

HW#4

Name: **Jingyi Wu**

Due Date: April 14, 2019

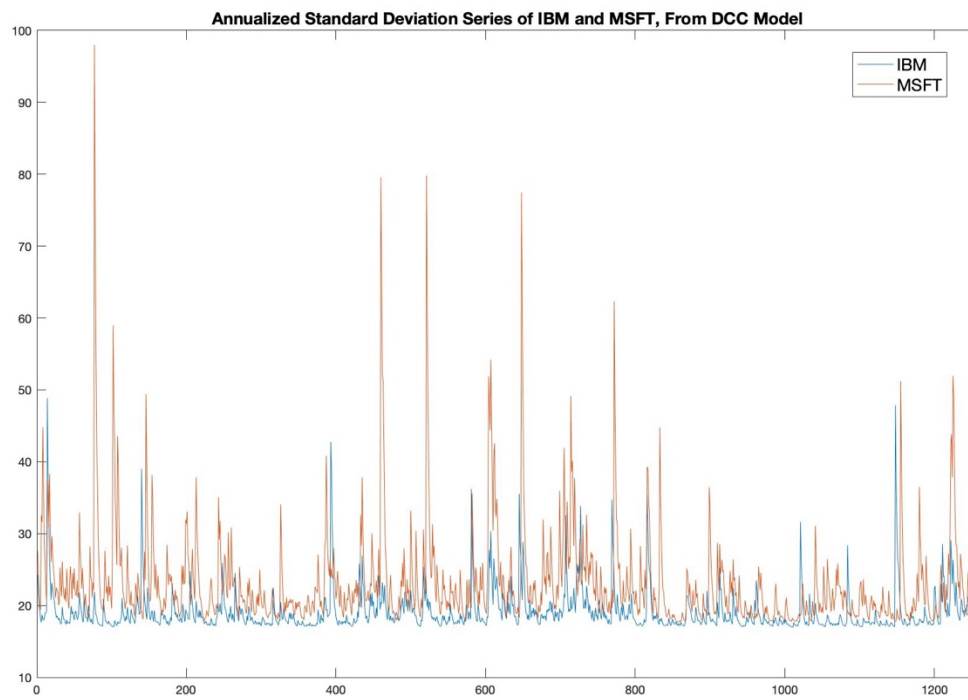
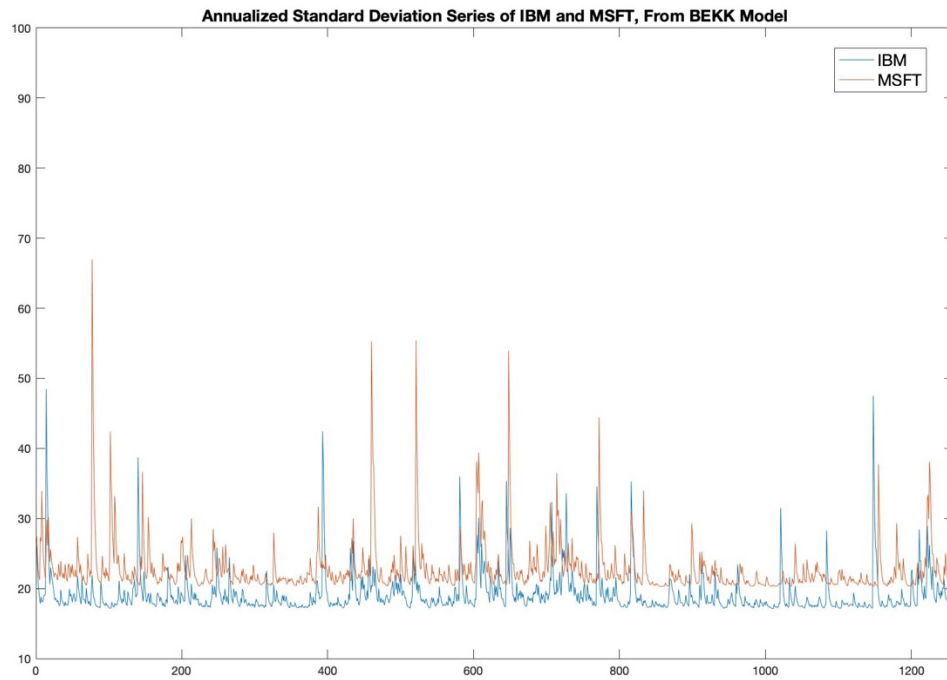
Course Code: ECON 623 Forecasting Financial Markets

