

FINANCIAL ECONOMETRICS- R_LAB

NGUYEN NGOC PHUNG

5/6/2021

Contents

REGRESSION MODEL AND TESTINGS	1
UKHP DATASET	2
SANDPHEDGE DATASET	3
MACRO DATASET	8
ARIMA MODELS	10
IMPORT LIBRARIES	10
IMPORT AND VIEW THE DATASET	11
EXTRACT THE HOUSE PRICE TIME SERIES AND ESTIMATE THE SIMPLE RETURNS . .	11
PLOTING OF HP AND DHP	11
DESCRIPTIVE STATISTICS OF RETURNS	11
ACF AND PACF PLOTS OF HOUSE PRICE (HP)	12
ACF AND PACF PLOTS OF SIMPLE RETURNS (DHP)	13
AUGMENTED DICKEY-FULLER TEST	14
FIND THE COEFFICIENT MATRIX OF ARIMA MODEL BY <code>coefest()</code> FUNCTION	14
FIND AIC (INFORMATION CRITERIA) OF ARIMA(1,0,2) AND ARIMA(1,0,1)	15
FORECASTING IN TIME SERIES	15
FINDING THE BEST ARIMA MODEL ACCORDING TO AIC, BIC VALUE	16
LJUNG-BOX TEST TO CHECK IF THE RESIDUALS SERIES IS WHITE NOISE PROCESS .	17

REGRESSION MODEL AND TESTINGS

IMPORT THE LIBRARIES

```
library(psych)
library(readxl)
library(car)
library(tseries)
```

```
library(lmtest)
library(MASS)
library(foreign)
library(sandwich)
```

UKHP DATASET

UKHP DATASET AND RETURNS

```
HP <- read_excel("UKHP.xls")
```

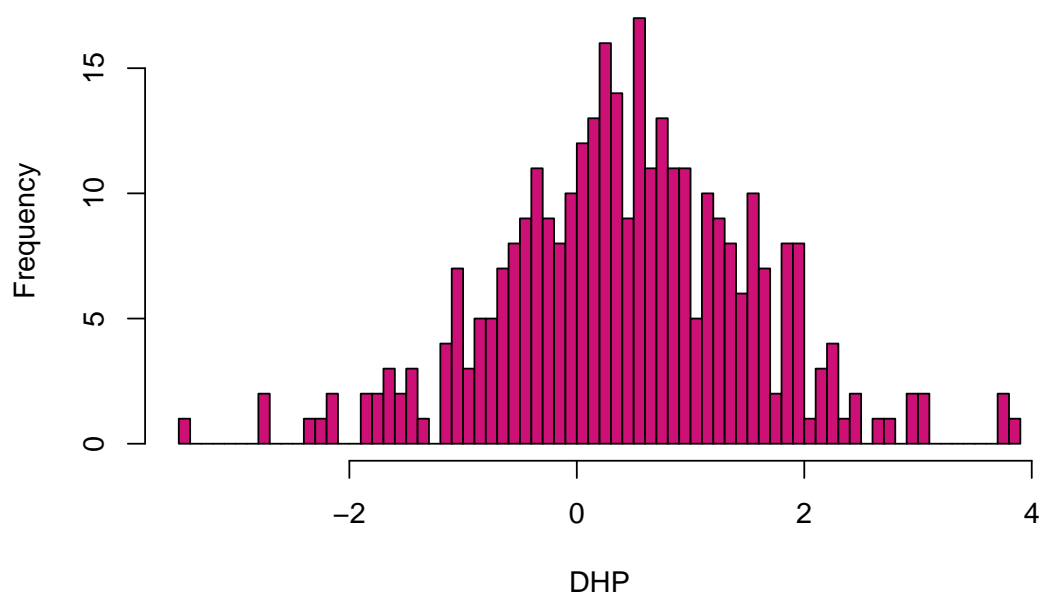
```
Z<-ts(HP$`Average House Price`)
DHP<-100*diff(Z)/lag(Z,-1)
DHP[1:20]
```

```
## [1] 0.8389504256 -1.1289220193 1.4833261822 1.3195330084 1.3269075933
## [6] -1.0275463501 -0.9174932632 -1.4456792164 0.3889743215 -0.8390243074
## [11] 0.1623230805 -2.1709901200 -0.1944398560 0.2456052867 0.0005801514
## [16] 1.4136151132 0.3660040378 -0.3377067136 -0.7873699395 -2.3221311513
```

HISTOGRAM OF RETURNS TIME SERIES (DHP)

```
hist(DHP, breaks = 70, col='deeppink3', main='HISTOGRAM OF SIMPLE RETURN (DHP)')
```

HISTOGRAM OF SIMPLE RETURN (DHP)



