate_single <- lm(nettf_a ~ inc + age, data = subset(pension_df, marr == 0))

summary(OLS_estimate_single)
```

```
##
## Call:
## lm(formula = nettf_a ~ inc + age, data = subset(pension_df, marr ==
## 0))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -182.70  -14.17   -3.73    5.77  1113.35
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -43.97943    3.50496  -12.55  <2e-16 ***
## inc         0.87987     0.05169   17.02  <2e-16 ***
## age         0.80046     0.07884   10.15  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 47.77 on 3442 degrees of freedom
## Multiple R-squared:  0.1076, Adjusted R-squared:  0.1071
## F-statistic: 207.5 on 2 and 3442 DF,  p-value: < 2.2e-16
```

#2d. MEANING OF INTERCEPT FROM THE REGRESSION # Answer: The intercept shows that single people have negative net financial wealth of -43-97943 if income and age equals zero. However, this negative net financial wealth position is worse for married people at an intercept of -60-69654.

#2g. OLS ESTIMATION OF THE MODEL

```
OLS_estimate1 <- lm(nettf_a ~ inc, data = pension_df)

summary(OLS_estimate1)
```

```
##
## Call:
## lm(formula = nettf_a ~ inc, data = pension_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -504.39  -18.10   -4.29    6.73  1475.04
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) -20.17948    1.17643   -17.15   <2e-16 ***
## inc          0.99991    0.02554    39.15   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 59.26 on 9273 degrees of freedom
## Multiple R-squared:  0.1418, Adjusted R-squared:  0.1417
## F-statistic: 1532 on 1 and 9273 DF,  p-value: < 2.2e-16

#2h. OLS ESTIMATION OF THE UPDATED MODEL
OLS_estimate2 <- lm(netffa ~ inc + age + incsq + agesq + fsize, data = pension_df)

summary(OLS_estimate2)

##
## Call:
## lm(formula = netffa ~ inc + age + incsq + agesq + fsize, data = pension_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -523.05  -15.85   -2.97    5.62  1466.00
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 13.9925513   9.9919763   1.400   0.1614
## inc         -0.1266417   0.0732205  -1.730   0.0837 .
## age         -1.2580134   0.4918240  -2.558   0.0105 *
## incsq        0.0094893   0.0005837  16.257 < 2e-16 ***
## agesq        0.0263771   0.0056527   4.666 3.11e-06 ***
## fsize       -1.9792085   0.4011045  -4.934 8.18e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 57.26 on 9269 degrees of freedom
## Multiple R-squared:  0.1989, Adjusted R-squared:  0.1985
## F-statistic: 460.4 on 5 and 9269 DF,  p-value: < 2.2e-16

#2i. F-TEST OF THE RESTRICTION IN MODEL (OLS_estimate2) #

library(car)

## Loading required package: carData
## Warning: package 'carData' was built under R version 4.1.2
##
## Attaching package: 'car'
## The following object is masked from 'package:dplyr':
##
##      recode
nullhyp <- c("incsq", "agesq")
linearHypothesis(OLS_estimate2, nullhyp)

## Linear hypothesis test
##
## Hypothesis:
```

```
## incsq = 0
## agesq = 0
##
## Model 1: restricted model
## Model 2: nettfa ~ inc + age + incsq + agesq + fsize
##
##   Res.Df      RSS Df Sum of Sq    F    Pr(>F)
## 1   9271 31361180
## 2   9269 30394935  2    966245 147.33 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Answer: The probability value of the f-test is less than 1%, hence we reject the null hypothesis at 1% significance level and conclude that the coefficients of incsq and agesq do not equal zero.

#2j. RE-ESTIMATION OF THE MODEL (OLS_estimate2) AFTER RESCALING inc BY DIVIDING IT BY 10#

```
pension_df$inc_adjusted <- pension_df$inc/10

OLS_estimate2_adj <- lm(nettfa ~ inc_adjusted + age + I(inc^2) + I(age^2) + fsize, data = pension_df)

summary(OLS_estimate2_adj)
```

```
##
## Call:
## lm(formula = nettfa ~ inc_adjusted + age + I(inc^2) + I(age^2) +
##     fsize, data = pension_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -523.05  -15.85   -2.97    5.62  1466.00
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  13.9925514  9.9919763   1.400   0.1614
## inc_adjusted -1.2664166  0.7322049  -1.730   0.0837 .
## age         -1.2580134  0.4918240  -2.558   0.0105 *
## I(inc^2)      0.0094893  0.0005837  16.257 < 2e-16 ***
## I(age^2)      0.0263771  0.0056527   4.666 3.11e-06 ***
## fsize        -1.9792085  0.4011045  -4.934 8.18e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 57.26 on 9269 degrees of freedom
## Multiple R-squared:  0.1989, Adjusted R-squared:  0.1985
## F-statistic: 460.4 on 5 and 9269 DF,  p-value: < 2.2e-16
```

#2k. TESTING THE MODEL (OLS_estimate) FOR HETEROSKEDASTICITY USING THE BREUSCH-PAGAN TEST #

```
library(lmtest)
```

```
## Loading required package: zoo
##
## Attaching package: 'zoo'
```



```
## The following objects are masked from 'package:base':
##
##   as.Date, as.Date.numeric
```

```
bptest(OLS_estimatea)
```

```
##
## studentized Breusch-Pagan test
##
## data: OLS_estimatea
## BP = 189.51, df = 2, p-value < 2.2e-16
```

Answer: the p-value is less than the significance level at 1% or 5%, hence we reject the null hypothesis and conclude that heteroskedasticity is present

#2l. ESTIMATING THE MODEL (OLS_estimate) WITH HETEROSKEDASTICITY-ROBUST (e.g. WHITE) STANDARD ERRORS #

```
library(lmtest)
library(sandwich)
coeftest(OLS_estimatea, vcov = sandwich)
```

```
##
## t test of coefficients:
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -60.696537   3.750371 -16.184 < 2.2e-16 ***
## inc          0.953357   0.069936  13.632 < 2.2e-16 ***
## age          1.030777   0.065054  15.845 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

#2m. RE-ESTIMATING THE MODEL (OLS_estimate) AFTER STANDARDISING ALL VARIABLES IN THE MODEL. INTERPRETE COEFFICIENTS ON inc AND age #

```
OLS_estimatea <- lm(nettfa ~ inc + age, data = pension_df)
pension_df$resi <- OLS_estimatea$residuals
varfunc.ols <- lm(log(resi^2) ~ log(inc) + log(age), data = pension_df)
pension_df$varfunc <- exp(varfunc.ols$fitted.values)
OLS_estimatea_gls <- lm(nettfa ~ inc + age, weights = 1/sqrt(varfunc), data = pension_df)

summary(OLS_estimatea)
```

```
##
## Call:
## lm(formula = nettfa ~ inc + age, data = pension_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -509.27  -18.71   -4.09   10.02  1464.74
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept) -60.69654    2.59633  -23.38  <2e-16 ***
## inc          0.95336    0.02528   37.72  <2e-16 ***
## age          1.03078    0.05912   17.43  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## Residual standard error: 58.31 on 9272 degrees of freedom
## Multiple R-squared:  0.1691, Adjusted R-squared:  0.1689
## F-statistic: 943.2 on 2 and 9272 DF,  p-value: < 2.2e-16

summary(OLS_estimatea_gls)