Instructor: Professor Andrew Patton

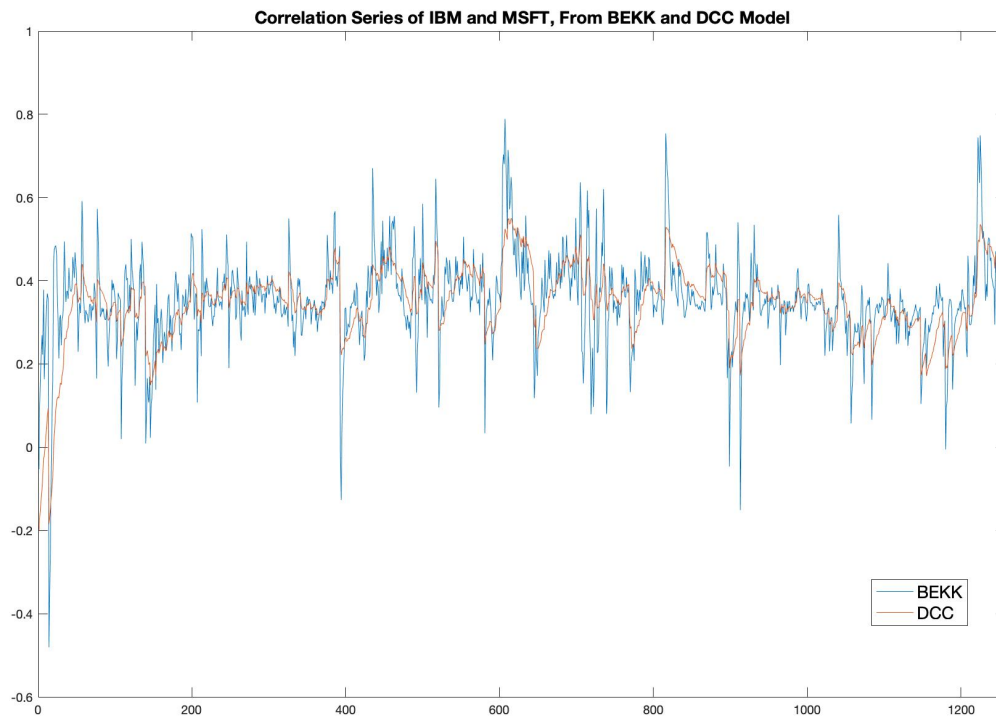
Exercise 1

Data: IBM and MSFT on 4/2/13-3/29/18
Finance.yahoo.com

(c)



(d)



(d)

MZ test: $H_0: \beta_0 = 0$ and $\beta_1 = 1$ $H_1: \beta_0 \neq 0$ and $\beta_1 \neq 1$

1. test for IBM in bekk

TestValue1 =

12.1192

CriticalValue1 =

5.9915

ans =

logical

0

MZ test statistics is larger than the critical value. So, we can reject the null. Therefore, there is evidence to show that BEKK model for IBM is optimal.

2. test for MSFT in BEKK

```
TestValue2 =
```

```
2.2549
```

```
CriticalValue2 =
```

```
5.9915
```

```
ans =
```

```
logical
```

```
1
```

MZ test statistics is smaller than the critical value. So, we cannot reject the null. Therefore, there is no evidence to show that BEKK model for MSFT is optimal.

3. test for IBM in DCC

```
TestValue3 =
```

```
12.9797
```

```
CriticalValue3 =
```

```
5.9915
```

```
ans =
```

```
logical
```

```
0
```

MZ test statistics is larger than the critical value. So, we can reject the null. Therefore, there is evidence to show that DCC model for IBM is optimal.

4. test for MSFT in DCC

```
TestValue4 =
```

```
46.8541
```

```
CriticalValue4 =
```

```
5.9915
```

```
ans =
```

```
logical
```

```
0
```

MZ test statistics is larger than the critical value. So, we can reject the null. Therefore, there is evidence to show that DCC model for MSFT is optimal.

5. test for Covariance in BEKK

TestValue5 =

4.2994

CriticalValue5 =

5.9915

ans =

logical

1

MZ test statistics is smaller than the critical value. So, we cannot reject the null. Therefore, there is no evidence to show that BEKK model for Covariance is optimal.