DESCRIPTIVE STATISTICS OF RETURNS (DHP)

```
summary(DHP)
```

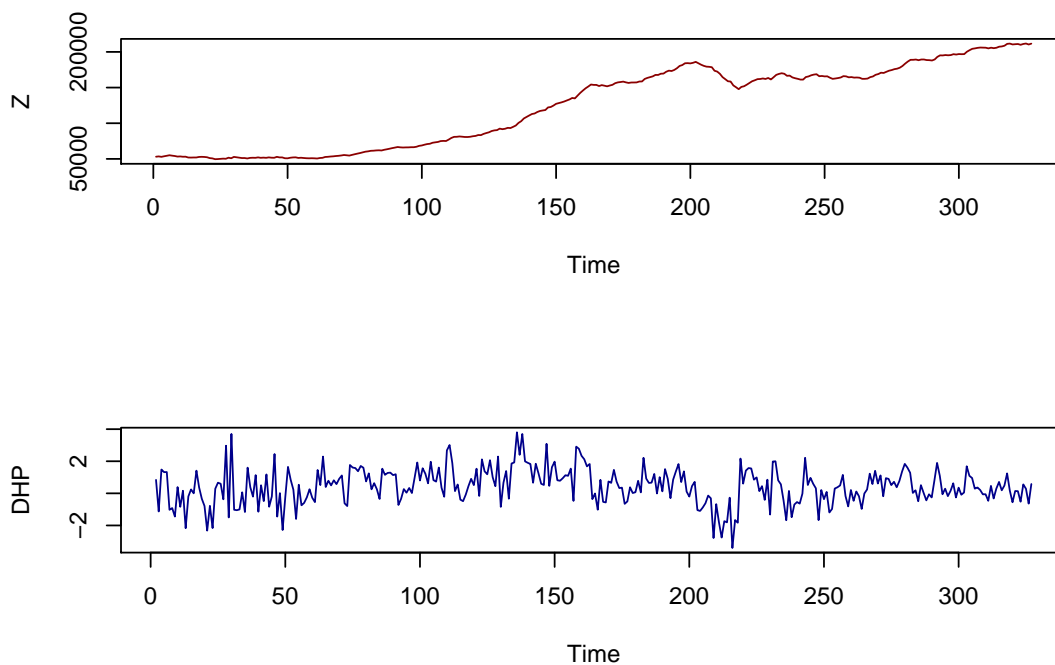
```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -3.4047 -0.2557   0.4483   0.4315   1.1522   3.8022
```

```
describe(DHP)
```

```
##      vars    n mean    sd median trimmed  mad   min max range  skew kurtosis   se
## X1      1 326 0.43 1.12   0.45   0.44 1.04 -3.4 3.8   7.21 -0.08    0.59 0.06
```

PLOTTING OF HOUSE PRICE AND RETURNS TIME SERIES

```
layout(matrix(c(1,2,1,2),2,2))
plot(Z,type='l',col='darkred')
plot(DHP,type='l',col='darkblue')
```



SANDPHEDGE DATASET

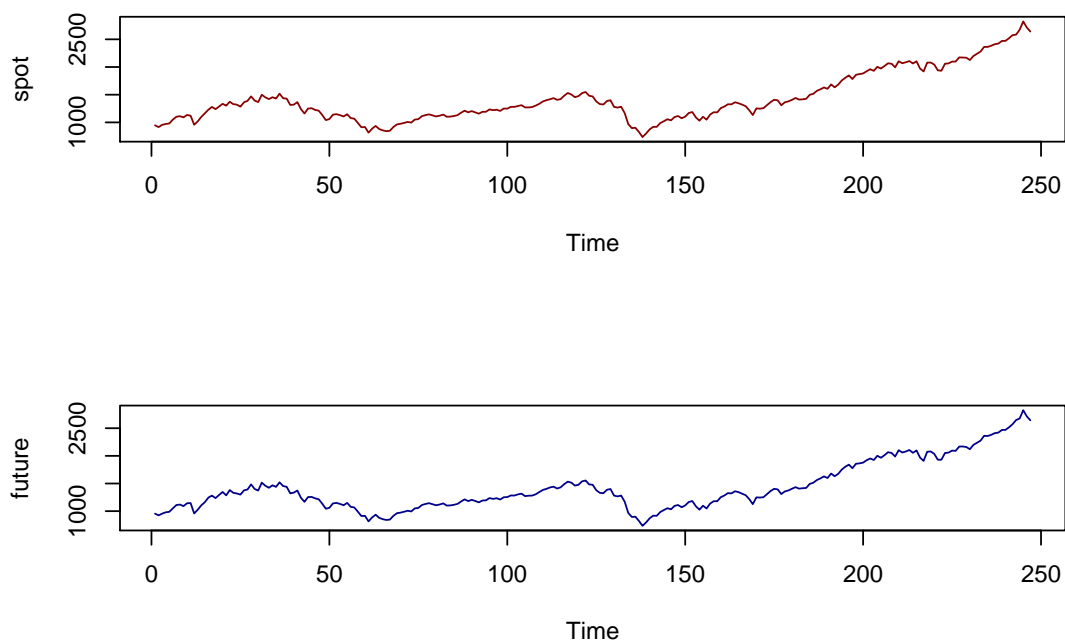
SANDPHEDGE DATASET AND RETURNS

```
# The dataset include the spot and future price @-
SandPhedge <- read_excel('SandPhedge.xls')
```

```
spot<-SandPhedge$Spot ; future<- SandPhedge$Futures
```

```
spot<-ts(spot) ; future<-ts(future)
```

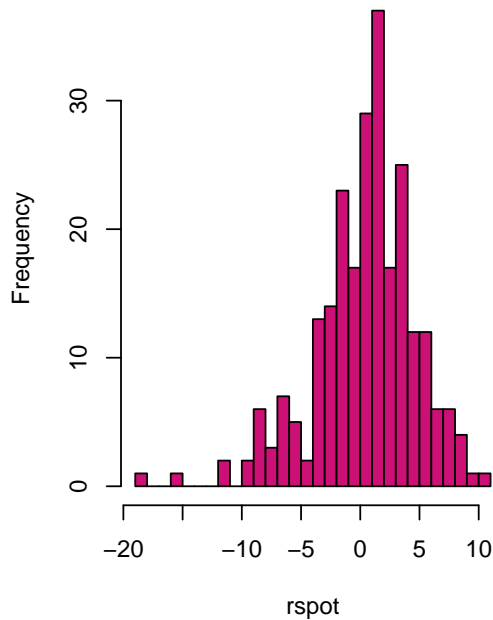
```
# Plotting of spot and future price 8-*
layout(matrix(c(1,2,1,2),2,2))
plot(spot,type='l',col='darkred')
plot(future,type='l',col='darkblue')
```



