

Group Project 2

CAPM and a multivariate model of stock returns

Variables:

Sector portfolio excess returns (return minus the risk free rate)

BMATRUSER Building materials

INDUSUSER Industrials

CNSMGUSER Consumer goods

FINANUSER Financials

TECNOUSER Technology

MktRF: Market return - risk free rate

SMB Fama-French Small minus Big

HML Fama- French High minus Low

RF risk free rate

```
% import data
```

```
%rename variables
```

```
BMATRUSER = returns{ : ,2};
```

```
INDUSUSER = returns{ : ,3};
```

```
CNSMGUSER = returns{ : ,4};
```

```
FINANUSER = returns{ : ,5};
```

```
TECNOUSER = returns{ : ,6};
```

```
MktRF = returns{ : ,7};
```

```
SMB = returns{ : ,8};
```

```
HML = returns{ : ,9};
```

CAPM estimation

Q1. (10 points) Estimate CAPM regressions for each of the 5 portfolios and report the results.

Is the CAPM rejected?

%INSERT CODE

```
x1 = table(MktRF,BMATRUSER);  
lm1 = fitlm(x1)
```

```
lm1 =  
Linear regression model:  
BMATRUSER ~ 1 + MktRF
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.73685	0.27321	2.697	0.0072198
MktRF	0.081883	0.061673	1.3277	0.18485

Number of observations: 532, Error degrees of freedom: 530
Root Mean Squared Error: 6.23
R-squared: 0.00331, Adjusted R-Squared: 0.00143
F-statistic vs. constant model: 1.76, p-value = 0.185

```
x2 = table(MktRF,INDUSUSER);  
lm1 = fitlm(x2),
```

```
lm1 =  
Linear regression model:  
INDUSUSER ~ 1 + MktRF
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.91565	0.23429	3.9081	0.00010506
MktRF	0.10704	0.052887	2.024	0.043473

Number of observations: 532, Error degrees of freedom: 530
Root Mean Squared Error: 5.34
R-squared: 0.00767, Adjusted R-Squared: 0.0058
F-statistic vs. constant model: 4.1, p-value = 0.0435

```
x3 = table(MktRF,CNSMGUSER);  
lm1 = fitlm(x3)
```

```
lm1 =  
Linear regression model:  
CNSMGUSER ~ 1 + MktRF
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.67519	0.22407	3.0133	0.0027079
MktRF	0.028628	0.050579	0.566	0.57163

Number of observations: 532, Error degrees of freedom: 530
Root Mean Squared Error: 5.11

R-squared: 0.000604, Adjusted R-Squared: -0.00128
 F-statistic vs. constant model: 0.32, p-value = 0.572

```
x4 = table(MktRF,FINANUSER);
lm1 = fitlm(x4)
```

lm1 =
 Linear regression model:
 FINANUSER ~ 1 + MktRF

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.80996	0.24665	3.2838	0.0010918
MktRF	0.14648	0.055677	2.6309	0.0087637

Number of observations: 532, Error degrees of freedom: 530
 Root Mean Squared Error: 5.63
 R-squared: 0.0129, Adjusted R-Squared: 0.011
 F-statistic vs. constant model: 6.92, p-value = 0.00876