6. test for Covariance in DCC

TestValue6 =

4.2630

CriticalValue6 =

5.9915

ans =

logical

1

MZ test statistics is smaller than the critical value. So, we cannot reject the null. Therefore, there is no evidence to show that DCC model for Covariance is optimal.

(e)

DM test: $H_0: E(\bar{d}) = 0$. $H_1: E(\bar{d}) \neq 0$

1.w1 = [1,0]

<u>DM1</u>	<u>CriticalValue</u>
-1.7642	1.96

DM test statistics is smaller than the critical value. So, we cannot reject the null. The two estimation perform the same.

2.w2 = [0.5,0.5]

<u>DM2</u>	<u>CriticalValue</u>
-1.4141	1.96

DM test statistics is smaller than the critical value. So, we cannot reject the null. The two estimation perform the same.

3.w2 = [0,1]

DM3	CriticalValue
-2.6146	1.96

DM test statistics is larger than the critical value. So, we can reject the null. The DCC model perform better.

Code:

```
%% Exercise 1
cd('/Users/killshadows/Desktop/DUKE/COURSES/SPRING2019/ECON623/TA Session/5')
addpath('/Users/killshadows/Desktop/DUKE/COURSES/SPRING2019/ECON623/TA Session/5/mfe-toolbox-master/multivariate')
addpath('/Users/killshadows/Desktop/DUKE/COURSES/SPRING2019/ECON623/TA Session/5/mfe-toolbox-master/utility')
addpath('/Users/killshadows/Desktop/DUKE/COURSES/SPRING2019/ECON623/TA Session/5/mfe-toolbox-master/distributions')
addpath('/Users/killshadows/Desktop/DUKE/COURSES/SPRING2019/ECON623/TA Session/5/mfe-toolbox-master/univariate')

IBM = csvread('IBM.csv', 1, 1);
MSFT = csvread('MSFT.csv', 1, 1);

IBM = IBM(:, 5);
MSFT = MSFT(:, 5);

IBM = 100*price2ret(IBM);
MSFT = 100*price2ret(MSFT);

% (a) BEKK Model
data = [IBM MSFT];
data = [data(:, 1)-mean(data(:, 1)) data(:, 2)-mean(data(:, 2))]; %obtain residual

[~, ~, Ht_a, ~, ~] = bekk(data,[], 1,0, 1); %estimate multivariate volatility by BEKK

vol_bekk_IBM = squeeze(Ht_a(1,1,:)); %obtain volatility of IBM from BEKK model
sd_bekk_IBM = sqrt(252) * sqrt(vol_bekk_IBM); %annualized sd of IBM from BEKK model

vol_bekk_MSFT = squeeze(Ht_a(2,2,:)); %obtain volatility of MSFT from BEKK model
sd_bekk_MSFT = sqrt(252) * sqrt(vol_bekk_MSFT); %annualized sd of IBM from BEKK model

plot(sd_bekk_IBM)
hold on
plot(sd_bekk_MSFT)
hold off
title('Annualized Standard Deviation Series of IBM and MSFT, From BEKK Model')
```

```

% (b) DCC Model
[~, ~, Ht_b, ~, ~] = dcc(data,[], 1,0, 1); %estimate multivariate volatility
by DCC

vol_dcc_IBM = squeeze(Ht_b(1,1,:)); %obtain volatility of IBM from DCC model
sd_dcc_IBM = sqrt(252) * sqrt(vol_dcc_IBM); %annualized sd of IBM from DCC
model

vol_dcc_MSFT = squeeze(Ht_b(2,2,:)); %obtain volatility of MSFT from DCC
model
sd_dcc_MSFT = sqrt(252) * sqrt(vol_dcc_MSFT); %annualized sd of IBM from DCC
model

plot(sd_dcc_IBM)
hold on
plot(sd_dcc_MSFT)
hold off
title('Annualized Standard Deviation Series of IBM and MSFT, From DCC Model')

% (d) correlation of two model
cov_bekk = squeeze(Ht_a(1,2,:));
corr_bekk = cov_bekk./sqrt(vol_bekk_IBM.*vol_bekk_MSFT);

cov_dcc = squeeze(Ht_b(1,2,:));
corr_dcc = cov_dcc./sqrt(vol_dcc_IBM.*vol_dcc_MSFT);

plot(corr_bekk)
hold on
plot(corr_dcc)
hold off
title('Correlation Series of IBM and MSFT, From BEKK and DCC Model')

% (e) Mincer-Zarnowitz test
proxy_var_IBM = data(:,1).^2;
proxy_var_MSFT = data(:,2).^2;
proxy_cov = data(:,1).*data(:,2);

% test for IBM in bekk
Intercept = ones(size(vol_bekk_IBM, 1), 1);
X = [Intercept vol_bekk_IBM];
Y = proxy_var_IBM;
results = nwest(Y,X);
beta1 = results.beta;
se1 = results.se;
vcv1 = results.vcv;
R1 = [1 0
      0 1];
%Calculate test statistics
TestValue1 = (R1*beta1 - [0; 1])'*((R1'*vcv1*R1)\(R1*beta1 - [0; 1]))
%Calculate Critical Value for 5% significance level
CriticalValue1 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue1 < CriticalValue1 %reject

% test for MSFT in bekk
Intercept = ones(size(vol_bekk_MSFT, 1), 1);