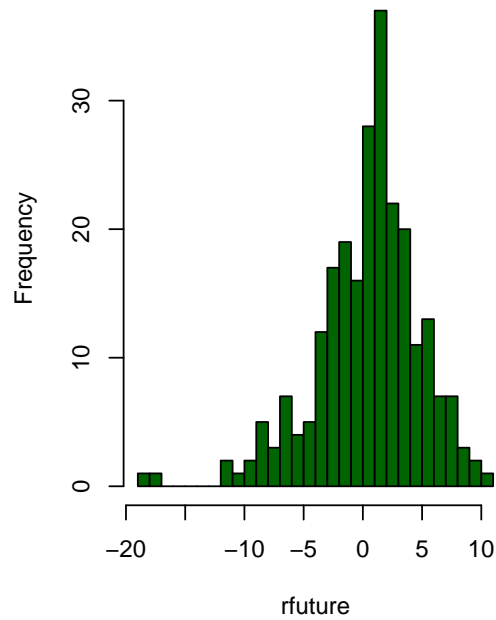
```
# Now, we calculate the log-returns of both spot and future prices @-
rspot <- 100*diff(log(spot))
rfuture <- 100*diff(log(future))
```

```
# Histogram of both returns time series
layout(matrix(c(1,1,2,2),2,2))
hist(rspot,breaks=30, col='deeppink3',main='RETURNS OF SPOT PRICE')
hist(rfuture,breaks=30, col='darkgreen',main='RETURNS OF FUTURE PRICE')
```

RETURNS OF SPOT PRICE



RETURNS OF FUTURE PRICE



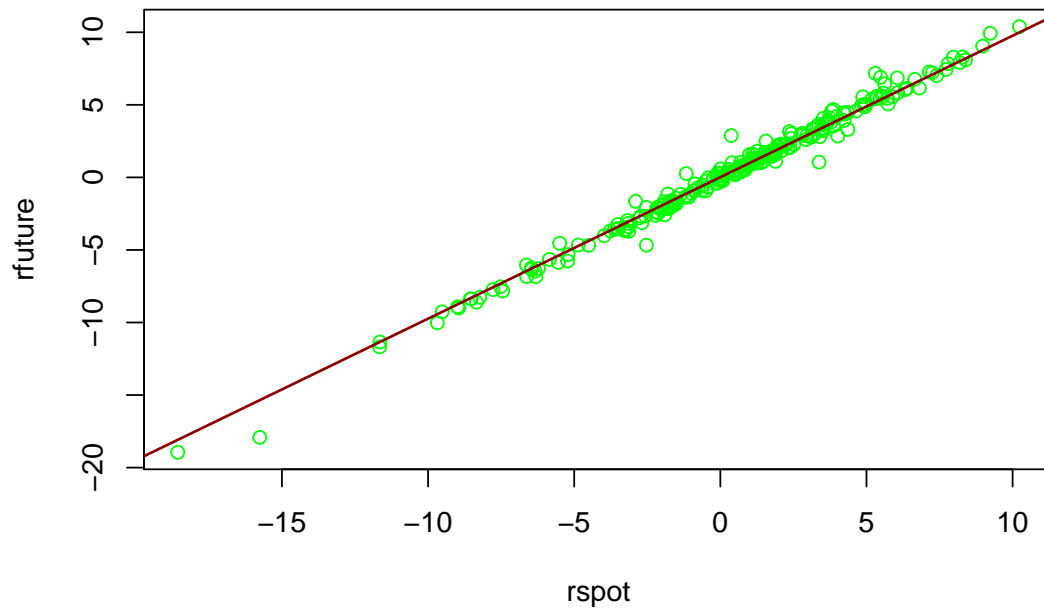
SIMPLE REGRESSION MODEL

We concern about whether the return of future price explain the changes of return in spot price or no

```
ReturnReg<-lm(rspot~rfuture)
summary(ReturnReg)
```

```
##
## Call:
## lm(formula = rspot ~ rfuture)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.45284 -0.16401  0.00236  0.23692  2.33789
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.013077   0.029473   0.444   0.658
## rfuture      0.975077   0.006654 146.543 <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.4602 on 244 degrees of freedom
## Multiple R-squared:  0.9888, Adjusted R-squared:  0.9887
## F-statistic: 2.147e+04 on 1 and 244 DF, p-value: < 2.2e-16
```

```
# Plotting of regression model @-@
plot(rspot, rfuture,col='green')
abline(ReturnReg,col='darkred',lwd=1.5)
```



GOODNESS-OF-FIT STATISTIC - DEVIANCE AND LOG-LIKELIHOOD

```
# in order to assess how good our model fit the actual values, we calculate the deviance (SSR)
deviance(ReturnReg)
```

```
## [1] 51.68444
```

```
logLik(ReturnReg) # return the logLik objects s.a the degree of freedom
```

```
## 'log Lik.' -157.1574 (df=3)
```

BREUSCH-PAGAN TEST

```
bptest(ReturnReg)
```

```
##
## studentized Breusch-Pagan test
##
## data: ReturnReg
## BP = 0.070052, df = 1, p-value = 0.7913
```