```
x5 = table(MktRF,TECNOUSER);
lm1 = fitlm(x5)
```

lm1 =
 Linear regression model:
 TECNOUSER ~ 1 + MktRF

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.0117	0.29754	3.4002	0.000724
MktRF	0.11309	0.067165	1.6838	0.092812

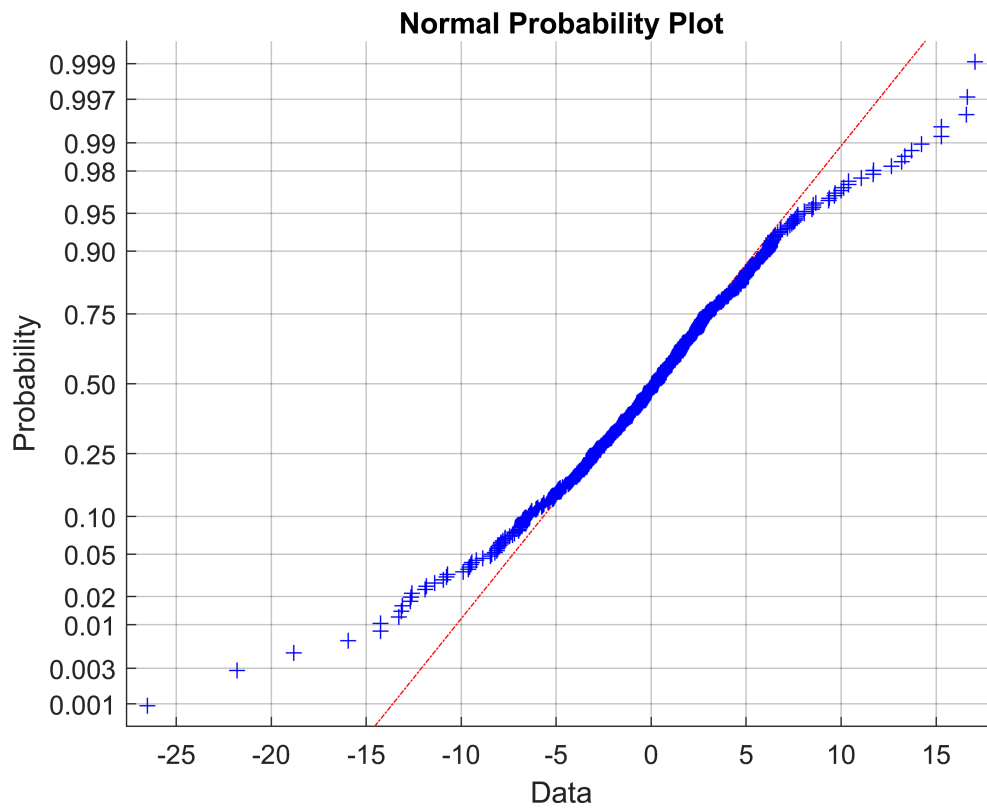
Number of observations: 532, Error degrees of freedom: 530