```

```

X = [Intercept vol_bekk_MSFT];
Y = proxy_var_MSFT;
results = nwest(Y,X);
beta2 = results.beta;
se2 = results.se;
vcv2 = results.vcv;
R2 = [1 0
      0 1];
%Calculate test statistics
TestValue2 = (R2*beta2 - [0; 1])'*((R2'*vcv2*R2)\(R2*beta2 - [0; 1]))
%Calculate Critical Value for 5% significance level
CriticalValue2 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue2 < CriticalValue2 %fail to reject

% test for IBM in dcc
Intercept = ones(size(vol_dcc_IBM, 1), 1);
X = [Intercept vol_dcc_IBM];
Y = proxy_var_IBM;
results = nwest(Y,X);
beta3 = results.beta;
se3 = results.se;
vcv3 = results.vcv;
R3 = [1 0
      0 1];
%Calculate test statistics
TestValue3 = (R3*beta3 - [0; 1])'*((R3'*vcv3*R3)\(R3*beta3 - [0; 1]))
%Calculate Critical Value for 5% significance level
CriticalValue3 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue3 < CriticalValue3 %reject

% test for MSFT in dcc
Intercept = ones(size(vol_dcc_MSFT, 1), 1);
X = [Intercept vol_dcc_MSFT];
Y = proxy_var_MSFT;
results = nwest(Y,X);
beta4 = results.beta;
se4 = results.se;
vcv4 = results.vcv;
R4 = [1 0
      0 1];
%Calculate test statistics
TestValue4 = (R4*beta4 - [0; 1])'*((R4'*vcv4*R4)\(R4*beta4 - [0; 1]))
%Calculate Critical Value for 5% significance level
CriticalValue4 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue4 < CriticalValue4 %reject

% test for covariance in bekk
Intercept = ones(size(cov_bekk, 1), 1);
X = [Intercept cov_bekk];
Y = proxy_cov;
results = nwest(Y,X);
beta5 = results.beta;
se5 = results.se;
vcv5 = results.vcv;
R5 = [1 0
      0 1];