BREUSCH-GODFREY TEST

```
bgtest(ReturnReg)
```

```
##  
## Breusch-Godfrey test for serial correlation of order up to 1  
##  
## data: ReturnReg  
## LM test = 59.696, df = 1, p-value = 1.107e-14
```

DURBIN-WATSON TEST FOR AUTO-CORRELATED ERRORS

```
dwtest(ReturnReg)
```

```
##  
## Durbin-Watson test  
##  
## data: ReturnReg  
## DW = 2.9694, p-value = 1  
## alternative hypothesis: true autocorrelation is greater than 0
```

```
durbinWatsonTest(ReturnReg,max.lag=2)
```

```
## lag Autocorrelation D-W Statistic p-value  
## 1 -0.4860789 2.969363 0.000  
## 2 -0.2293546 2.431056 0.002  
## Alternative hypothesis: rho[lag] != 0
```

NON-CONSTANT VARIANCE SCORE TEST

```
ncvTest(ReturnReg)
```

```
## Non-constant Variance Score Test  
## Variance formula: ~ fitted.values  
## Chisquare = 0.3534588, Df = 1, p = 0.55216
```

AUGMENTED DICKEY-FULLER TEST FOR RETURNS (SPOT & FUTURE) AND PRICE (SPOT & FUTURE)

```
# returns  
adf.test(na.omit(rspot))
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: na.omit(rspot)  
## Dickey-Fuller = -5.6811, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
adf.test(na.omit(rfuture))
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: na.omit(rfuture)  
## Dickey-Fuller = -5.7291, Lag order = 6, p-value = 0.01  
## alternative hypothesis: stationary
```

```
# prices  
adf.test(spot)
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: spot  
## Dickey-Fuller = -0.88531, Lag order = 6, p-value = 0.953  
## alternative hypothesis: stationary
```

```
adf.test(future)
```

```
##  
## Augmented Dickey-Fuller Test  
##  
## data: future  
## Dickey-Fuller = -0.9341, Lag order = 6, p-value = 0.9476  
## alternative hypothesis: stationary
```

MACRO DATASET

MACRO DATASET AND RETURNS

```
macro <- read_excel('macro.xls')
```

```
spread<-ts(macro$BMINUSA)  
credit<-ts(macro$CCREDIT)  
prod<-ts(macro$INDPRO)  
msoft<-ts(macro$MICROSOFT)  
sandp<-ts(macro$SANDP)  
money<-ts(macro$M1SUPPLY)  
cpi<-ts(macro$CPI)
```

```
dsread<-diff(spread)  
dcredit<-diff(credit)  
dprod<-diff(prod)  
dmoney<-diff(money)
```

```
rmsoft<-100*diff(log(msoft))  
rsandp<-100*diff(log(sandp))
```



```
term<-ts(macro$USTB10Y)-ts(macro$USTB3M)
inflation<-100*diff(log(cpi))
dinflation<-diff(inflation)
mustb3m<-ts(macro$USTB3M)/12
rterm<-diff(term)
ermsoft<-rmsoft - mustb3m
ersandp<-rsandp - mustb3m
```

MULTIPLE REGRESSION MODEL

```
msoftreg <- lm(ermsoft[2:325]~ersandp[2:325] + dprod[2:325] + dcredit[2:325] +
               dinflation[2:325] + dmoney[2:325] + dspread[2:325] + rterm[2:325])
summary(msoftreg)
```

```
##
## Call:
## lm(formula = ermsoft[2:325] ~ ersandp[2:325] + dprod[2:325] +
##     dcredit[2:325] + dinflation[2:325] + dmoney[2:325] + dspread[2:325] +
##     rterm[2:325])
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -36.924  -4.671  -0.384   4.574  24.382
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    1.38426    0.53388   2.593  0.00996 **
## ersandp[2:325]  1.32625    0.10302  12.873 < 2e-16 ***
## dprod[2:325]   -0.91455    0.82171  -1.113  0.26656
## dcredit[2:325] -0.01683    0.03488  -0.482  0.62986
## dinflation[2:325] -3.14294    1.38715  -2.266  0.02414 *
## dmoney[2:325]  -0.01194    0.02317  -0.516  0.60652
## dspread[2:325]  0.72326    4.38207   0.165  0.86901
## rterm[2:325]    4.75868    1.86769   2.548  0.01131 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.202 on 316 degrees of freedom
## Multiple R-squared:  0.3608, Adjusted R-squared:  0.3466
## F-statistic: 25.48 on 7 and 316 DF, p-value: < 2.2e-16
```