 Root Mean Squared Error: 6.79
 R-squared: 0.00532, Adjusted R-Squared: 0.00344
 F-statistic vs. constant model: 2.84, p-value = 0.0928

Answer:

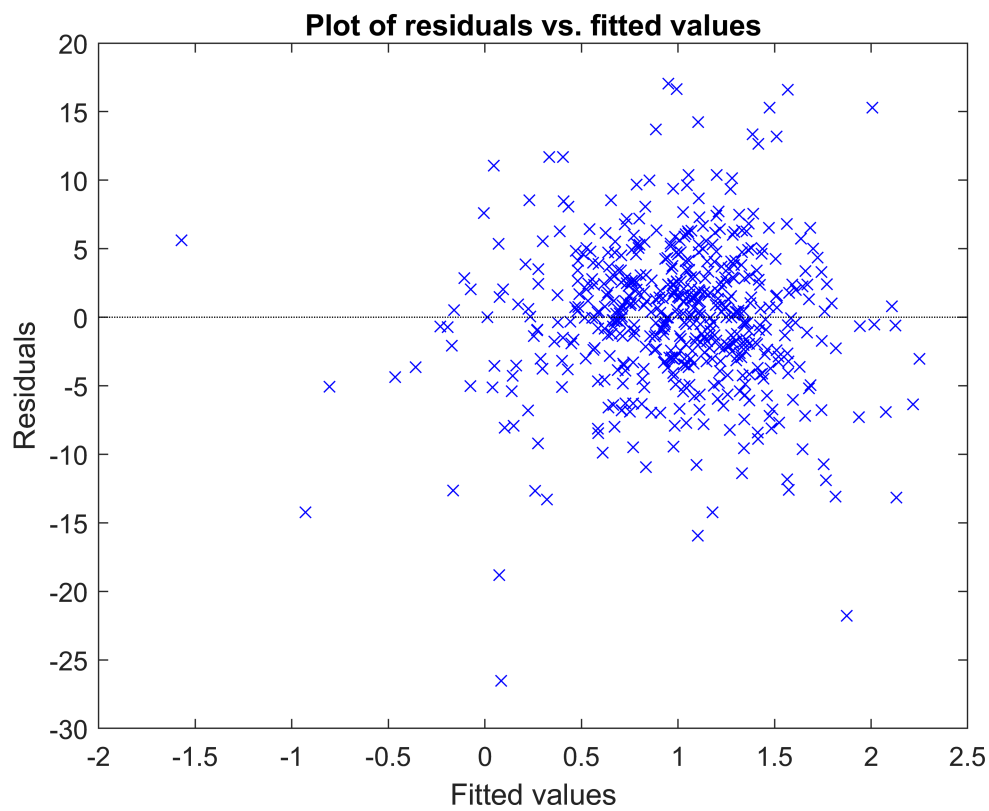
Q2. (6 points) Do the residuals of the CAPM regressions appear to be distributed normally, exhibit heteroskedasticity and autocorrelation?