```



```

%Calculate test statistics
TestValue5 = (R5*beta5 - [0; 1])'*((R5'*vcv5*R5)\(R5*beta5 - [0; 1]))
%Calculate Critical Value for 5% significance level
CriticalValue5 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue5 < CriticalValue5 %fail to reject

% test for covariance in dcc
Intercept = ones(size(cov_dcc, 1), 1);
X = [Intercept cov_dcc];
Y = proxy_cov;
results = nwest(Y,X);
beta6 = results.beta;
se6 = results.se;
vcv6 = results.vcv;
R6 = [1 0
      0 1];
%Calculate test statistics
TestValue6 = (R6*beta6 - [0; 1])'*((R6'*vcv6*R6)\(R6*beta6 - [0; 1]))
%Calculate Critical Value for 5% significance level
CriticalValue6 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue6 < CriticalValue6 %fail to reject

% (f) Diebold-Mariano test
%w1 = [1,0]
e_bekk_w1_squared = (proxy_var_IBM - vol_bekk_IBM).^2;
e_dcc_w1_squared = (proxy_var_IBM - vol_dcc_IBM).^2;
d1bar = mean(e_bekk_w1_squared - e_dcc_w1_squared);
d1var = var(e_bekk_w1_squared - e_dcc_w1_squared);
DM1 = d1bar/sqrt(d1var/1258);
CriticalValue = 1.96;
Summary1 = table(DM1, CriticalValue) %fail to reject, perform the same

%w2 = [0.5,0.5]
w2 = [0.5, 0.5];
proxy_w2 = (w2*data').^2';
vol_bekk_w2 = 0.25*vol_bekk_IBM+0.25*vol_bekk_MSFT+0.5*cov_bekk;
vol_dcc_w2 = 0.25*vol_dcc_IBM+0.25*vol_dcc_MSFT+0.5*cov_dcc;
e_bekk_w2_squared = (proxy_w2 - vol_bekk_w2).^2;
e_dcc_w2_squared = (proxy_w2 - vol_dcc_w2).^2;
d2bar = mean(e_bekk_w2_squared - e_dcc_w2_squared);
d2var = var(e_bekk_w2_squared - e_dcc_w2_squared);
DM2 = d2bar/sqrt(d2var/1258);
CriticalValue = 1.96;
Summary2 = table(DM2, CriticalValue) %fail to reject, perform the same

%w3 = [0,1]
e_bekk_w3_squared = (proxy_var_MSFT - vol_bekk_MSFT).^2;
e_dcc_w3_squared = (proxy_var_MSFT - vol_dcc_MSFT).^2;
d3bar = mean(e_bekk_w3_squared - e_dcc_w3_squared);
d3var = var(e_bekk_w3_squared - e_dcc_w3_squared);
DM3 = d3bar/sqrt(d3var/1258);
CriticalValue = 1.96;
Summary3 = table(DM3, CriticalValue) %reject, DCC perform better

```

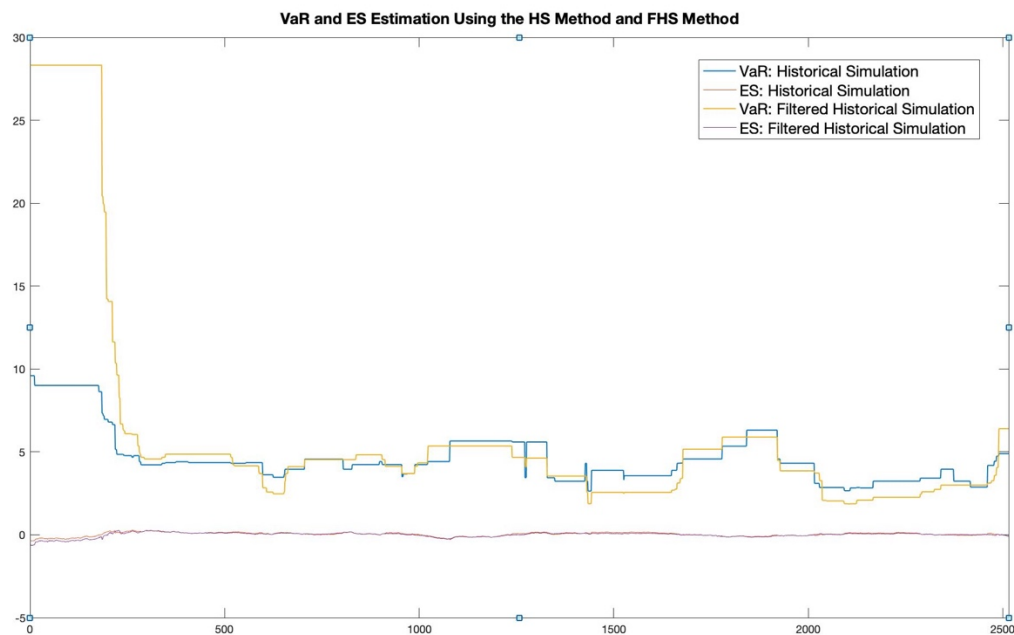
Exercise 2

data input: AAPL from 1/1/2008 - 12/31/2018

(a)

VaR_un_1	VaR_un_5	VaR_un_10
-5.6014	-2.8754	-2.1142
ES_un_1	ES_un_5	ES_un_10
-7.6757	-4.6601	-3.5614

(d)



(e)

1. test HS method

```
beta7 =  
    0.9855  
   -0.0143  
    0.0038  
  
TestValue7 =  
    9.1222e+06  
  
CriticalValue7 =  
    5.9915  
  
ans =  
  
logical  
0
```