##
## Call:
## lm(formula = nettfa ~ inc + age, data = pension_df, weights = 1/sqrt(varfunc))
##
## Weighted Residuals:
##      Min       1Q   Median       3Q      Max
## -160.311   -4.565   -1.772    1.881   301.278
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -38.56580     1.69027  -22.82  <2e-16 ***
## inc           0.72560     0.02222   32.65  <2e-16 ***
## age           0.69655     0.04104   16.97  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

## Residual standard error: 11.85 on 9272 degrees of freedom
## Multiple R-squared:  0.1382, Adjusted R-squared:  0.138
## F-statistic: 743.6 on 2 and 9272 DF,  p-value: < 2.2e-16
```

QUESTION 3

IMPORT DATA INTO R

```
library("haven")
library("sandwich")
library(foreign)

## Warning: package 'foreign' was built under R version 4.1.2

library("lmtest")
library("zoo")
library(dplyr)

ceo_salary_df <- read_dta("Documents/4328 - Applied Financial Econometrics/Assignment/Assignment 1/HW1data.dta")

#3a. Estimation of the model using OLS

OLS_estimate_salary <- lm(lsalary ~ lsales + lmktval + ceoten + ceotensq, data = ceo_salary_df)

summary(OLS_estimate_salary)
```

```
##
## Call:
## lm(formula = lsalary ~ lsales + lmktval + ceoten + ceotensq,
##      data = ceo_salary_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
```

```
## -2.41976 -0.28791 0.00253 0.28615 1.74966
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.3685503  0.2587397  16.884 < 2e-16 ***
## lsales       0.1646331  0.0386393   4.261 3.35e-05 ***
## lmktval      0.1085285  0.0488257   2.223 0.02753 *
## ceoten       0.0451169  0.0141169   3.196 0.00166 **
## ceotensq     -0.0012102  0.0004747  -2.549 0.01167 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4969 on 172 degrees of freedom
## Multiple R-squared:  0.343, Adjusted R-squared:  0.3277
## F-statistic: 22.45 on 4 and 172 DF, p-value: 6.257e-15
```

#3b. COUNT, MEAN, STANDARD DEVIATION, MINIMUM AND MAXIMUM OF EXPLANATORY VARIABLES. EXPLAIN REASON FOR NATURAL LOG

```
lsales. <- c(length(ceo_salary_df$lsales), mean(ceo_salary_df$lsales), sd(ceo_salary_df$lsales), min(ceo_salary_df$lsales), max(ceo_salary_df$lsales))
lmktval. <- c(length(ceo_salary_df$lmktval), mean(ceo_salary_df$lmktval), sd(ceo_salary_df$lmktval), min(ceo_salary_df$lmktval), max(ceo_salary_df$lmktval))
ceoten. <- c(length(ceo_salary_df$ceoten), mean(ceo_salary_df$ceoten), sd(ceo_salary_df$ceoten), min(ceo_salary_df$ceoten), max(ceo_salary_df$ceoten))
ceotensq. <- c(length(ceo_salary_df$ceotensq), mean(ceo_salary_df$ceotensq), sd(ceo_salary_df$ceotensq), min(ceo_salary_df$ceotensq), max(ceo_salary_df$ceotensq))

salary_summary_df <- data.frame(lsales., lmktval., ceoten., ceotensq.)
row.names(salary_summary_df) <- c("COUNT", "MEAN", "STANDARD DEVIATION", "MINIMUM", "MAXIMUM")
salary_summary_df
```

```
##             lsales.  lmktval.  ceoten. ceotensq.
## COUNT          177.000000 177.000000 177.000000 177.0000
## MEAN             7.231025  7.399410  7.954802 114.1243
## STANDARD DEVIATION 1.432086  1.133414  7.150826 212.5660
## MINIMUM          3.367296  5.958425  0.000000  0.0000
## MAXIMUM          10.845446 10.723268 37.000000 1369.0000
```

#3c. RESTIMATION OF MODEL WITH WHITE STANDARD ERRORS AND THE T-STATS

```
coeftest(OLS_estimate_salary, vcov = sandwich)
```

```
##
## t test of coefficients:
##
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.36855032  0.26010837 16.7951 < 2.2e-16 ***
## lsales       0.16463314  0.03791554  4.3421 2.404e-05 ***
## lmktval      0.10852852  0.04877684  2.2250 0.027382 *
## ceoten       0.04511688  0.01412824  3.1934 0.001672 **
## ceotensq     -0.00121019  0.00054608  -2.2161 0.027994 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
coeftest(OLS_estimate_salary, vcov = vcovHC(OLS_estimate_salary, "HCO"))
```

```
##
## t test of coefficients:
##
##             Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) 4.36855032 0.26010837 16.7951 < 2.2e-16 ***
## lsales      0.16463314 0.03791554 4.3421 2.404e-05 ***
## lmktval     0.10852852 0.04877684 2.2250 0.027382 *
## ceoten      0.04511688 0.01412824 3.1934 0.001672 **
## ceotensq    -0.00121019 0.00054608 -2.2161 0.027994 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
coeftest(OLS_estimate_salary, vcov = vcovHC(OLS_estimate_salary))
```

```
##
## t test of coefficients:
##
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.36855032 0.27604986 15.8252 < 2.2e-16 ***
## lsales      0.16463314 0.03961859 4.1555 5.108e-05 ***
## lmktval     0.10852852 0.05163631 2.1018 0.037029 *
## ceoten      0.04511688 0.01636159 2.7575 0.006454 **
## ceotensq    -0.00121019 0.00068165 -1.7754 0.077603 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
summary(OLS_estimate_salary)
```

```
##
## Call:
## lm(formula = lsalary ~ lsales + lmktval + ceoten + ceotensq,
##     data = ceo_salary_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.41976 -0.28791  0.00253  0.28615  1.74966
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.3685503  0.2587397  16.884 < 2e-16 ***
## lsales      0.1646331  0.0386393   4.261 3.35e-05 ***
## lmktval     0.1085285  0.0488257   2.223 0.02753 *
## ceoten      0.0451169  0.0141169   3.196 0.00166 **
## ceotensq    -0.0012102  0.0004747  -2.549 0.01167 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4969 on 172 degrees of freedom
## Multiple R-squared:  0.343, Adjusted R-squared:  0.3277
## F-statistic: 22.45 on 4 and 172 DF, p-value: 6.257e-15
```

#3d. OBTAIN RESIDUALS AND STANDARDIZE