Answer this question for INDUSUSER, FINANUSER, and TECNOUSER

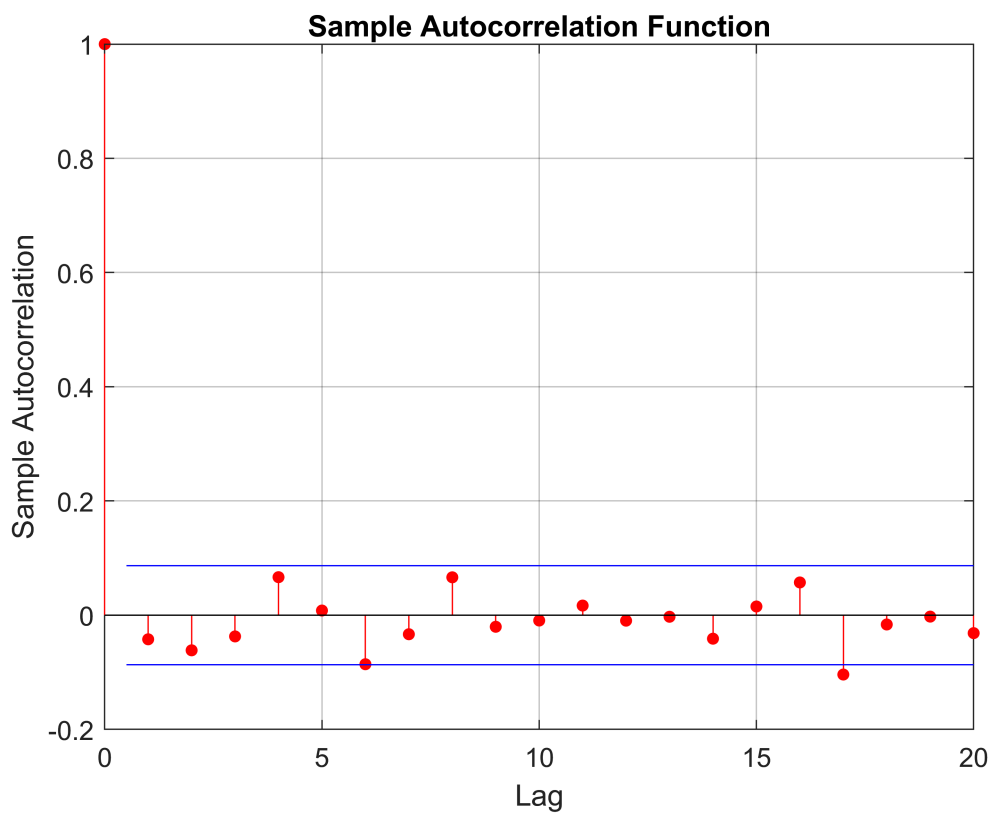
```
% INSERT CODE
x1 = table(MktRF,INDUSUSER);
lm1 = fitlm(x1);
resid = Md1.Residuals.Raw(~isnan(Md1.Residuals.Raw));
normplot(resid)
```



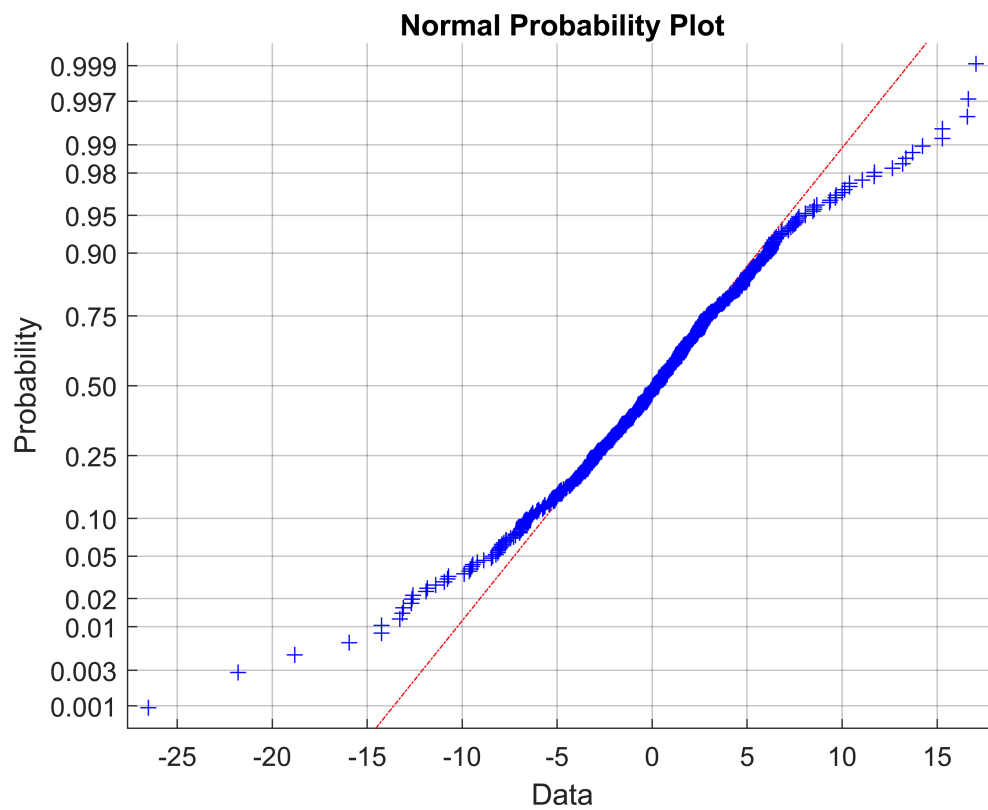