Reject the null

2. test FHS method

```
beta8 =  
    0.9758  
   -0.0035  
    0.0013  
  
TestValue8 =  
    4.1544e+05  
  
CriticalValue8 =  
    5.9915  
  
ans =  
  
logical  
0
```

Reject the null

(f)

1.test HS method

beta9 =

16.8097
18.9243
1.3289

TestValue9 =

2.5595e+03

CriticalValue9 =

5.9915

ans =

logical

0

Reject the null

2. test FHS method

beta10 =

16.4936
9.2802
2.2405

TestValue10 =

2.6902e+03

CriticalValue10 =

5.9915

ans =

logical

0

Reject the null

(g)

DM4	CriticalValue
<hr/>	<hr/>
-13.394	1.96

DM test statistics is larger than the critical value. So, we can reject the null. The HS model perform better.

(h)

1x2 [table](#)

DM5	CriticalValue
$-0.49177 + 1.5234e-05i$	1.96

Weird output because there is some $ES > 0$.

Code:

```
%% Exercise 2
% data input: AAPL from 1/1/2008 - 12/31/2018
AAPL = csvread('AAPL.csv', 1, 1);
AAPL = AAPL(:, 5);
AAPL = 100*price2ret(AAPL);

% (a) unconditional VaR and ES
VaR_un_1 = quantile(AAPL, .01);
ES_un_1 = mean(AAPL(AAPL < VaR_un_1));

VaR_un_5 = quantile(AAPL, .05);
ES_un_5 = mean(AAPL(AAPL < VaR_un_5));

VaR_un_10 = quantile(AAPL, .10);
ES_un_10 = mean(AAPL(AAPL < VaR_un_10));

Summary_VaR_un = table(VaR_un_1, VaR_un_5, VaR_un_10)
Summary_ES_un = table(ES_un_1, ES_un_5, ES_un_10)

% (b) historical simulation
testwindow = 254:2768; %start at 1/2/2009
windowsize = 250;
pVaR = 0.01;
historical1 = zeros(length(testwindow), 1);
for t = testwindow
    i = t - 254 + 1;
    estimationwindow = t-windowsize:t-1;
    X = AAPL(estimationwindow);
    historical1(i) = -quantile(X, pVaR);
    es1(i) = mean(X(X < historical1(i)));
end

% (c) filtered historical simulation
sd_old = sqrt(var(AAPL)); %old volatility

results = nwest(AAPL, ones(length(AAPL), 1)); %use GARCH model to get new
volatility
resid = AAPL - results.yhat;
parameters = tarch(resid, 1, 0, 1);
```

```

sigmasqr = NaN(length(resid)+1, 1);
sigmasqr(1) = var(resid);
omega = parameters(1);
alpha = parameters(2);
beta = parameters(3);

for i = 2:length(resid)+1

    sigmasqr(i) = omega + alpha*resid(i-1)^2 + beta*sigmasqr(i-1);

end

sd_new = sqrt(sigmasqr); %obtain new volatility
sd_new = sd_new(1:2768,:);
AAPL_filtered = AAPL.*sd_new/sd_old; %obtain filtered return

historical2 = zeros(length(testwindow),1); %do historical simulation
for t = testwindow
    i = t - 254 + 1;
    estimationwindow = t-windowsize:t-1;
    X = AAPL_filtered(estimationwindow);
    historical2(i) = -quantile(X,pVaR);
    es2(i) = mean(X(X < historical2(i)));
end

% (d) plot
plot(historical1)
hold on
plot(es1)
plot(historical2)
plot(es2)
hold off
title('VaR and ES Estimation Using the HS Method and FHS Method')

% (e)
% test HS method
hit_hs = AAPL(254:2768);
for i = 1:2515
    if hit_hs(i) <= historical1(i)
        hit_hs(i) = 1;
    else
        hit_hs(i) = 0;
    end
end

Intercept = ones(size(hit_hs(1:end-1), 1), 1);
X = [Intercept hit_hs(1:end-1) historical1(2:end)];
Y = hit_hs(2:end);
results = nwest(Y,X);
beta7 = results.beta
se7 = results.se;
vcv7 = results.vcv;
R7 = [1 0 0
      0 1 0
      0 0 1];
%Calculate test statistics
TestValue7 = (R7*beta7 - [0.01; 0; 0])'*((R7'*vcv7*R7)\(R7*beta7 - [0.01; 0;
0]))