```
residuals <- residuals(OLS_estimate_salary)
residuals
```

```
##           1           2           3           4           5
## 0.0745482432 0.0087696527 -0.1600566223 -0.1995056406 0.0013666286
##           6           7           8           9          10
## -0.1718538017 0.4195761498 0.2619258405 -0.0289205553 0.2237905010
##          11          12          13          14          15
```

##	-0.0751345583	-0.4910547216	0.9017545887	-0.2879058883	0.0659469317
##	16	17	18	19	20
##	-0.6177548927	0.2638780054	0.8461136004	0.2355350816	0.1106099584
##	21	22	23	24	25
##	-0.5525803494	0.0914100372	0.0025346251	-0.5289003987	-0.2840587361
##	26	27	28	29	30
##	-0.2457134327	0.2917024589	-0.3934139595	0.0074422476	-1.3925994169
##	31	32	33	34	35
##	0.3898379325	0.1076785660	-0.4662147270	0.0451350470	-0.3769320893
##	36	37	38	39	40
##	0.1540314879	-0.1983508542	-0.9945930624	0.3447092973	0.0545626236
##	41	42	43	44	45
##	-0.2096120725	0.0752916896	0.3024517243	0.1751854880	0.7778985571
##	46	47	48	49	50
##	-0.4092876556	-0.0665065024	-0.0569332802	-0.3050195721	-0.1622872106
##	51	52	53	54	55
##	-0.3527929027	-0.6704555158	0.3191934229	0.0870063838	0.6387602152
##	56	57	58	59	60
##	0.1508932495	-0.0648512329	-0.0001929742	0.0825673759	0.6734141736
##	61	62	63	64	65
##	-0.0790527852	0.3404305505	-0.2913127871	-0.1156633731	-0.4660592342
##	66	67	68	69	70
##	0.2085487346	0.1531184154	-0.1542875842	0.6367520667	-0.2144786245
##	71	72	73	74	75
##	0.3866402140	-0.0015212773	0.0096457889	1.4453521716	-0.5360540507
##	76	77	78	79	80
##	-0.5195719905	0.4484531571	-0.3020273083	-0.0445622912	-0.0144437724
##	81	82	83	84	85
##	0.0052252415	0.5739850587	0.2799007270	0.3831818594	-0.1468072788
##	86	87	88	89	90
##	-0.7933580273	-0.4888879508	0.1953072497	0.6119965614	-0.6819625978
##	91	92	93	94	95
##	-0.2384441873	-0.1715401714	0.4220239862	0.3861182413	-0.4564460425
##	96	97	98	99	100
##	-0.2475566759	-0.0749548485	-0.3302797128	0.1928829140	0.4438447239
##	101	102	103	104	105
##	0.2671760231	0.0214916075	1.7496562215	-0.2152739652	-0.0060845290
##	106	107	108	109	110
##	0.0416992543	-0.0952660488	0.6401604124	-0.0581459973	0.3631657152
##	111	112	113	114	115
##	0.1955686962	0.2861454428	-2.4197629906	-0.2601548728	-0.2179390655
##	116	117	118	119	120
##	-0.0959731913	0.5784300415	-0.7325487610	0.4872881785	-0.2373378236
##	121	122	123	124	125
##	0.2477222495	-0.4745729914	-0.4186439442	0.5777408225	0.3506959115
##	126	127	128	129	130
##	-0.4865395860	-0.3066477603	-0.3696171351	-0.0441448841	1.1227601097
##	131	132	133	134	135
##	-0.1223753913	-0.1788500726	-0.4046291129	0.2596750265	0.1788586731
##	136	137	138	139	140
##	0.4878536866	-0.3841228046	0.1042148250	-0.1251903926	-0.0164320691
##	141	142	143	144	145
##	-0.7354238081	0.2860410507	0.4175497064	0.3921764334	-0.2677096983
##	146	147	148	149	150

```
## 0.6842659805 0.6810975131 0.0591282829 0.3254489960 0.5329426486
## 151 152 153 154 155
## -0.2972669862 0.2048235407 0.3765686191 0.1474531422 0.4177142940
## 156 157 158 159 160
## -0.3741389456 -0.2562147395 -0.8403145924 -0.5428789985 0.5250499136
## 161 162 163 164 165
## 0.8060669846 -0.5842859240 0.1227519861 0.4799655898 0.6134723818
## 166 167 168 169 170
## -0.4770502259 0.0995704022 0.0416215944 -0.7248350873 0.0688208279
## 171 172 173 174 175
## -0.2124112209 -0.8596249461 -0.5437982863 -0.9722212184 -0.4605385424
## 176 177
## 1.3079955513 0.0919640239
## attr("label")
## [1] "log(salary)"
## attr("format.stata")
## [1] "%9.0g"
```

```
standardised_residuals <- rstandard(OLS_estimate_salary)
standardised_residuals
```

```
## 1 2 3 4 5
## 0.1537627164 0.0178263441 -0.3267869642 -0.4074059778 0.0027723001
## 6 7 8 9 10
## -0.3512278504 0.8505510750 0.5309998045 -0.0586667840 0.4525889639
## 11 12 13 14 15
## -0.1525508901 -0.9957036137 2.1528202078 -0.5867724651 0.1342193252
## 16 17 18 19 20
## -1.2575305050 0.5340577232 1.7154888376 0.4846139495 0.2247724829
## 21 22 23 24 25
## -1.1160667091 0.1855321639 0.0051281525 -1.0818914339 -0.5770761291
## 26 27 28 29 30
## -0.5102404192 0.5905041483 -0.8033058227 0.0150514529 -2.8813141742
## 31 32 33 34 35
## 0.7925881606 0.2185908134 -0.9491191323 0.0927831452 -0.7654534673
## 36 37 38 39 40
## 0.3131279139 -0.4008619471 -2.0402159042 0.7072065077 0.1134104329
## 41 42 43 44 45
## -0.4253861902 0.1539058558 0.6216351178 0.3631083441 1.5844149320
## 46 47 48 49 50
## -0.8284146227 -0.1360160992 -0.1159230761 -0.6327069977 -0.3351307270
## 51 52 53 54 55
## -0.7170123272 -1.3636324861 0.6472105596 0.1774469810 1.2959091880
## 56 57 58 59 60
## 0.3093756637 -0.1325058772 -0.0003903838 0.1680835095 1.3663694247
## 61 62 63 64 65
## -0.1609851635 0.6986723401 -0.5918317348 -0.2340797765 -0.9417097109
## 66 67 68 69 70
## 0.4252500821 0.3141939400 -0.3124566393 1.3022810687 -0.4382440547
## 71 72 73 74 75
## 0.7825914100 -0.0031073294 0.0200674291 2.9296866253 -1.0844160179
## 76 77 78 79 80
## -1.0499724674 0.9178886591 -0.6140356504 -0.0905215539 -0.0292933890
## 81 82 83 84 85
## 0.0106195242 1.1636551859 0.5665266203 0.7748161526 -0.2997521486
```