```
plotResiduals(lm1,'fitted')
```



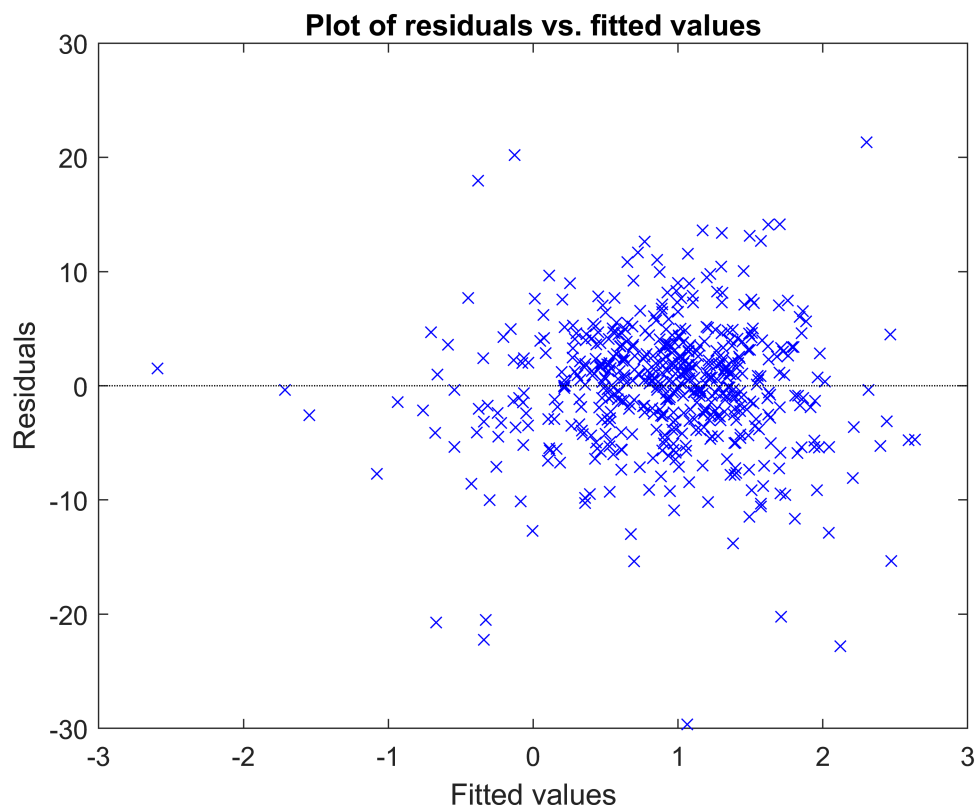
```
autocorr(resid)
```



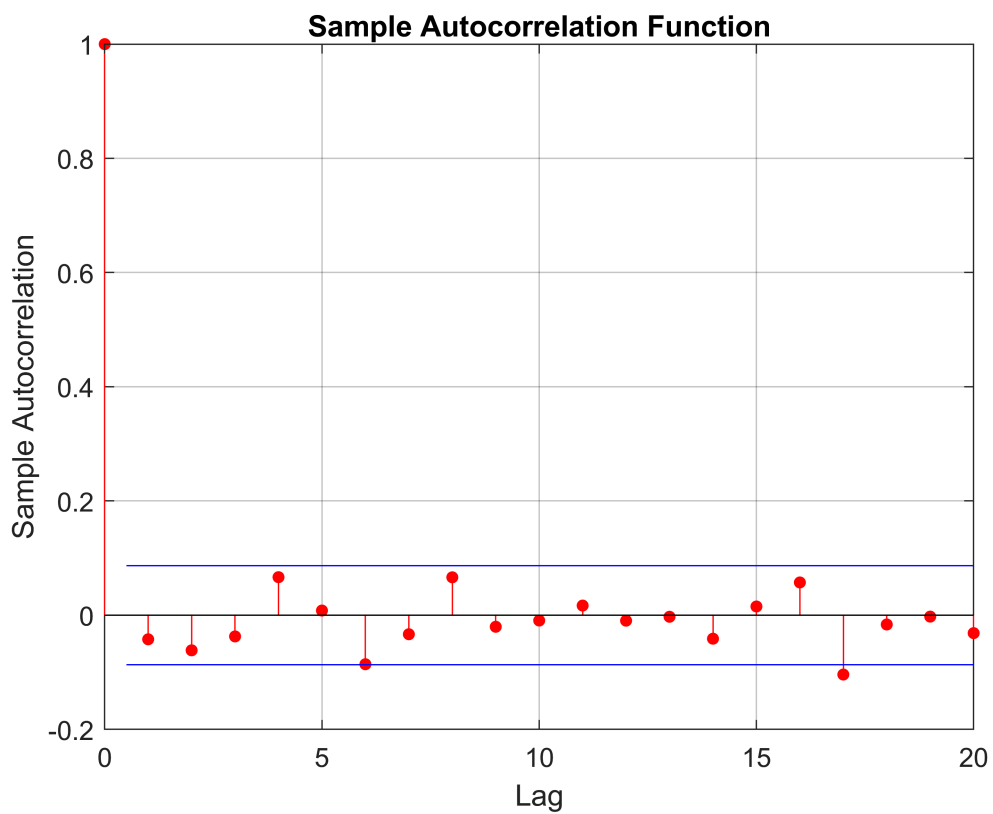
```
x1 = table(MktRF,FINANUSER);  
lm1 = fitlm(x1);  
resid = Md1.Residuals.Raw(~isnan(Md1.Residuals.Raw));  
normplot(resid)
```