```

```

%Calculate Critical Value for 5% significance level
CriticalValue7 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue7 < CriticalValue7 %reject

%test FHS method
hit_fhs = AAPL(254:2768);
for i = 1:2515
if hit_fhs(i) <= historical2(i)
    hit_fhs(i) = 1;
else
    hit_fhs(i) = 0;
end
end

Intercept = ones(size(hit_fhs(1:end-1), 1), 1);
X = [Intercept hit_fhs(1:end-1) historical2(2:end)];
Y = hit_fhs(2:end);
results = nwest(Y,X);
beta8 = results.beta
se8 = results.se;
vcv8 = results.vcv;
R8 = [1 0 0
      0 1 0
      0 0 1];
%Calculate test statistics
TestValue8 = (R8*beta8 - [0.01; 0; 0])'*((R8'*vcv8*R8)\(R8*beta8 - [0.01; 0; 0]))
%Calculate Critical Value for 5% significance level
CriticalValue8 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue8 < CriticalValue8 %reject

% (f)
for i = 1:2515
a = AAPL(254:2768).*hit_hs;
es_proxy(i) = 100*mean(AAPL(254:254+i-1));
end
es_proxy = es_proxy';

es1 = es1';
es2 = es2';

%test HS method
Intercept = ones(size(hit_hs(1:end-1), 1), 1);
X = [Intercept es1(2:end) hit_hs(1:end-1)];
Y = es_proxy(2:end);
results = nwest(Y,X);
beta9 = results.beta
se9 = results.se;
vcv9 = results.vcv;
R9 = [1 0 0
      0 1 0
      0 0 1];
%Calculate test statistics
TestValue9 = (R9*beta9 - [0; 1; 0])'*((R9'*vcv9*R9)\(R9*beta9 - [0; 1; 0]))
%Calculate Critical Value for 5% significance level
CriticalValue9 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value

```

```

TestValue9 < CriticalValue9 %reject

%test FHS method
Intercept = ones(size(hit_fhs(1:end-1), 1), 1);
X = [Intercept es2(2:end) hit_fhs(1:end-1)];
Y = es_proxy(2:end);
results = nwest(Y,X);
beta10 = results.beta
se10 = results.se;
vcv10 = results.vcv;
R10 = [1 0 0
       0 1 0
       0 0 1];
%Calculate test statistics
TestValue10 = (R10*beta10 - [0; 1; 0])'*((R10'*vcv10*R10)\(R10*beta10 - [0;
1; 0]))
%Calculate Critical Value for 5% significance level
CriticalValue10 = chi2inv(1-0.05,2)
%Test if test statistics is less than the Critical Value
TestValue10 < CriticalValue10 %reject

% (g)
loss_hs = (hit_hs-0.01).*(historical1-AAPL(254:2768));
loss_fhs = (hit_fhs-0.01).*(historical2-AAPL(254:2768));
dbar4 = mean(loss_hs-loss_fhs);
dvar4 = var(loss_hs-loss_fhs);
dbarvar4 = dvar4/2515;
DM4 = dbar4/sqrt(dbarvar4);
CriticalValue = 1.96;
Summary4 = table(DM4, CriticalValue)

% (h)
loss_hs2 = ((hit_hs.*(AAPL(254:2768) - historical1))./(0.01*es1)) - ((es1-
historical1)./es1) + log(-es1);
loss_fhs2 = ((hit_fhs.*(AAPL(254:2768) - historical2))./(0.01*es2)) - ((es2-
historical2)./es2) + log(-es2);
dbar5 = mean(loss_hs2-loss_fhs2);
dvar5 = var(loss_hs2-loss_fhs2);
dbarvar5 = dvar5/2515;
DM5 = dbar5/sqrt(dbarvar5);
CriticalValue = 1.96;
Summary5 = table(DM5, CriticalValue)