NEWBY-WEST TEST FOR HETEROSCEDASTICITY AND AUTOCORRELATION

```
# Simply stated, this function return teh HAC Covariance matrix Estimation
NeweyWest(msoftreg)
```

```
##              (Intercept) ersandp[2:325] dprod[2:325] dcredit[2:325]
## (Intercept)    0.305520096   5.178095e-03 -0.0158598406  -9.926349e-03
```

```
## ersandp[2:325]      0.005178095   1.110974e-02 -0.0150081345  -5.396904e-04
## dprod[2:325]       -0.015859841  -1.500813e-02  0.2711811688  -2.192278e-04
## dcredit[2:325]     -0.009926349  -5.396904e-04 -0.0002192278   7.688322e-04
## dinflation[2:325] -0.072675457  -6.173293e-02  0.1534943327   8.265343e-03
## dmoney[2:325]      -0.001846171   2.103895e-05  0.0019349167   2.355234e-06
## dspread[2:325]     0.317658305   4.140168e-03  0.0055105657  -1.437012e-02
## rterm[2:325]       0.120233907   5.964648e-02 -0.1669457032  -1.309566e-02
##
##      dinflation[2:325] dmoney[2:325] dspread[2:325] rterm[2:325]
## (Intercept)          -0.072675457 -1.846171e-03   0.317658305   0.12023391
## ersandp[2:325]       -0.061732930  2.103895e-05   0.004140168   0.05964648
## dprod[2:325]         0.153494333   1.934917e-03   0.005510566  -0.16694570
## dcredit[2:325]       0.008265343   2.355234e-06  -0.014370121  -0.01309566
## dinflation[2:325]    1.801701889  -2.244382e-03   0.116674663  -0.90465206
## dmoney[2:325]       -0.002244382   3.158284e-04  -0.026330103  -0.01021057
## dspread[2:325]       0.116674663  -2.633010e-02   7.288247914   0.83785431
## rterm[2:325]        -0.904652057  -1.021057e-02   0.837854309   3.03324948
```

```
bwNeweyWest(msoftreg)
```

```
## [1] 7.695043
```

```
NeweyWest(msoftreg, lag = 4, prewhite = FALSE)
```

```
##      (Intercept) ersandp[2:325] dprod[2:325] dcredit[2:325]
## (Intercept)      0.293262795   7.658101e-03 -0.039611730  -8.789173e-03
## ersandp[2:325]   0.007658101   1.142868e-02 -0.012309860  -6.050553e-04
## dprod[2:325]     -0.039611730  -1.230986e-02  0.325312764  -1.624426e-03
## dcredit[2:325]   -0.008789173  -6.050553e-04 -0.001624426   7.790433e-04
## dinflation[2:325] -0.028438008  -5.576858e-02  0.164841144   3.318870e-03
## dmoney[2:325]    -0.001499844  -5.774214e-05  0.002728501  -2.556316e-05
## dspread[2:325]    0.241051820   5.773703e-02  0.019427240  -1.446388e-02
## rterm[2:325]     0.101233416   4.471238e-02 -0.195826416  -1.046251e-02
##
##      dinflation[2:325] dmoney[2:325] dspread[2:325] rterm[2:325]
## (Intercept)          -0.028438008 -1.499844e-03   0.24105182   0.101233416
## ersandp[2:325]       -0.055768584 -5.774214e-05   0.05773703   0.044712382
## dprod[2:325]         0.164841144   2.728501e-03   0.01942724  -0.195826416
## dcredit[2:325]       0.003318870  -2.556316e-05  -0.01446388  -0.010462512
## dinflation[2:325]    1.886506252  -1.597368e-03  -0.12534855  -0.379002137
## dmoney[2:325]       -0.001597368   3.224004e-04  -0.02163564  -0.009477304
## dspread[2:325]       -0.125348548  -2.163564e-02   7.39953192   0.706771660
## rterm[2:325]        -0.379002137  -9.477304e-03   0.70677166   2.946691601
```

ARIMA MODELS

IMPORT LIBRARIES

```
library(psych)
library(lmtest)
library(forecast)
library(readxl)
library(tseries)
```

IMPORT AND VIEW THE DATASET

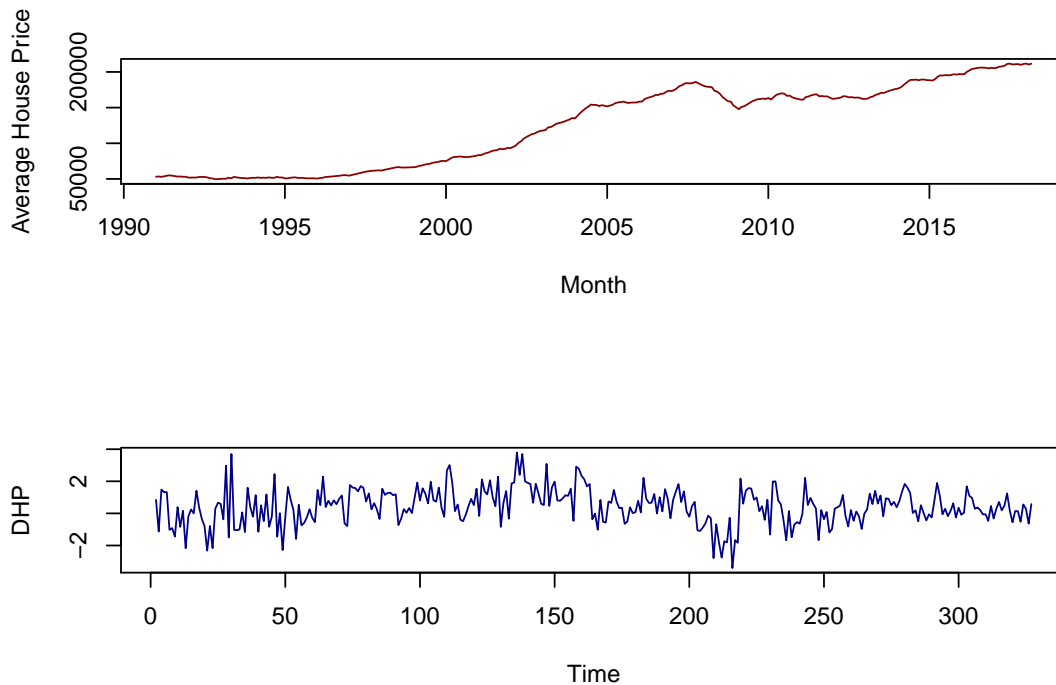
```
HP <- read_excel("UKHP.xls")
```

EXTRACT THE HOUSE PRICE TIME SERIES AND ESTIMATE THE SIMPLE RETURNS

```
Z<-HP$`Average House Price`  
Z<-ts(Z, start=1990, end=2018, freq="quarter")  
DHP<-100*diff(Z)/lag(Z,-1)
```

PLOTTING OF HP AND DHP

```
layout(matrix(c(1,2,1,2),2,2))  
plot(HP,type='l',col='darkred')  
plot(DHP,type='l',col='darkblue')
```



DESCRIPTIVE STATISTICS OF RETURNS

```
summary(DHP)
```

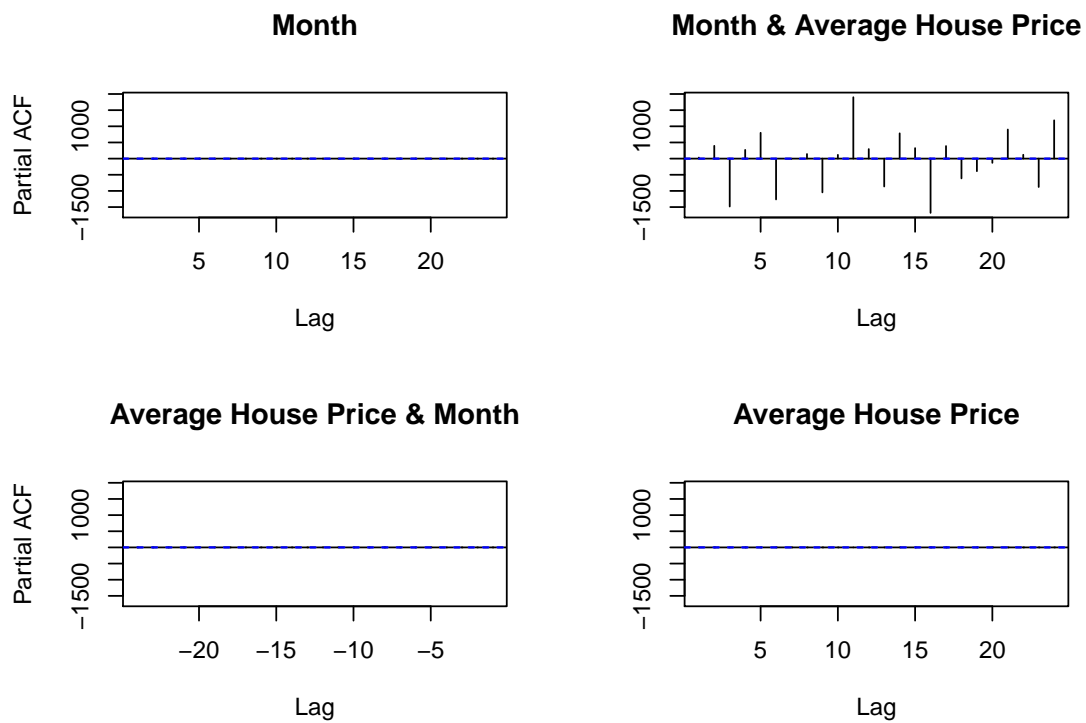
```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## -3.4047 -0.2557  0.4483  0.4315  1.1522  3.8022
```

ACF AND PACF PLOTS OF HOUSE PRICE (HP)

```
acf(HP, lag=24)
```

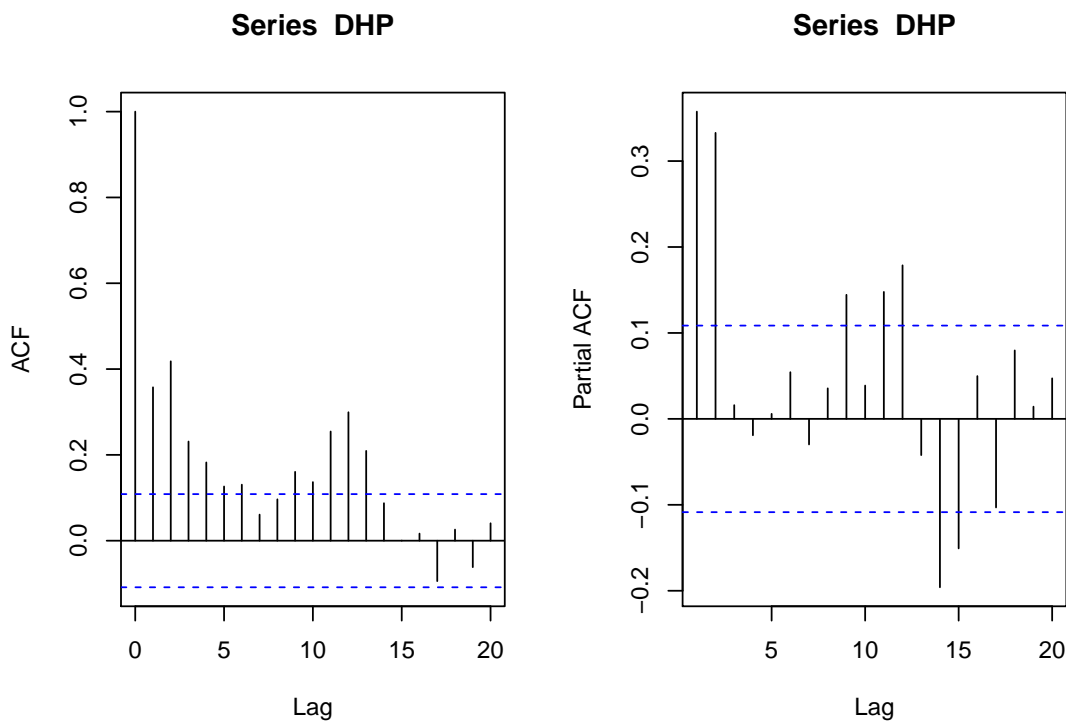


```
pacf(HP, lag=24)
```



ACF AND PACF PLOTS OF SIMPLE RETURNS (DHP)

```
layout(matrix(c(1,1,2,2),2,2))
acf(DHP,lag=20)
pacf(DHP,lag=20)
```



AUGMENTED DICKEY-FULLER TEST

In order to test whether the DHP time series is stationary or not(unit root), we perform the ADF test
`adf.test(DHP)`

```
##
## Augmented Dickey-Fuller Test
##
## data: DHP
## Dickey-Fuller = -5.1732, Lag order = 6, p-value = 0.01
## alternative hypothesis: stationary
```

FIND THE COEFFICIENT MATRIX OF ARIMA MODEL BY `coeftest()` FUNCTION

Setting the order: ARIMA(0,1,2), we can estimate the coefficient matrix (including estimates, SEs, t
`arima11<-arima(DHP,order=c(0,1,2))`
`coeftest(arima11)`

```
##
## z test of coefficients:
##
## Estimate Std. Error z value Pr(>|z|)
## ma1 -0.758418 0.052550 -14.432 <2e-16 ***
```

```
## ma2 -0.067894    0.057246   -1.186    0.2356
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

FIND AIC (INFORMATION CRITERIA) OF ARIMA(1,0,2) AND ARIMA(1,0,1)

```
# ARIMA(1,0,2)
arima11<-arima(DHP,order=c(1,0,2))
AIC(arima11)
```

```
## [1] 925.6487
```

```
# ARIMA(1,0,1)
arima11<-arima(DHP,order=c(1,0,1))
AIC(arima11)
```

```
## [1] 933.4199
```

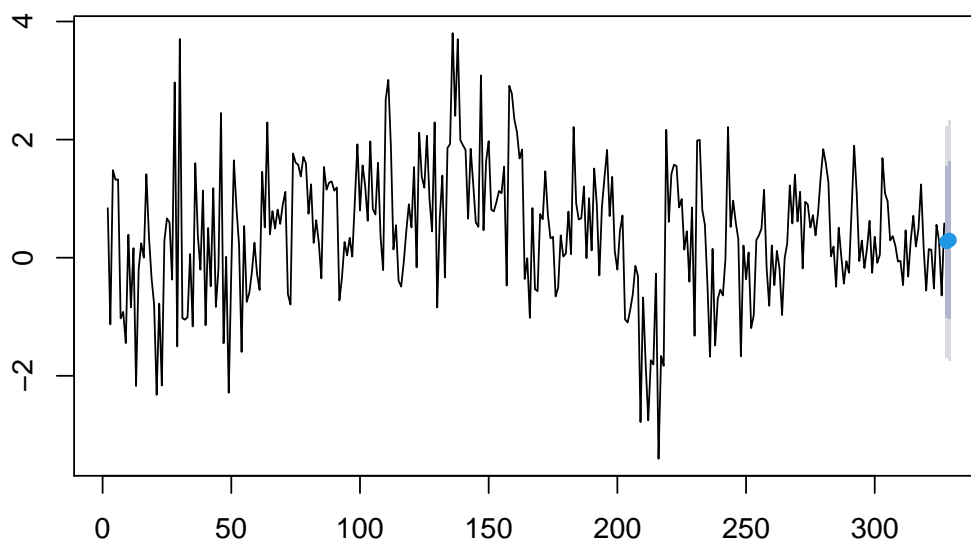
FORECASTING IN TIME SERIES

```
# forecasting for ARIMA(1,0,1) over 2 periods (h=2)
fcast <- forecast(arima11, h=2)
summary(fcast)
```

```
##
## Forecast method: ARIMA(1,0,1) with non-zero mean
##
## Model Information:
##
## Call:
## arima(x = DHP, order = c(1, 0, 1))
##
## Coefficients:
##          ar1          ma1  intercept
##          0.8224   -0.5417         0.4286
## s.e.    0.0596    0.0877         0.1414
##
## sigma^2 estimated as 0.9999:  log likelihood = -462.71,  aic = 933.42
##
## Error measures:
##              ME          RMSE          MAE          MPE          MAPE          MASE
## Training set 5.715104e-06 0.9999633 0.7567761 86.34112 206.557 0.8067148
##              ACF1
## Training set -0.05935702
##
## Forecasts:
##      Point Forecast      Lo 80      Hi 80      Lo 95      Hi 95
## 328      0.2706600 -1.010844 1.552164 -1.689232 2.230552
## 329      0.2987119 -1.032304 1.629727 -1.736901 2.334324
```

```
# Plotting
plot(fcast)
```

Forecasts from ARIMA(1,0,1) with non-zero mean



FINDING THE BEST ARIMA MODEL ACCORDING TO AIC, BIC VALUE

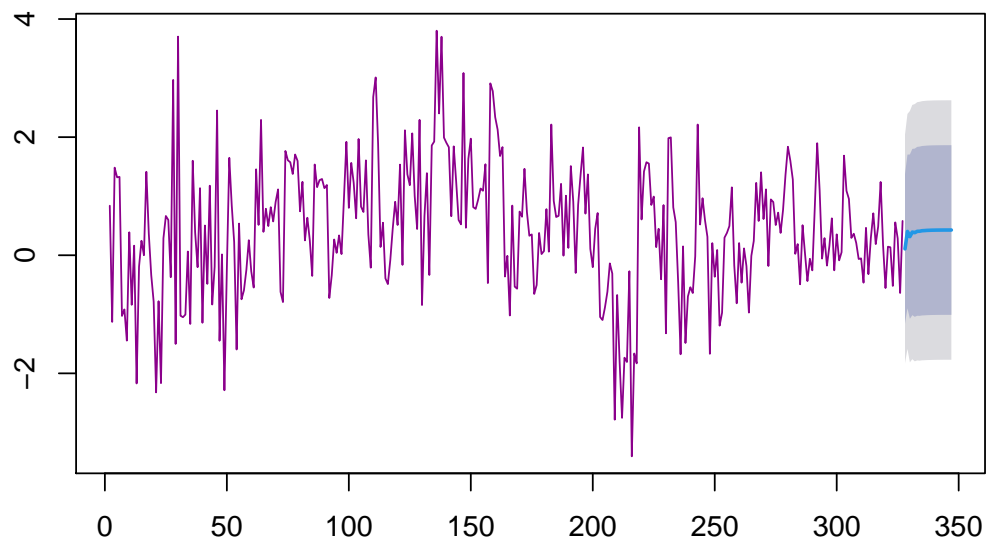
```
# Instead of finding the best ARIMA model for time series, we can use the auto.arima() function to find
auto.arima(DHP, max.order = 5) #example with max order =5
```

```
## Series: DHP
## ARIMA(2,0,0) with non-zero mean
##
## Coefficients:
##          ar1      ar2      mean
##          0.2361  0.3340  0.4275
## s.e.      0.0521  0.0523  0.1259
##
## sigma^2 estimated as 0.9756: log likelihood=-457.23
## AIC=922.46  AICc=922.58  BIC=937.61
```

```
# we find optimal ARIMA model for returns (DHP) using max arima order =10
fit<-auto.arima(DHP, max.order = 10)

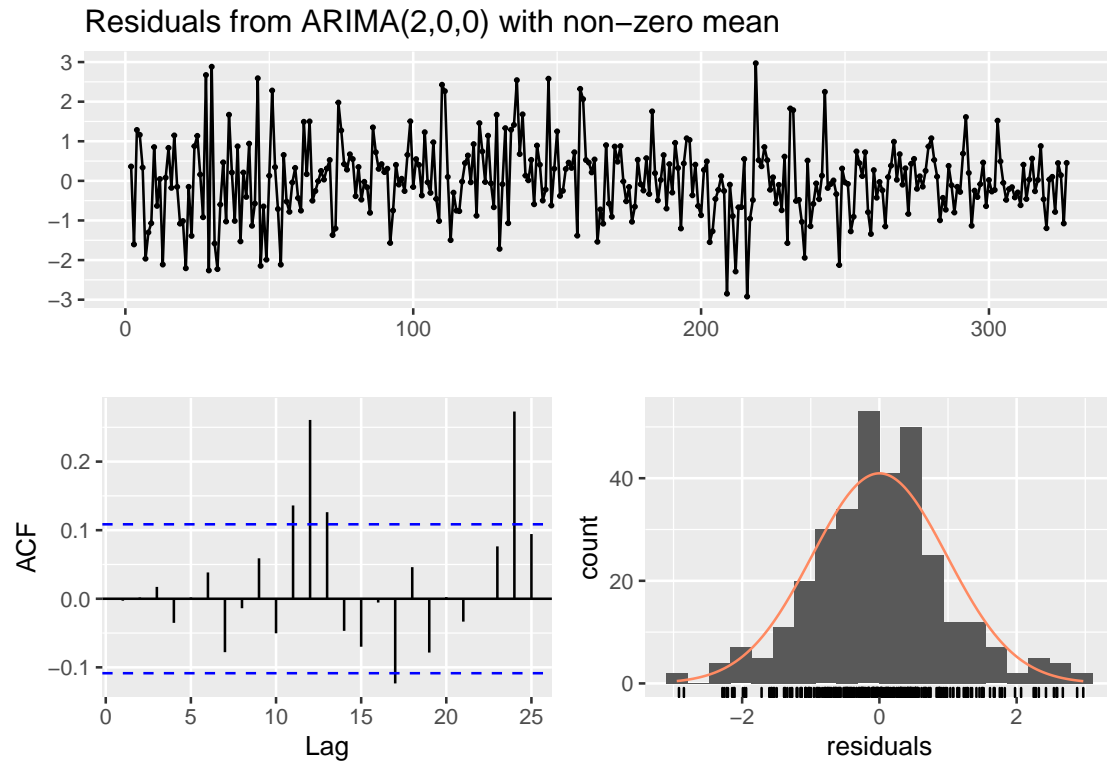
# forecasting of 'fit' over 20 periods (extra)
plot(forecast(fit,h=20),col='darkmagenta')
```


Forecasts from ARIMA(2,0,0) with non-zero mean



LJUNG-BOX TEST TO CHECK IF THE RESIDUALS SERIES IS WHITE NOISE PROCESS

```
checkresiduals(fit)
```



```
##
##  Ljung-Box test
##
## data:  Residuals from ARIMA(2,0,0) with non-zero mean
## Q* = 5.1323, df = 7, p-value = 0.6438
##
## Model df: 3.   Total lags used: 10
```