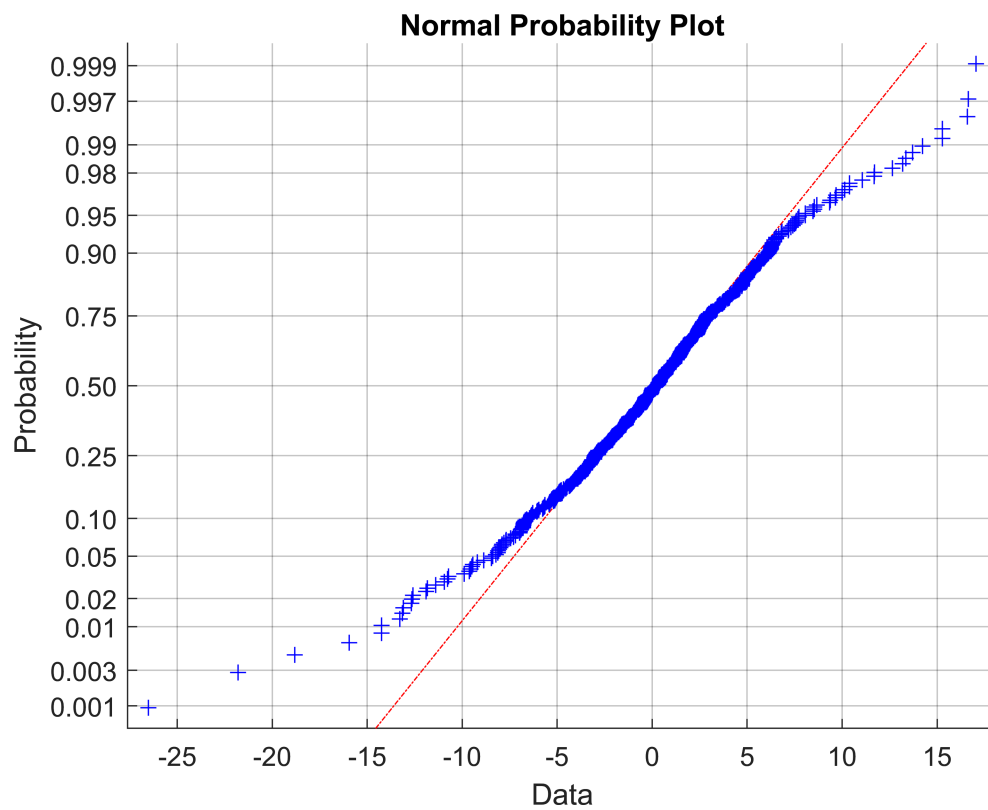
```
plotResiduals(lm1, 'fitted')
```



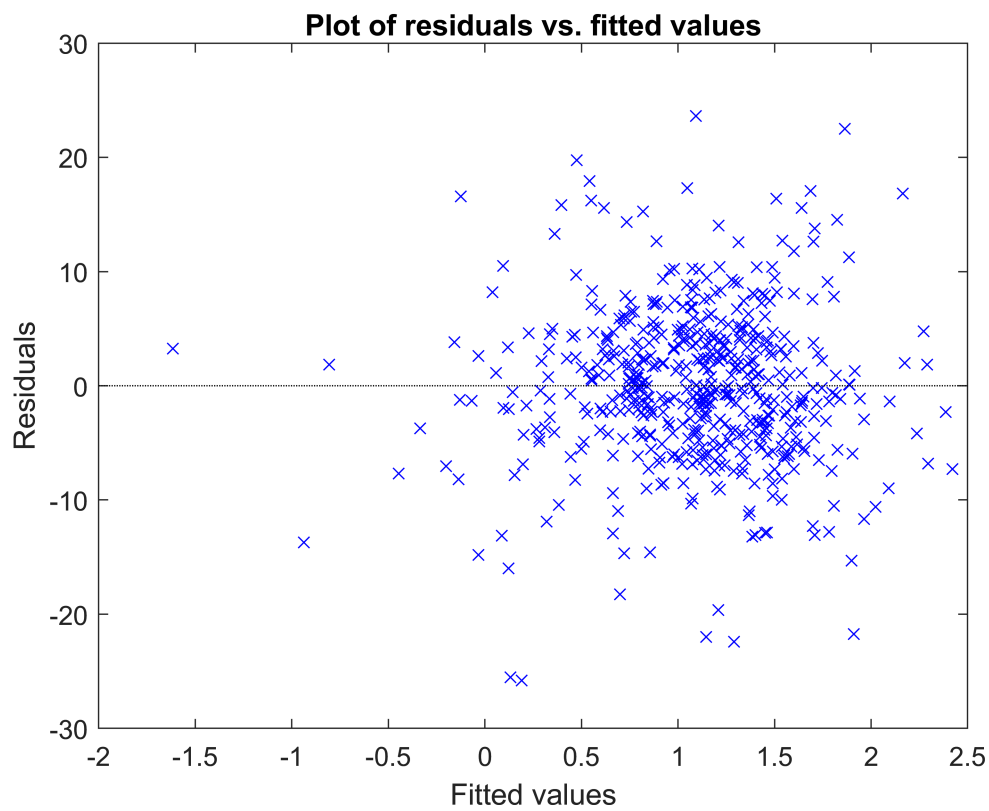
```
autocorr(resid)
```



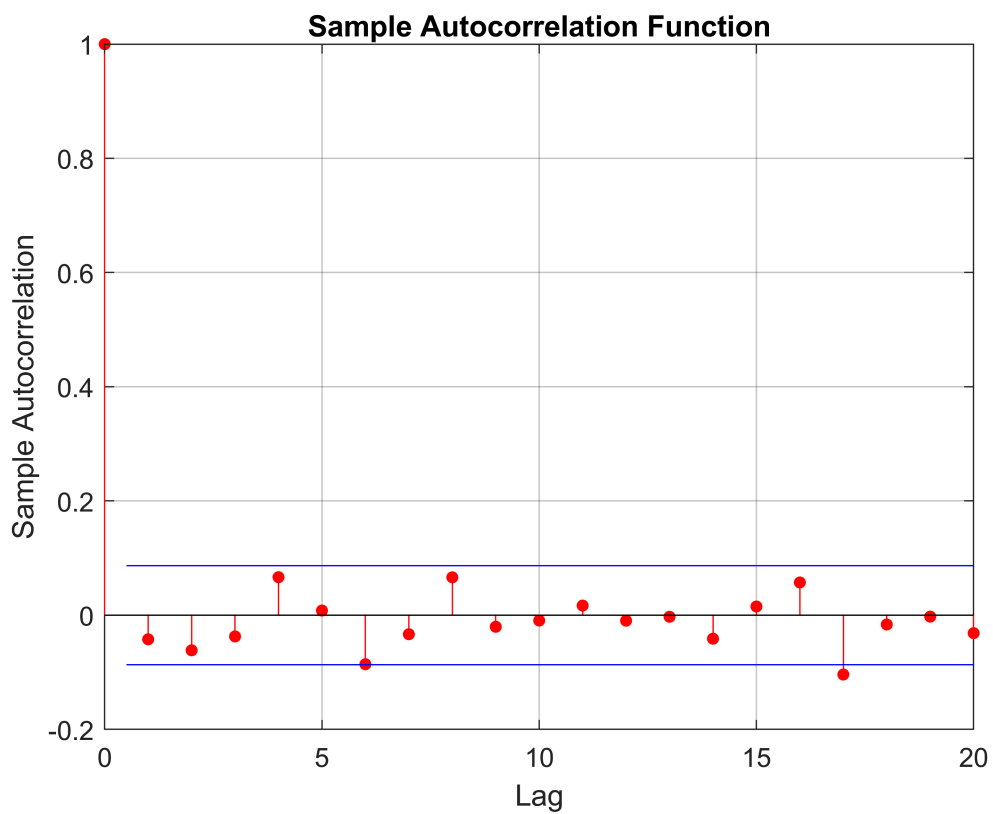
```
x1 = table(MktRF,TECNOUSER);  
lm1 = fitlm(x1);  
resid = Md1.Residuals.Raw(~isnan(Md1.Residuals.Raw));  
normplot(resid)
```