```

Exercise 2

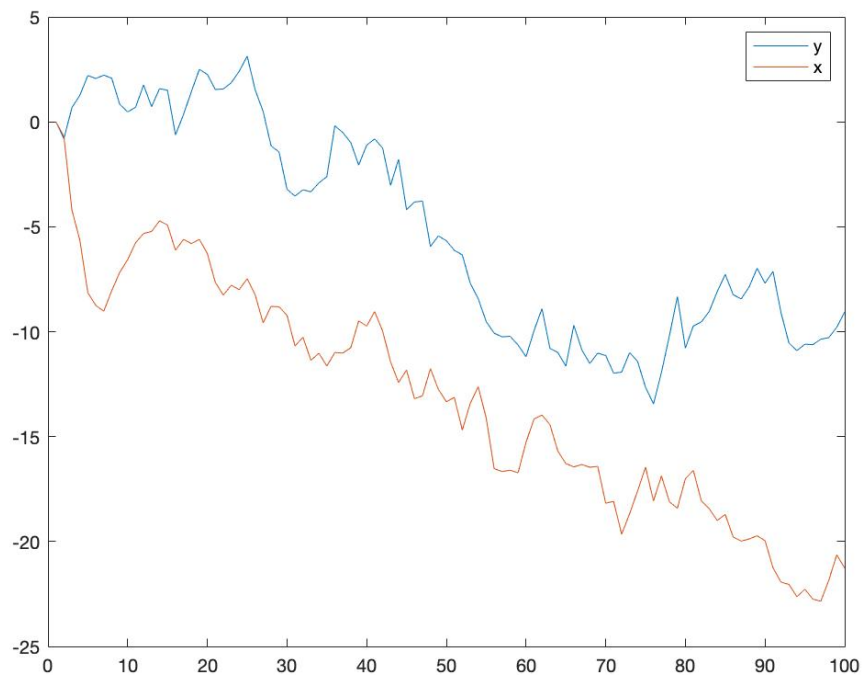
(b)


```

% (c)
x = zeros(100,1);
y = zeros(100,1);
Tstatistics = zeros(1000,1);
for j = 1:1000
    for i = 2:100
        epsx = randn(100,1);
        x(i) = x(i-1) + epsx(i);
        epsy = randn(100,1);
        y(i) = y(i-1) + epsy(i);
    end
    results = ols(y, [ones(length(x),1) x]);
    Tstatistics(j) = results.tstat(2);
end

plot(y)
hold on
plot(x)
hold off

```



(c)

	StandardNormal	sim
Pr[tstat<-2.58]	0.005	0.353
Pr[tstat<-1.96]	0.025	0.385
Pr[tstat<-1.65]	0.05	0.409
Pr[tstat<0]	0.5	0.513
Pr[tstat<1.65]	0.05	0.393
Pr[tstat<1.96]	0.025	0.368
Pr[tstat<2.58]	0.005	0.34

Code:

```
%% Exercise 3
% (a)
x = zeros(100,1);
y = zeros(100,1);
for i = 2:100
    epsx = randn(100,1);
    x(i) = x(i-1) + epsx(i);
    epsy = randn(100,1);
    y(i) = y(i-1) + epsy(i);
end

% (b)
results = ols(y, [ones(length(x),1) x]);
Tstatistics_default = results.tstat(2);

% (c)
x = zeros(100,1);
y = zeros(100,1);
Tstatistics = zeros(1000,1);
for j = 1:1000
    for i = 2:100
        epsx = randn(100,1);
        x(i) = x(i-1) + epsx(i);
        epsy = randn(100,1);
        y(i) = y(i-1) + epsy(i);
    end
    results = ols(y, [ones(length(x),1) x]);
    Tstatistics(j) = results.tstat(2);
end

plot(y)
hold on
plot(x)
hold off

% (d)
sim = zeros(7,1);
t1 = Tstatistics < -2.58;
sim(1) = sum(t1)/1000;
t2 = Tstatistics < -1.96;
sim(2) = sum(t2)/1000;
t3 = Tstatistics < -1.65;
sim(3) = sum(t3)/1000;
t4 = Tstatistics < 0;
sim(4) = sum(t4)/1000;
t5 = Tstatistics > 1.65;
sim(5) = sum(t5)/1000;
t6 = Tstatistics > 1.96;
sim(6) = sum(t6)/1000;
t7 = Tstatistics > 2.58;
sim(7) = sum(t7)/1000;
StandardNormal = [0.005; 0.025; 0.05; 0.5; 0.05; 0.025; 0.005];
table = table(StandardNormal,sim);
table.Properties.RowNames = {'Pr[tstat<-2.58]'; 'Pr[tstat<-1.96]';
'Pr[tstat<-1.65]'; 'Pr[tstat<0]'; 'Pr[tstat<1.65]'; 'Pr[tstat<1.96]';
'Pr[tstat<2.58]']}
```