```
##          86          87          88          89          90
## -1.6479671869 -0.9879370241  0.3966670349  1.2376762344 -1.3818081388
##          91          92          93          94          95
## -0.4883501827 -0.3483918126  0.8627610744  0.7831483885 -0.9259884038
##          96          97          98          99         100
## -0.5008282826 -0.1520852924 -0.6694739059  0.3945736799  0.8993774245
##         101         102         103         104         105
##  0.5451288813  0.0436587030  3.5522777545 -0.4367748226 -0.0123065805
##         106         107         108         109         110
##  0.0854605068 -0.1956556283  1.2958826924 -0.1181536708  0.7375290244
##         111         112         113         114         115
##  0.4008544625  0.5878715853 -5.0449182642 -0.5285123564 -0.4419209749
##         116         117         118         119         120
## -0.1955726841  1.1970194818 -1.5050156993  0.9921361568 -0.4815370193
##         121         122         123         124         125
##  0.5931891944 -0.9876213219 -0.8476258513  1.1890639423  0.7083833390
##         126         127         128         129         130
## -0.9928729580 -0.6280643880 -0.7584891517 -0.0902097464  2.2801892152
##         131         132         133         134         135
## -0.2488673862 -0.3620685545 -0.8242829121  0.5260487068  0.3756878393
##         136         137         138         139         140
##  1.0883011287 -0.7833841134  0.2127112604 -0.2545212330 -0.0332731061
##         141         142         143         144         145
## -1.4952336813  0.5798309280  0.8460231410  0.7971370503 -0.5439654235
##         146         147         148         149         150
##  1.3975018824  1.3790141943  0.1201795586  0.6611144651  1.0796482172
##         151         152         153         154         155
## -0.6027205460  0.4150867380  0.7684289917  0.3006363144  0.8449449324
##         156         157         158         159         160
## -0.7643260609 -0.5200605645 -1.7056737168 -1.1040363055  1.0682295111
##         161         162         163         164         165
##  1.6447236359 -1.1928278604  0.2486804970  0.9704040578  1.2411025278
##         166         167         168         169         170
## -0.9648321177  0.2033136786  0.0865968456 -1.4710890131  0.1405514993
##         171         172         173         174         175
## -0.4322073847 -1.7433700061 -1.1036728652 -1.9795851440 -0.9367339473
##         176         177
##  2.6691162788  0.1881773844
## attr("label")
## [1] "log(salary)"
## attr("format.stata")
## [1] "%9.0g"
```

#3e. RE-ESTIMATE MODEL BY ADDING DUMMY “COLLEGE” AND INTERACTION TERM "COLLEGE*LSALES". INTERPRETE COEFFICIENT OF INTERACTION TERM

```
OLS_reestimate_salary <- lm(lsalary ~ lsales + lmktval + ceoten + ceotensq + college + college*sales, data = ceo_salary_df)
summary(OLS_reestimate_salary)
```

```
##
## Call:
## lm(formula = lsalary ~ lsales + lmktval + ceoten + ceotensq +
##      college + college * sales, data = ceo_salary_df)
##
## Residuals:
```

```
##      Min      1Q   Median      3Q      Max
## -2.44555 -0.29258  0.00315  0.29093  1.74839
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.704e+00  4.724e-01   9.959  < 2e-16 ***
## lsales       1.627e-01  4.319e-02   3.766  0.000229 ***
## lmktval      1.159e-01  5.189e-02   2.234  0.026821 *
## ceoten       4.586e-02  1.421e-02   3.227  0.001503 **
## ceotensq     -1.209e-03  4.764e-04  -2.537  0.012072 *
## college      -3.852e-01  3.357e-01  -1.147  0.252866
## sales        -9.017e-05  5.901e-05  -1.528  0.128363
## college:sales 9.107e-05  5.829e-05   1.562  0.120072
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4977 on 169 degrees of freedom
## Multiple R-squared:  0.3524, Adjusted R-squared:  0.3255
## F-statistic: 13.14 on 7 and 169 DF,  p-value: 1.847e-13
```

#3f. RE-ESTIMATE MODEL WITH VARIABLE “LSALES_ADJUSTED” INSTEAD OF “LSALES”.
COMPARE COEFFICIENT ON “LSALES_ADJUSTED” WITH “LSALES”

```
ceo_salary_df$lsales_adjusted <- ceo_salary_df$lsales * 0.9
OLS_reestimate_salary2 <- lm(lsalary ~ lsales_adjusted + lmktval + ceoten + ceotensq + college + college * sales, data = ceo_salary_df)
summary(OLS_reestimate_salary2)
```

```
##
## Call:
## lm(formula = lsalary ~ lsales_adjusted + lmktval + ceoten + ceotensq +
##      college + college * sales, data = ceo_salary_df)
##
## Residuals:
##      Min      1Q   Median      3Q      Max
## -2.44555 -0.29258  0.00315  0.29093  1.74839
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  4.704e+00  4.724e-01   9.959  < 2e-16 ***
## lsales_adjusted 1.807e-01  4.799e-02   3.766  0.000229 ***
## lmktval      1.159e-01  5.189e-02   2.234  0.026821 *
## ceoten       4.586e-02  1.421e-02   3.227  0.001503 **
## ceotensq     -1.209e-03  4.764e-04  -2.537  0.012072 *
## college      -3.852e-01  3.357e-01  -1.147  0.252866
## sales        -9.017e-05  5.901e-05  -1.528  0.128363
## college:sales 9.107e-05  5.829e-05   1.562  0.120072
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4977 on 169 degrees of freedom
## Multiple R-squared:  0.3524, Adjusted R-squared:  0.3255
## F-statistic: 13.14 on 7 and 169 DF,  p-value: 1.847e-13
```

Answer: The original coefficient of lsales is less than the coefficient of lsales_adjusted.

QUESTION 4

IMPORT DATA INTO R

```
library(haven)
library(sandwich)
library(foreign)
library(lmtest)
library(zoo)
library(dplyr)
library("plm")

## Warning: package 'plm' was built under R version 4.1.2

##
## Attaching package: 'plm'

## The following objects are masked from 'package:dplyr':
##
##      between, lag, lead

local_returns_df <- read_dta("Documents/4328 - Applied Financial Econometrics/Assignment/Assignment 1/HW

#4a. CONFIRM IF BALANCED OR UNBALANCED PANEL. PROVIDE AGGREGATE STATISTICS

local_returns_panel <- pdata.frame(local_returns_df, index = c("permno", "date"))
is.pbalanced(local_returns_panel)

## [1] FALSE

aggregate(local_returns_panel, list(local_returns_df$year), summary)

##      Group.1 permno.10001 permno.10012 permno.10016 permno.10025 permno.10026
## 1      2000           12           12           12           12           12
## 2      2001           12           12           12           12           12
## 3      2002           12           12           12           12           12
## 4      2003           12           12           12           12           12
## 5      2004           12           12           12           12           12
## 6      2005           12           12           12           12           12
## 7      2006           12           12           12           12           12
## 8      2007           12           12           12           12           12
## 9      2008           12           12           12           12           12
## 10     2009           12           12           12           12           12
## 11     2010           12           12           12           12           12
##      permno.10028 permno.10035 permno.10037 permno.10039 permno.10042
## 1              12              12              12              12              12
## 2              12              12              12              12              12
## 3              12              12              12              12              12
## 4              12              12              12              12              12
## 5              12              12              12              12              12
## 6              12              12              12              12              12
## 7              12              12              12              12              12
## 8              12              12              12              12              12
## 9              12              12              12              12              12
## 10             12              12              12              12              12
## 11             12              12              12              12              12
##      permno.10056 permno.10065 permno.10078 permno.10085 permno.10089
```

## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10092	permno.10100	permno.10104	permno.10107	permno.10108
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10114	permno.10119	permno.10122	permno.10123	permno.10127
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10138	permno.10143	permno.10147	permno.10149	permno.10155
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10160	permno.10170	permno.10177	permno.10180	permno.10192
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12

## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10200	permno.10205	permno.10216	permno.10225	permno.10230
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10235	permno.10238	permno.10239	permno.10244	permno.10252
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10256	permno.10257	permno.10259	permno.10271	permno.10275
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10294	permno.10297	permno.10299	permno.10302	permno.10304
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10318	permno.10324	permno.10325	permno.10333	permno.10341

## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10346	permno.10353	permno.10355	permno.10362	permno.10363
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10371	permno.10375	permno.10382	permno.10383	permno.10395
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10401	permno.10404	permno.10418	permno.10423	permno.10453
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12
## 7	12	12	12	12	12
## 8	12	12	12	12	12
## 9	12	12	12	12	12
## 10	12	12	12	12	12
## 11	12	12	12	12	12
##	permno.10463	permno.10467	permno.10484	permno.10501	permno.10507
## 1	12	12	12	12	12
## 2	12	12	12	12	12
## 3	12	12	12	12	12
## 4	12	12	12	12	12
## 5	12	12	12	12	12
## 6	12	12	12	12	12

## 7	12	12	12	12	12	
## 8	12	12	12	12	12	
## 9	12	12	12	12	12	
## 10	12	12	12	12	12	
## 11	12	12	12	12	12	
##	permno.10516	permno.10517	permno.10547	permno.10550	permno.10562	
## 1	12	12	12	12	12	
## 2	12	12	12	12	12	
## 3	12	12	12	12	12	
## 4	12	12	12	12	12	
## 5	12	12	12	12	12	
## 6	12	12	12	12	12	
## 7	12	12	12	12	12	
## 8	12	12	12	12	12	
## 9	12	12	12	12	12	
## 10	12	12	12	12	12	
## 11	12	12	12	12	12	
##	permno.10563	permno.10564	permno.10574	permno.10588	permno.10594	
## 1	12	12	12	12	12	
## 2	12	12	12	12	12	
## 3	12	12	12	12	12	
## 4	12	12	12	12	12	
## 5	12	12	12	12	12	
## 6	12	12	12	12	12	
## 7	12	12	12	12	12	
## 8	12	12	12	12	12	
## 9	12	12	12	12	12	
## 10	12	12	12	12	12	
## 11	12	12	12	12	12	
##	permno.10606	permno.10622	permno.10623	permno.10628	permno.10629	
## 1	12	12	12	12	12	
## 2	12	12	12	12	12	
## 3	12	12	12	12	12	
## 4	12	12	12	12	12	
## 5	12	12	12	12	12	
## 6	12	12	12	12	12	
## 7	12	12	12	12	12	
## 8	12	12	12	12	12	
## 9	12	12	12	12	12	
## 10	12	12	12	12	12	
## 11	12	12	12	12	12	
##	permno.10638	permno.10644	permno.10645	permno.10656	permno.(Other)	year.Min.
## 1	12	12	12	12	53911	2000
## 2	12	12	12	12	49672	2001
## 3	12	12	12	12	45674	2002
## 4	12	12	12	12	42583	2003
## 5	12	12	12	12	41804	2004
## 6	12	12	12	12	42084	2005
## 7	12	12	12	12	41871	2006
## 8	12	12	12	12	41753	2007
## 9	12	12	12	12	40823	2008
## 10	12	12	12	12	38012	2009
## 11	12	12	12	12	36327	2010
##	year.1st Qu.	year.Median	year.Mean	year.3rd Qu.	year.Max.	ret.Min.

## 1	2000	2000	2000	2000	2000	-0.9615384340		
## 2	2001	2001	2001	2001	2001	-0.9812949896		
## 3	2002	2002	2002	2002	2002	-0.9136441946		
## 4	2003	2003	2003	2003	2003	-0.9418604374		
## 5	2004	2004	2004	2004	2004	-0.9187499881		
## 6	2005	2005	2005	2005	2005	-0.7966101766		
## 7	2006	2006	2006	2006	2006	-0.8791540861		
## 8	2007	2007	2007	2007	2007	-0.9526842237		
## 9	2008	2008	2008	2008	2008	-0.9641255736		
## 10	2009	2009	2009	2009	2009	-0.9111764431		
## 11	2010	2010	2010	2010	2010	-0.8269895911		
##	ret.1st Qu.	ret.Median	ret.Mean	ret.3rd Qu.	ret.Max.			
## 1	-0.1319444478	-0.0121457493	-0.0100748903	0.0762799233	5.6029410362			
## 2	-0.0912109297	0.0041152174	0.0246395190	0.0951333176	8.6666669846			
## 3	-0.0986666381	-0.0070422466	-0.0082690645	0.0588293783	5.6399998665			
## 4	-0.02977778100	0.0220364723	0.0483819156	0.0924368799	4.6250000000			
## 5	-0.0434782617	0.0107526523	0.0166063852	0.0633954648	3.7428572178			
## 6	-0.0512122111	0.0006675515	0.0039563130	0.0502267452	3.3032786846			
## 7	-0.0381936785	0.0082634622	0.0139527424	0.0573521405	4.0229883194			
## 8	-0.0571625941	-0.0044943779	-0.0050265961	0.0421818122	5.8010077477			
## 9	-0.1347608566	-0.0310461428	-0.0431089926	0.0432299748	4.8999996185			
## 10	-0.0542604206	0.0251198402	0.0481144160	0.1123742256	15.7741928101			
## 11	-0.0464502219	0.0150435138	0.0233956206	0.0778974518	6.1071429253			
##	city.Min.	city.1st Qu.	city.Median	city.Mean	city.3rd Qu.	city.Max.		
## 1	1.00000	5.00000	11.00000	10.44184	15.00000	20.00000		
## 2	1.00000	5.00000	11.00000	10.43189	15.00000	20.00000		
## 3	1.00000	5.00000	11.00000	10.42132	15.00000	20.00000		
## 4	1.00000	5.00000	11.00000	10.35514	15.00000	20.00000		
## 5	1.00000	5.00000	11.00000	10.33350	15.00000	20.00000		
## 6	1.00000	5.00000	11.00000	10.34364	15.00000	20.00000		
## 7	1.00000	5.00000	12.00000	10.34998	15.00000	20.00000		
## 8	1.00000	5.00000	12.00000	10.36830	15.00000	20.00000		
## 9	1.00000	5.00000	12.00000	10.40040	15.00000	20.00000		
## 10	1.00000	5.00000	11.00000	10.33793	15.00000	20.00000		
## 11	1.00000	5.00000	11.00000	10.28741	15.00000	20.00000		
##	ind.Min.	ind.1st Qu.	ind.Median	ind.Mean	ind.3rd Qu.	ind.Max.	date.480	
## 1	1.000000	6.000000	9.000000	8.102216	11.000000	12.000000	4642	
## 2	1.000000	6.000000	9.000000	8.112328	11.000000	12.000000	4455	
## 3	1.000000	6.000000	9.000000	8.170074	11.000000	12.000000	4032	
## 4	1.000000	6.000000	10.000000	8.264376	11.000000	12.000000	3776	
## 5	1.000000	6.000000	10.000000	8.321316	11.000000	12.000000	3596	
## 6	1.000000	6.000000	10.000000	8.400166	11.000000	12.000000	3618	
## 7	1.000000	6.000000	10.000000	8.414710	11.000000	12.000000	3598	
## 8	1.000000	6.000000	10.000000	8.444424	11.000000	12.000000	3603	
## 9	1.000000	6.000000	10.000000	8.472424	11.000000	12.000000	3570	
## 10	1.000000	6.000000	10.000000	8.461531	11.000000	12.000000	3373	
## 11	1.000000	6.000000	10.000000	8.429695	11.000000	12.000000	3172	
##	date.484	date.481	date.483	date.482	date.488	date.486	date.485	date.487
## 1	4621	4610	4609	4605	4600	4598	4592	4590
## 2	4426	4384	4337	4282	4240	4213	4155	4130
## 3	3998	3972	3948	3934	3919	3894	3864	3853
## 4	3754	3710	3684	3657	3639	3620	3600	3594
## 5	3593	3592	3586	3585	3581	3581	3580	3577
## 6	3612	3610	3608	3607	3607	3606	3606	3605

## 7	3595	3590	3590	3590	3590	3589	3587	3586
## 8	3597	3591	3588	3583	3579	3577	3574	3569
## 9	3565	3557	3542	3522	3521	3502	3485	3473
## 10	3352	3329	3302	3284	3258	3248	3237	3225
## 11	3157	3151	3142	3140	3138	3125	3119	3112
##	date.489	date.490	date.491	date.492	date.493	date.494	date.495	date.496
## 1	4588	4545	4499	0	0	0	0	0
## 2	4112	4070	4056	0	0	0	0	0
## 3	3824	3820	3804	0	0	0	0	0
## 4	3585	3577	3575	0	0	0	0	0
## 5	3575	3574	3572	0	0	0	0	0
## 6	3603	3597	3593	0	0	0	0	0
## 7	3582	3581	3581	0	0	0	0	0
## 8	3563	3562	3555	0	0	0	0	0
## 9	3454	3423	3397	0	0	0	0	0
## 10	3209	3198	3185	0	0	0	0	0
## 11	3099	3087	3073	0	0	0	0	0
##	date.497	date.498	date.499	date.500	date.501	date.502	date.503	date.504
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.505	date.506	date.507	date.508	date.509	date.510	date.511	date.512
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.513	date.514	date.515	date.516	date.517	date.518	date.519	date.520
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.521	date.522	date.523	date.524	date.525	date.526	date.527	date.528

## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.529	date.530	date.531	date.532	date.533	date.534	date.535	date.536
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.537	date.538	date.539	date.540	date.541	date.542	date.543	date.544
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.545	date.546	date.547	date.548	date.549	date.550	date.551	date.552
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.553	date.554	date.555	date.556	date.557	date.558	date.559	date.560
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0

## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.561	date.562	date.563	date.564	date.565	date.566	date.567	date.568
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.569	date.570	date.571	date.572	date.573	date.574	date.575	date.576
## 1	0	0	0	0	0	0	0	0
## 2	0	0	0	0	0	0	0	0
## 3	0	0	0	0	0	0	0	0
## 4	0	0	0	0	0	0	0	0
## 5	0	0	0	0	0	0	0	0
## 6	0	0	0	0	0	0	0	0
## 7	0	0	0	0	0	0	0	0
## 8	0	0	0	0	0	0	0	0
## 9	0	0	0	0	0	0	0	0
## 10	0	0	0	0	0	0	0	0
## 11	0	0	0	0	0	0	0	0
##	date.577	date.578	date.(Other)	city_returns.Min.	city_returns.1st Qu.			
## 1	0	0	0	-0.270534486	-0.069581896			
## 2	0	0	0	-0.230681196	-0.052691292			
## 3	0	0	0	-0.201631397	-0.060558319			
## 4	0	0	0	-0.057042830	0.016218796			
## 5	0	0	0	-0.149902940	-0.007337616			
## 6	0	0	0	-0.123852029	-0.028083174			
## 7	0	0	0	-0.118696563	-0.005877608			
## 8	0	0	0	-0.157961637	-0.019134125			
## 9	0	0	0	-0.299503535	-0.102640361			
## 10	0	0	0	-0.171920344	0.001067523			
## 11	0	0	0	-0.123967253	-0.027859902			
##	city_returns.Median	city_returns.Mean	city_returns.3rd Qu.	city_returns.Max.				
## 1	-0.021834882	-0.009747969	0.055983528	0.344486326				
## 2	0.029200688	0.024860797	0.079776831	0.422849029				
## 3	-0.012173467	-0.008786427	0.032210965	0.315815747				
## 4	0.043276556	0.049337457	0.077847213	0.272160679				
## 5	0.015159218	0.017589902	0.050050143	0.187615231				
## 6	0.002705623	0.003818538	0.032051869	0.115589522				
## 7	0.014673841	0.014150576	0.035127830	0.179933280				
## 8	0.001114333	-0.004535199	0.018248394	0.094374500				
## 9	-0.027299616	-0.043640268	0.022113474	0.150761425				
## 10	0.063041970	0.050115572	0.092088118	0.483421147				
## 11	0.038584746	0.024098869	0.071188457	0.200951740				
##	MRP.Min.	MRP.1st Qu.	MRP.Median	MRP.Mean	MRP.3rd Qu.			

## 1	-0.107200000	-0.047400000	-0.027600000	-0.013748502	0.046400000
## 2	-0.100500000	-0.072600000	-0.019400000	-0.011749072	0.031300000
## 3	-0.103500000	-0.057600000	-0.014400000	-0.019482284	0.005000000
## 4	-0.025700000	-0.012400000	0.014200000	0.022684816	0.042900000
## 5	-0.040600000	0.000800000	0.014300000	0.008737202	0.021500000
## 6	-0.027600000	-0.019700000	0.004900000	0.002760376	0.036100000
## 7	-0.035700000	-0.003000000	0.014600000	0.008264604	0.020300000
## 8	-0.048300000	-0.019600000	0.009200000	0.001176638	0.032200000
## 9	-0.172300000	-0.078600000	-0.030900000	-0.036624601	0.015300000
## 10	-0.101000000	-0.025900000	0.033300000	0.022435878	0.077200000
## 11	-0.078900000	-0.047700000	0.020000000	0.014747901	0.063100000
##	MRP.Max.	HML.Min.	HML.1st Qu.	HML.Median	HML.Mean
## 1	0.070300000	-0.126100000	-0.012300000	0.048100000	0.0293780268
## 2	0.079400000	-0.070600000	-0.043600000	0.017800000	0.0134248230
## 3	0.078400000	-0.064600000	0.011500000	0.022900000	0.0107105373
## 4	0.082200000	-0.020600000	-0.014400000	0.000900000	0.0027210893
## 5	0.045400000	-0.016700000	-0.003600000	0.003900000	0.0071065012
## 6	0.039200000	-0.017900000	-0.006900000	0.011300000	0.0070781267
## 7	0.032300000	-0.017600000	-0.001000000	0.011900000	0.0102491883
## 8	0.034900000	-0.029900000	-0.021400000	-0.010700000	-0.0101178827
## 9	0.046000000	-0.049300000	-0.010600000	-0.000300000	0.0020498893
## 10	0.101900000	-0.098600000	-0.043700000	0.005600000	-0.0006602551
## 11	0.095400000	-0.042800000	-0.023500000	0.000900000	-0.0013699960
##	HML.3rd Qu.	HML.Max.	RF.Min.	RF.1st Qu.	RF.Median
## 1	0.076500000	0.123200000	4.000000e-03	4.300000e-03	4.800000e-03
## 2	0.056400000	0.138800000	1.500000e-03	2.800000e-03	3.100000e-03
## 3	0.038700000	0.042300000	1.100000e-03	1.300000e-03	1.400000e-03
## 4	0.014600000	0.027100000	7.000000e-04	7.000000e-04	9.000000e-04
## 5	0.016700000	0.045200000	6.000000e-04	8.000000e-04	1.000000e-03
## 6	0.017500000	0.028500000	1.600000e-03	2.100000e-03	2.400000e-03
## 7	0.025200000	0.032700000	3.400000e-03	3.600000e-03	4.000000e-03
## 8	-0.000200000	0.003900000	2.700000e-03	3.400000e-03	4.000000e-03
## 9	0.030700000	0.044100000	0.000000e+00	1.300000e-03	1.500000e-03
## 10	0.026000000	0.075700000	0.000000e+00	0.000000e+00	1.000000e-04
## 11	0.020500000	0.033600000	0.000000e+00	1.000000e-04	1.000000e-04
##	RF.Mean	RF.3rd Qu.	RF.Max.	MOM.Min.	MOM.1st Qu.
## 1	4.773604e-03	5.000000e-03	5.600000e-03	-0.0906000000	-0.0680000000
## 2	3.163787e-03	3.900000e-03	5.400000e-03	-0.2501000000	-0.0812000000
## 3	1.350708e-03	1.400000e-03	1.500000e-03	-0.1628000000	0.0175000000
## 4	8.517397e-04	1.000000e-03	1.000000e-03	-0.1076000000	-0.0102000000
## 5	9.838900e-04	1.100000e-03	1.600000e-03	-0.0536000000	-0.0150000000
## 6	2.450143e-03	2.900000e-03	3.200000e-03	-0.0131000000	0.0004000000
## 7	3.924729e-03	4.100000e-03	4.300000e-03	-0.0370000000	-0.0182000000
## 8	3.809546e-03	4.200000e-03	4.400000e-03	-0.0133000000	-0.0016000000
## 9	1.324948e-03	1.800000e-03	2.100000e-03	-0.0781000000	-0.0405000000
## 10	7.516582e-05	1.000000e-04	2.000000e-04	-0.3458000000	-0.1153000000
## 11	8.312941e-05	1.000000e-04	1.000000e-04	-0.0536000000	-0.0298000000
##	MOM.Median	MOM.Mean	MOM.3rd Qu.	MOM.Max.	indret.Min.
## 1	-0.001200000	0.0163667526	0.0570000000	0.1838000000	-0.231696367
## 2	0.021700000	-0.0038333641	0.0838000000	0.1248000000	-0.199510977
## 3	0.037600000	0.0239051321	0.0679000000	0.0962000000	-0.198041797
## 4	-0.001900000	-0.0151266249	0.0152000000	0.0375000000	-0.065995745
## 5	0.002000000	0.0003643864	0.0259000000	0.0526000000	-0.125920936
## 6	0.005400000	0.0117245031	0.0246000000	0.0350000000	-0.093874954

```
## 7 -0.0026000000 -0.0053262152 0.0124000000 0.0276000000 -0.115569755
## 8 0.0030000000 0.0175373489 0.0279000000 0.0654000000 -0.154650033
## 9 0.0322000000 0.0157833353 0.0611000000 0.1245000000 -0.309580952
## 10 -0.0500000000 -0.0540570281 0.0264000000 0.0531000000 -0.205032170
## 11 0.0136000000 0.0047365454 0.0321000000 0.0368000000 -0.119565889
## indret.1st Qu. indret.Median indret.Mean indret.3rd Qu. indret.Max.
## 1 -0.052548133 -0.007632487 -0.000873050 0.028119655 0.416101128
## 2 -0.032575607 0.023592338 0.025786685 0.071124643 0.480341494
## 3 -0.047627989 0.005931754 -0.005631135 0.035642218 0.285326600
## 4 0.014074927 0.043373980 0.046605819 0.066260874 0.298667014
## 5 -0.004751768 0.018252801 0.018300291 0.040691782 0.260545909
## 6 -0.019288585 0.001436898 0.004841844 0.029061180 0.173220903
## 7 -0.001106735 0.015605695 0.013783091 0.030768951 0.195191547
## 8 -0.028508164 -0.001335449 -0.006057680 0.014753194 0.104822814
## 9 -0.097998999 -0.014430972 -0.040644081 0.010713807 0.232047364
## 10 -0.012664497 0.049110822 0.041098763 0.096727975 0.390912533
## 11 -0.033044849 0.038807660 0.021072395 0.067907922 0.180234089
```

Answer: The panel is unbalanced.

#4c. ESTIMATE PANEL REGRESSION

```
local_returns_estimation <- lm(ret ~ city_returns + indret, data = local_returns_panel)
summary(local_returns_estimation)
```

```
##
## Call:
## lm(formula = ret ~ city_returns + indret, data = local_returns_panel)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.1930 -0.0739 -0.0067  0.0577 15.6429
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.0014572  0.0002687  -5.424 5.83e-08 ***
## city_returns  0.2121668  0.0051632  41.092 < 2e-16 ***
## indret        0.8558370  0.0049870 171.613 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1857 on 487579 degrees of freedom
## Multiple R-squared:  0.1502, Adjusted R-squared:  0.1502
## F-statistic: 4.31e+04 on 2 and 487579 DF, p-value: < 2.2e-16
```

#4d. ESTIMATE PANEL REGRESSION USING POOLED OLS WITHOUT DUMMIES

```
local_returns_estimation_pooled <- plm(ret ~ city_returns + indret, data = local_returns_panel, model =
summary(local_returns_estimation_pooled)
```

```
## Pooling Model
##
## Call:
## plm(formula = ret ~ city_returns + indret, data = local_returns_panel,
##      model = "pooling")
##
## Unbalanced Panel: n = 6783, T = 1-132, N = 487582
```

```
##
## Residuals:
##      Min.      1st Qu.      Median      3rd Qu.      Max.
## -1.1929686 -0.0739028 -0.0067361  0.0576988 15.6428882
##
## Coefficients:
##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept) -0.00145716  0.00026866  -5.4239 5.834e-08 ***
## city_returns  0.21216676  0.00516315  41.0925 < 2.2e-16 ***
## indret        0.85583696  0.00498702 171.6127 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Total Sum of Squares:    19778
## Residual Sum of Squares: 16807
## R-Squared:    0.15022
## Adj. R-Squared: 0.15022
## F-statistic: 43095.7 on 2 and 487579 DF, p-value: < 2.22e-16
```

#4e. TEST FOR HETEROSKEDASTICITY IN THE RESIDUALS

```
bptest(local_returns_estimation_pooled, data = local_returns_panel, studentize = F)
```

```
##
## Breusch-Pagan test
##
## data: local_returns_estimation_pooled
## BP = 85052, df = 2, p-value < 2.2e-16
```

Answer: The p-value is less than 5%. Hence, at 5% significance level we reject the null hypothesis and conclude that there is heteroskedasticity in the residuals.

#4g. ESTIMATE PANEL REGRESSION USING POOLED OLS WITH TIME AND FIRM FIXED EFFECTS. COMPARE RESULTS

```
library(Matrix)
library(lfe)
```

```
##
## Attaching package: 'lfe'
##
## The following object is masked from 'package:plm':
##
##      sargan
##
## The following object is masked from 'package:lmtest':
##
##      waldtest
```

```
local_returns_estimation_pooled_results_bothtimeandfirm <- felm(ret ~ city_returns + indret | permno + year
summary(local_returns_estimation_pooled_results_bothtimeandfirm)
```

```
##
## Call:
## felm(formula = ret ~ city_returns + indret | permno + year | 0 | 0, data = local_returns_panel)
##
## Residuals:
##      Min      1Q  Median      3Q      Max
## -1.3031 -0.0743 -0.0075  0.0566 14.9957
```

```
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## city_returns 0.211815   0.005294  40.01  <2e-16 ***
## indret       0.854208   0.005032 169.75  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1852 on 480787 degrees of freedom
## Multiple R-squared(full model): 0.166   Adjusted R-squared: 0.1542
## Multiple R-squared(proj model): 0.138   Adjusted R-squared: 0.1258
## F-statistic(full model):14.09 on 6794 and 480787 DF, p-value: < 2.2e-16
## F-statistic(proj model): 3.849e+04 on 2 and 480787 DF, p-value: < 2.2e-16

#4h. ESTIMATE PANEL REGRESSION USING POOLED OLS WITH TIME AND FIRM FIXED
EFFECTS, AND WITH DOUBLE CLUSTERING ON BOTH FIRM AND TIME DIMENSIONS

local_returns_estimation_pooled_results_bothtimeandfirm_clust <-felm(ret ~ city_returns + indret | permno + year | permno + year, data = local_returns_df)
summary(local_returns_estimation_pooled_results_bothtimeandfirm_clust)

##
## Call:
##      felm(formula = ret ~ city_returns + indret | permno + year | permno + year, data = local_returns_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.3031 -0.0743 -0.0075  0.0566 14.9957
##
## Coefficients:
##           Estimate Cluster s.e. t value Pr(>|t|)
## city_returns 0.21182      0.02087   10.15 1.39e-06 ***
## indret       0.85421      0.05159   16.56 1.35e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1852 on 480787 degrees of freedom
## Multiple R-squared(full model): 0.166   Adjusted R-squared: 0.1542
## Multiple R-squared(proj model): 0.138   Adjusted R-squared: 0.1258
## F-statistic(full model, *iid*):14.09 on 6794 and 480787 DF, p-value: < 2.2e-16
## F-statistic(proj model): 785.5 on 2 and 10 DF, p-value: 1.012e-11

#4i. ESTIMATE PANEL REGRESSION USING FIXED EFFECT (FE) ESTIMATOR

local_returns_estimation_firmfe <-felm(ret ~ city_returns + indret | permno | 0 | 0, data=local_returns_df)
summary(local_returns_estimation_firmfe)

##
## Call:
##      felm(formula = ret ~ city_returns + indret | permno | 0 | 0, data = local_returns_df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1.3028 -0.0744 -0.0076  0.0566 14.9962
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
```

```
## city_returns 0.211136 0.005184 40.73 <2e-16 ***
## indret 0.854034 0.005014 170.34 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.1852 on 480797 degrees of freedom
## Multiple R-squared(full model): 0.166 Adjusted R-squared: 0.1542
## Multiple R-squared(proj model): 0.1511 Adjusted R-squared: 0.1391
## F-statistic(full model): 14.1 on 6784 and 480797 DF, p-value: < 2.2e-16
## F-statistic(proj model): 4.278e+04 on 2 and 480797 DF, p-value: < 2.2e-16

#4j. ESTIMATE PANEL REGRESSION USING FIRST DIFFERENCE (FD) ESTIMATOR
local_returns_estimation_firstdiff <- plm(ret ~ city_returns + indret, data=local_returns_df, model = "fd")

## Warning in pdata.frame(data, index): duplicate couples (id-time) in resulting pdata.frame
## to find out which, use, e.g., table(index(your_pdataframe), useNA = "ifany")
coef(summary(local_returns_estimation_firstdiff))

##              Estimate Std. Error t-value Pr(>|t|)
## (Intercept) -0.0002495018 0.0003832596 -0.6509997 0.515047
## city_returns 0.2173462840 0.0055628022 39.0713663 0.000000
## indret 0.8473646279 0.0052652998 160.9337866 0.000000
```