```
plotResiduals(lm1, 'fitted')
```

```
autocorr(resid)
```



Q3. (6 points) Compute the HAC standard errors for the CAPM regressions of INDUSUSER, FINANUSER, and TECNOUSER.

Are they different from the OLS standard errors?

% INSERT CODE

```
x1 = table(MktRF,INDUSUSER);
lm1 = fitlm(x1)
```

```
lm1 =
Linear regression model:
    INDUSUSER ~ 1 + MktRF
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.91565	0.23429	3.9081	0.00010506
MktRF	0.10704	0.052887	2.024	0.043473

Number of observations: 532, Error degrees of freedom: 530

Root Mean Squared Error: 5.34

R-squared: 0.00767, Adjusted R-Squared: 0.0058

F-statistic vs. constant model: 4.1, p-value = 0.0435

```
[EstCov,es,coeff] = hac(lm1)
```

```
Estimator type: HAC
Estimation method: BT
Bandwidth: 4.4884
Whitening order: 0
Effective sample size: 532
Small sample correction: on
```

Coefficient Covariances:

	Const	MktRF
Const	0.0510	-0.0050
MktRF	-0.0050	0.0036

EstCov = 2x2

```
0.0510    -0.0050
-0.0050    0.0036
```

es = 2x1

```
0.2258
0.0601
```

coeff = 2x1

```
0.9157
0.1070
```

```
x1 = table(MktRF,FINANUSER);
lm1 = fitlm(x1)
```

```
lm1 =
Linear regression model:
  FINANUSER ~ 1 + MktRF
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.80996	0.24665	3.2838	0.0010918
MktRF	0.14648	0.055677	2.6309	0.0087637

Number of observations: 532, Error degrees of freedom: 530
 Root Mean Squared Error: 5.63
 R-squared: 0.0129, Adjusted R-Squared: 0.011
 F-statistic vs. constant model: 6.92, p-value = 0.00876

```
[EstCov,es,coeff] = hac(lm1)
```

Estimator type: HAC
 Estimation method: BT
 Bandwidth: 6.1959
 Whitening order: 0
 Effective sample size: 532
 Small sample correction: on

Coefficient Covariances:

```

      |  Const  MktRF
-----
Const |  0.0555 -0.0034
MktRF | -0.0034  0.0029
EstCov = 2x2
    0.0555  -0.0034
   -0.0034   0.0029
es = 2x1
    0.2356
    0.0542
coeff = 2x1
    0.8100
    0.1465
```

```
x1 = table(MktRF,TECNOUSER);
lm1 = fitlm(x1)
```

```
lm1 =
Linear regression model:
  TECNOUSER ~ 1 + MktRF
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.0117	0.29754	3.4002	0.000724
MktRF	0.11309	0.067165	1.6838	0.092812

Number of observations: 532, Error degrees of freedom: 530
 Root Mean Squared Error: 6.79
 R-squared: 0.00532, Adjusted R-Squared: 0.00344
 F-statistic vs. constant model: 2.84, p-value = 0.0928

```
[EstCov,es,coeff] = hac(lm1)
```

```
Estimator type: HAC
Estimation method: BT
Bandwidth: 4.1440
Whitening order: 0
Effective sample size: 532
Small sample correction: on
```

Coefficient Covariances:

```
      |  Const   MktRF
-----
Const |  0.0836  -0.0055
MktRF | -0.0055   0.0049
EstCov = 2x2
    0.0836   -0.0055
   -0.0055    0.0049
es = 2x1
    0.2892
    0.0699
coeff = 2x1
    1.0117
    0.1131
```

Answer:

Q4. (2 points) Compute the correlation matrix of Mkt-RF, SMB, HML. Are these highly correlated?

```
% correlationn matrix
corrcoef(returns{ :,6:8})
```

```
ans = 3x3
    1.0000    0.0729    0.1283
    0.0729    1.0000    0.2407
    0.1283    0.2407    1.0000
```

Answer:

Q5. (20 points) Estimate a multivariate regression adding SMB and HML to the CAPM regressions for each of the 5 portfolios..

- Is the constant statistically significant?
- Are SMB and HML significant?
- Are there significant differences between the coefficient of MktRF in the original CAPM regression and in the multivariate regression?
- Any omitted variable bias?

`% insert code`

```
x1 = table(MktRF,SMB,HML, BMATRUSER);
lm1 = fitlm(x1)
```

```
lm1 =
Linear regression model:
    BMATRUSER ~ 1 + MktRF + SMB + HML
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.60822	0.2718	2.2377	0.025654
MktRF	0.065687	0.064067	1.0253	0.3057
SMB	0.35893	0.093689	3.8311	0.00014292
HML	0.25323	0.097814	2.5889	0.0098944

Number of observations: 532, Error degrees of freedom: 528
 Root Mean Squared Error: 6.14
 R-squared: 0.0371, Adjusted R-Squared: 0.0317
 F-statistic vs. constant model: 6.79, p-value = 0.00017

```
x2 = table(MktRF,SMB,HML,INDUSUSER);
lm1 = fitlm(x2)
```

```
lm1 =
Linear regression model:
    INDUSUSER ~ 1 + MktRF + SMB + HML
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.80926	0.23352	3.4654	0.00057247
MktRF	0.095806	0.055045	1.7405	0.082353
SMB	0.28649	0.080495	3.5591	0.00040564
HML	0.21224	0.084039	2.5254	0.011847

Number of observations: 532, Error degrees of freedom: 528
 Root Mean Squared Error: 5.27
 R-squared: 0.0377, Adjusted R-Squared: 0.0322
 F-statistic vs. constant model: 6.89, p-value = 0.000147

```
x3 = table(MktRF,SMB,HML,CNSMGUSER);
lm1 = fitlm(x3)
```

```
lm1 =
Linear regression model:
    CNSMGUSER ~ 1 + MktRF + SMB + HML
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.54264	0.22125	2.4527	0.014503
MktRF	0.01853	0.052151	0.35531	0.7225
SMB	0.33823	0.076263	4.4351	1.1206e-05
HML	0.2695	0.079621	3.3848	0.00076533

Number of observations: 532, Error degrees of freedom: 528

Root Mean Squared Error: 4.99

R-squared: 0.0489, Adjusted R-Squared: 0.0435

F-statistic vs. constant model: 9.04, p-value = 7.55e-06

```
x4 = table(MktRF,SMB,HML,FINANUSER);  
lm1 = fitlm(x4)
```

lm1 =

Linear regression model:

FINANUSER ~ 1 + MktRF + SMB + HML

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.69691	0.245	2.8446	0.0046198
MktRF	0.12495	0.05775	2.1637	0.030933
SMB	0.35042	0.08445	4.1494	3.8847e-05
HML	0.21306	0.088169	2.4165	0.01601

Number of observations: 532, Error degrees of freedom: 528

Root Mean Squared Error: 5.53

R-squared: 0.0493, Adjusted R-Squared: 0.0439

F-statistic vs. constant model: 9.13, p-value = 6.7e-06

```
x5 = table(MktRF,SMB,HML,TECNOUSER);  
lm1 = fitlm(x5)
```

lm1 =

Linear regression model:

TECNOUSER ~ 1 + MktRF + SMB + HML

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.87429	0.29713	2.9424	0.0033999
MktRF	0.10988	0.070039	1.5689	0.11728
SMB	0.31586	0.10242	3.0839	0.0021498
HML	0.28887	0.10693	2.7014	0.0071258

Number of observations: 532, Error degrees of freedom: 528

Root Mean Squared Error: 6.71

R-squared: 0.0317, Adjusted R-Squared: 0.0262

F-statistic vs. constant model: 5.76, p-value = 0.000696

Answer:
