

## **INTRODUCTION - UNIT ROOT TESTING IN R**

### **1. INTRODUCTION**

#### **1.1. Downloading and Installing R**

R is a widely used package for statistical analysis. The difference between R and many other statistical packages is that it is free software. R is used within a command - line interface. R is distributed by the “Comprehensive R Archive Network” (CRAN) - it is available from: <http://cran.r-project.org>. R can be installed by executing the downloaded file. The installation procedure is straightforward, one usually only has to specify the target directory in which to install R. After the installation, R can be started like any other application for Windows. That is, by double - clicking on the corresponding icon.

#### **1.2. Import Data in R**

Importing data into R can be carried out in various ways. Below, the command `read.table` is used:

```
mydatats1<-read.table("E:/Loukia/Teaching/TimeSeries/Notes-R/datats.txt")
y <- mydatats1$V1
```

Let us create a time series object using the function “`ts`” from a vector - single time-series or a matrix - multivariate time-series. The data consist of the Johnson & Johnson quarterly earnings per share from 1960:1 to 1980:4, i.e. 84 quarters.

```
j=ts(y, frequency=4, start = c(1960,1))
```

#### **1.3. Basic Plots in R**

A time series plot can be created using the `plot( )` command. The `plot( )` command allows for a number of optional arguments: the option `type="l"` sets the plot-type to “lines”, the option `lwd=2` (line width) controls the thickness of the plotted line, the option `col="red"` controls the colour used, the options `xlab` and `ylab` are used for labeling the axes, while `main` specifies the title of the plot.

For example, the time series plot for the Johnson & Johnson data is given by:

```
plot(j, type="l", col='red', lwd=1, main="Time Series plot of Johnson & Johnson data", ylab="Quarterly earnings per share")
```

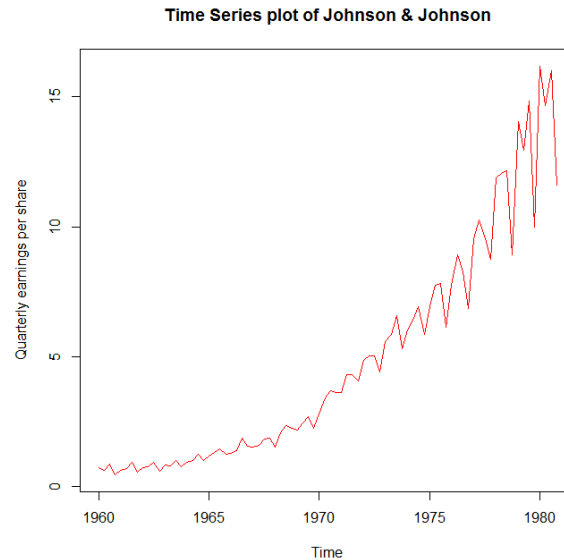


Figure 1: Time series plot of J&amp;J

Every R function has a corresponding help file, which can be accessed by typing a question mark and the command. It contains further details about the function and available options, references and examples of usage. For example, by typing `?plot` into the console opens the help file for the `plot()` command.

The `hist()` command creates a histogram of the data vector. Below, the time series plot of the J&J data, of the log of J&J and of the first differences of the log of J&J data, together with their corresponding histograms are presented. The log of J&J time series and the first differences of the log of J&J data are computed by:

```
lj=log(j)    # compute the logarithm of the J&J data
dlj=diff(lj) # compute the first differences of the log of J&J data
```

A nice graph can be produced by using the command `par(mfrow=c(rows,col))`, which splits the graph into (rows x col) subplots:

```
par(mfrow=c(3,2))    # set up the graphics
plot(j,type="l", col='red', lwd=1,main="Time Series plot of Johnson & Johnson", ylab="Quarterly earnings
per share")
hist(j, nclass=15, main="Histogram of Johnson & Johnson")
plot(lj,type="l", col='red', lwd=1,main="Log of Johnson & Johnson", ylab="Log of Quarterly earnings per
share")
hist(lj, nclass=15, main="Histogram of log of Johnson & Johnson")
```

```
plot(dlj,type="l", col='red', lwd=1,main="Differences of log of Johnson & Johnson")
hist(dlj, nclass=15, main="Histogram of differences of log of J&J")
```

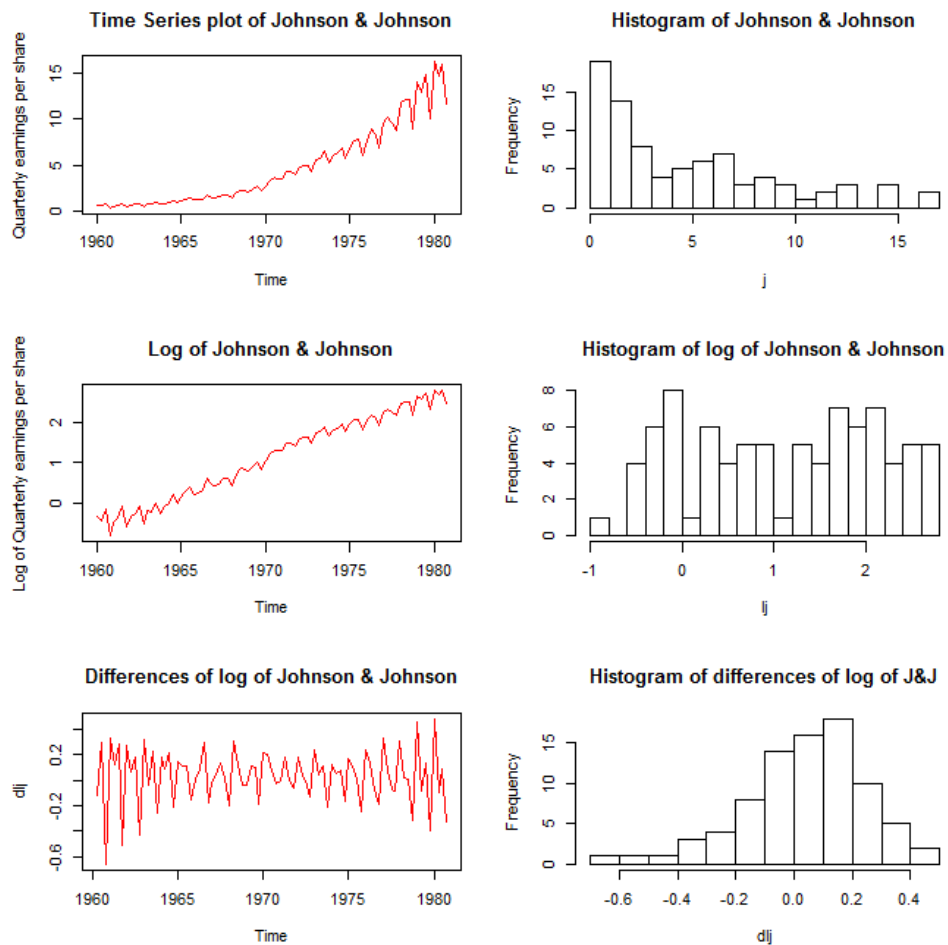


Figure 2: Time series plots and histograms for the J&J, log of J&J and the differences of log(J&J)

The autocorrelation and partial autocorrelation plots are useful to examine if there is dependence between lagged values of the analyzed series. The `acf()` command and the `pacf()` command create an autocorrelation and a partial autocorrelation plot, respectively. Below, the autocorrelation and partial autocorrelation plots of the J&J data, the log of J&J and of the first differences of the log of J&J data are presented using the command `par(mfrow=c(3,2))`.

```
par(mfrow=c(3,2)) # set up the graphics
acf(j, 48, main="ACF of J&J") # autocorrelation function plot
pacf(j, 48, main="PACF of J&J") # partial autocorrelation function
```

```
acf(lj, 48, main="ACF of log of J&J")
pacf(lj, 48, main="PACF of log of J&J")
acf(dlj, 48, main="ACF of differences of log of J&J")
pacf(dlj, 48, main="PACF of differences of log of J&J")
```

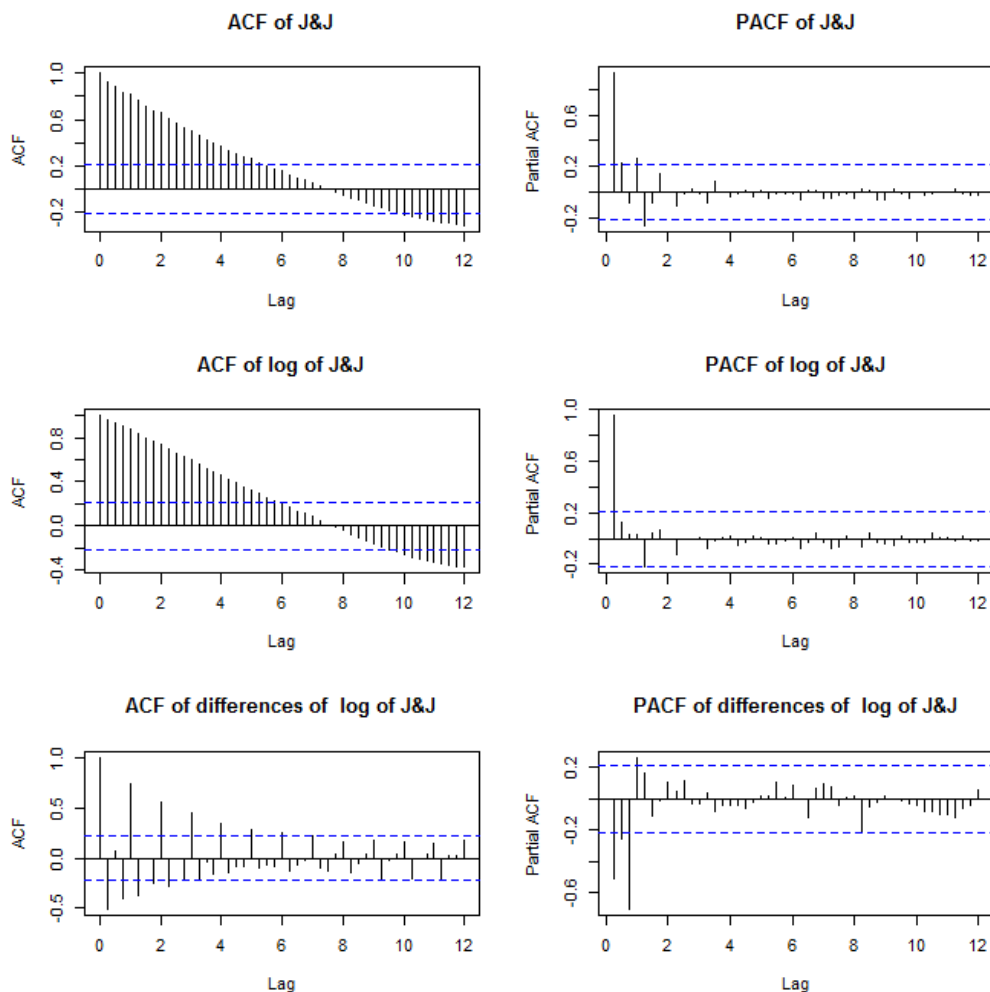


Figure 3: Autocorrelations (ACF) and partial autocorrelations (PACF) for the J&J, log of J&J and the differences of log(J&J)

Note that the lag values in the X axis are 1, 2, 3, 4, 5,... and correspond to lags 4, 8, 12, 16, 20,... because we have quarterly data, i.e. the frequency is 4. A better type of labeling can be produced by using the following set of commands:

```
par(mfrow=c(3,2)) # set up the graphics
```

```

acf(ts(j,freq=1), 48, main="ACF of J&J")    # autocorrelation function plot
pacf(ts(j,freq=1), 48, main="PACF of J&J")  # partial autocorrelation function plot
acf(ts(lj,freq=1), 48, main="ACF of log of J&J")
pacf(ts(lj,freq=1), 48, main="PACF of log of J&J")
acf(ts(dlj,freq=1), 48, main="ACF of differences of log of J&J")
pacf(ts(dlj,freq=1), 48, main="PACF of differences of log of J&J")

```

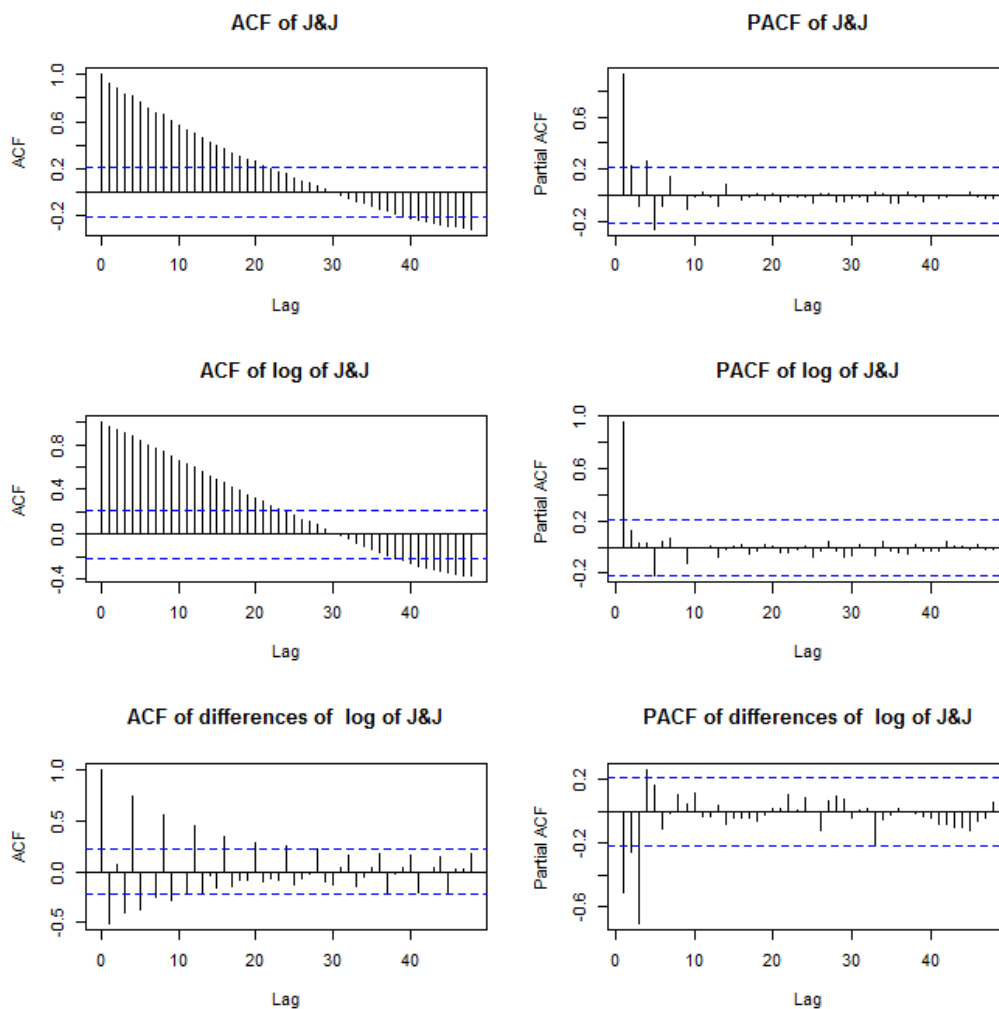


Figure 4: Autocorrelations (ACF) and partial autocorrelations (PACF) for the J&J, log of J&J and the differences of log(J&J)

Note that in the autocorrelation plots presented above, the dashed lines are the approximate two standard error confidence bounds computed by  $\pm 1.96 \cdot \left(1/\sqrt{T}\right)$ , where T is the number observations. If

the autocorrelation is within these bounds, it is not significantly different from zero at (approximately) 5% level of significance [Bartlett test].

The `Box.test()` command can be used to compute the Box-Pierce or the Ljung-Box test statistic for examining the null hypothesis that the autocorrelations of a given time series are zero. The command is: `Box.test(x, lag, type = c("Box-Pierce", "Ljung-Box"))`, where “x” is the analyzed time series, “lag” denotes the number of lags at which the statistic will be computed, while “type” determines the Box-Pierce or the Ljung-Box test statistic. For example, by running the commands

```
res1=Box.test(j,48,type="Box-Pierce")
res2=Box.test(j,48,type="Ljung-Box")
```

will give the following results:

```
res1: Box-Pierce test
data: j
X-squared = 695.06, df = 48, p-value < 2.2e-16

res2: Box-Ljung test
data: j
X-squared = 853.09, df = 48, p-value < 2.2e-16
```

## 2. UNIT ROOT TESTING

To perform a unit-root test, the command `ur.df(y, type = c("none", "drift", "trend"), lags = 1, selectlags = c("Fixed", "AIC", "BIC"))` can be used; “y” is the time series to be tested for a unit root, “type” corresponds to the three fitted models, i.e. a model without constant/trend (none), a model with constant only (drift), and a model with constant and time trend (trend). “lags” denotes the maximum number of lags for endogenous variable to be included, `selectlags` denotes the lag selection which can be achieved according to the Akaike "AIC" or the Bayes "BIC" information criteria.

First, the package “urca” is installed (or is loaded if it is already installed):

```
install.packages("urca")
library(urca)
```

Example 1: Unit root testing to the J&J time series

First, we can fit an autoregressive time series model to the J&J data, by selecting the complexity of the model based on AIC. Then, we perform an augmented Dickey-Fuller test of unit root, based on a model with constant and trend (see figure 2, time series plot for the J&J series):

```
m=ar(j)
m
m$order
m1=ur.df(j,type="trend",lags=m$order-1)
m1
summary(m1)
```

The results taken from R are presented below:

[m1](#)

```
#####
# Augmented Dickey-Fuller Test Unit Root / Cointegration Test #
#####
The value of the test statistic is: 1.9321 16.7049 19.2758
```

[summary\(m1\)](#)

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
Test regression trend
Call:
lm(formula = z.diff ~ z.lag.1 + 1 + tt + z.diff.lag)
Residuals:
    Min     1Q   Median     3Q    Max
-1.27266 -0.17348  0.01381  0.12299  1.18302
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) -0.088498  0.136166  -0.650  0.5178
z.lag.1      0.073336  0.037956   1.932  0.0573 .
tt           0.010052  0.006304   1.595  0.1152
z.diff.lag1 -1.069854  0.131507  -8.135 8.57e-12 ***
z.diff.lag2 -1.012388  0.145806  -6.943 1.41e-09 ***
z.diff.lag3 -1.006500  0.143949  -6.992 1.14e-09 ***
z.diff.lag4  0.092346  0.141368   0.653  0.5157
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.4127 on 72 degrees of freedom
Multiple R-squared:  0.9262,    Adjusted R-squared:  0.92
F-statistic: 150.5 on 6 and 72 DF, p-value: < 2.2e-16
Value of test-statistic is: 1.9321 16.7049 19.2758
Critical values for test statistics:
    1pct 5pct 10pct
tau3 -4.04 -3.45 -3.15
```

phi2	6.50	4.88	4.16
phi3	8.73	6.49	5.47

Similar results are obtained by using EViews:

Null Hypothesis: J has a unit root				
Exogenous: Constant, Linear Trend				
Lag Length: 4 (Fixed)				
			t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic			1.932145	1.0000
Test critical values:	1% level		-4.078420	
	5% level		-3.467703	
	10% level		-3.160627	
*MacKinnon (1996) one-sided p-values.				
Augmented Dickey-Fuller Test Equation				
Dependent Variable: D(J)				
Method: Least Squares				
Sample (adjusted): 1961Q2 1980Q4				
Included observations: 79 after adjustments				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
J(-1)	0.073336	0.037956	1.932145	0.0573
D(J(-1))	-1.069854	0.131507	-8.135353	0.0000
D(J(-2))	-1.012388	0.145806	-6.943389	0.0000
D(J(-3))	-1.006500	0.14949	-6.992067	0.0000
D(J(-4))	0.092346	0.141368	0.653236	0.5157
C	-0.088498	0.136166	-0.649925	0.5178
@TREND(1960Q1)	0.010052	0.006304	1.594602	0.1152

Obviously, the null hypothesis of non-stationarity for the J&J series is not rejected. Thus the J&J is not stationary. Next, we can test for a unit root for the logarithms of the J&J. We perform an augmented Dickey-Fuller test of unit root, based on a model with constant and trend (see figure 2, time series plot for the log(J&J) series):

```
#Unit root testing for the log(j)
m1=ur.df(lj,type="trend",lags=5)
m1
```



## summary(m1)

The results taken from R are presented below:

m1

```
#####
# Augmented Dickey-Fuller Test Unit Root / Cointegration Test #
#####
The value of the test statistic is: -1.4369 6.8046 1.2869
```

summary(m1)

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
Test regression trend
Call:
lm(formula = z.diff ~ z.lag.1 + 1 + tt + z.diff.lag)
Residuals:
    Min     1Q   Median     3Q    Max
-0.18785 -0.04923 -0.00168  0.04606  0.20635
Coefficients:
      Estimate Std. Error t value Pr(>|t|)
(Intercept) -0.037988  0.105676  -0.359 0.720323
z.lag.1      -0.179098  0.124646  -1.437 0.155214
tt           0.007347  0.005337   1.377 0.173026
z.diff.lag1 -0.589245  0.158531  -3.717 0.000403 ***
z.diff.lag2 -0.416588  0.164498  -2.532 0.013571 *
z.diff.lag3 -0.429622  0.143892  -2.986 0.003896 **
z.diff.lag4  0.380827  0.133094   2.861 0.005558 **
z.diff.lag5  0.154810  0.106817   1.449 0.151721
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.08266 on 70 degrees of freedom
Multiple R-squared:  0.8382,    Adjusted R-squared:  0.822
F-statistic: 51.8 on 7 and 70 DF, p-value: < 2.2e-16
Value of test-statistic is: -1.4369 6.8046 1.2869
Critical values for test statistics:
    1pct 5pct 10pct
tau3 -4.04 -3.45 -3.15
phi2  6.50  4.88  4.16
phi3  8.73  6.49  5.47
```

Similar results are obtained by using EVIEWS:

Null Hypothesis: LJ has a unit root Exogenous: Constant, Linear Trend Lag Length: 5 (Automatic based on AIC, MAXLAG=11)				
			t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic			-1.436854	0.8423
Test critical values:				
	1% level		-4.080021	
	5% level		-3.468459	
	10% level		-3.161067	
*MacKinnon (1996) one-sided p-values.				
Augmented Dickey-Fuller Test Equation				
Dependent Variable: D(LJ)				
Method: Least Squares				
Sample (adjusted): 1961Q3 1980Q4				
Included observations: 78 after adjustments				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
LJ(-1)	-0.179098	0.124646	-1.436854	0.1552
D(LJ(-1))	-0.589245	0.158531	-3.716900	0.0004
D(LJ(-2))	-0.416588	0.164498	-2.532476	0.0136
D(LJ(-3))	-0.429622	0.143892	-2.985722	0.0039
D(LJ(-4))	0.380827	0.133094	2.861342	0.0056
D(LJ(-5))	0.154810	0.106817	1.449295	0.1517
C	-0.037988	0.105676	-0.359474	0.7203
@TREND(1960Q1)	0.007347	0.005337	1.376592	0.1730

The null hypothesis of non-stationarity for the log(J&J) series is not rejected. Thus the log(J&J) is not stationary. Note, however, that if the number of the lagged variables in the augmented Dickey-Fuller model changes (for example, if the order is 1), the result of the unit root test is completely different. Maybe alternative models or more powerful unit root techniques could be used.

Next, we can test for a unit root for the differences of logarithms of the J&J. We perform an augmented Dickey-Fuller test of unit root, based on a model with constant (see figure 2, time series plot for the differences of log(J&J) series):

```
m=ar(dlj)
m
m$order
m1=ur.df(dlj,type="drift",lags=5)
```

```
m1
```

```
summary(m1)
```

The results taken from R are presented below:

```
m1
```

```
#####
# Augmented Dickey-Fuller Test Unit Root / Cointegration Test #
#####
The value of the test statistic is: -4.317 9.3192
```

```
summary(m1)
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
Test regression drift
Call:
lm(formula = z.diff ~ z.lag.1 + 1 + z.diff.lag)
Residuals:
    Min     1Q   Median     3Q    Max
-0.189424 -0.049737 -0.008657  0.048398  0.207341
Coefficients:
      Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.10117   0.02537   3.988 0.000162 ***
z.lag.1     -2.53130   0.58635  -4.317 5.11e-05 ***
z.diff.lag1  0.80961   0.53999   1.499 0.138290
z.diff.lag2  0.32762   0.45641   0.718 0.475258
z.diff.lag3 -0.22525   0.32533  -0.692 0.490988
z.diff.lag4  0.06502   0.21591   0.301 0.764209
z.diff.lag5  0.12319   0.10708   1.150 0.253900
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.08322 on 70 degrees of freedom
Multiple R-squared:  0.9438,    Adjusted R-squared:  0.939
F-statistic: 196.1 on 6 and 70 DF, p-value: < 2.2e-16
Value of test-statistic is: -4.317 9.3192
Critical values for test statistics:
      1pct 5pct 10pct
tau2 -3.51 -2.89 -2.58
phi1  6.70  4.71  3.86
```

Similar results are obtained by using EVIEWS:

Null Hypothesis: DLJ has a unit root				
Exogenous: Constant				
Lag Length: 4 (Automatic based on AIC, MAXLAG=11)				
			t-Statistic	Prob.*
Augmented Dickey-Fuller test statistic			-4.569037	0.0004
Test critical values:	1% level		-3.516676	
	5% level		-2.899115	
	10% level		-2.586866	
*MacKinnon (1996) one-sided p-values.				
Dependent Variable: D(DLJ)				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
DLJ(-1)	-2.338380	0.511788	-4.569037	0.0000
D(DLJ(-1))	0.600855	0.447643	1.342263	0.1837
D(DLJ(-2))	0.054690	0.323307	0.169157	0.8661
D(DLJ(-3))	-0.460666	0.207464	-2.220465	0.0295
D(DLJ(-4))	-0.126099	0.105703	-1.192958	0.2368
C	0.094353	0.022428	4.206999	0.0001

Therefore the differences of  $\log(J\&J)$  seem to be a stationary process.

Alternatively, other R commands can be used. The R package 'tseries' is used to test the null hypothesis that a series has a unit root, versus the alternative hypothesis that the process is stationary. We test the null hypothesis using the available Dickey-Fuller (DF), Augmented Dickey-Fuller (ADF) and Phillips-Perron (PP) tests; note that in each case, the general regression equation incorporates a constant and a linear trend. In the ADF test, the default number of AR components included in the model, say  $k$ , is  $\lceil [(n-1)^{1/3}] \rceil$ , which corresponds to the suggested upper bound on the rate at which the number of lags,  $k$ , should be made to grow with the sample size for the general ARMA( $p,q$ ) setup. For the PP test, the default value of  $k$  is  $\lceil [0.04n^{1/4}] \rceil$ . The distinction between the two tests (DF and PP) is that the Phillips-Perron procedure estimates the autocorrelations in the stationary process directly (using a kernel smoother) rather than assuming an AR approximation, and for this reason the Phillips-Perron test is described as semi-parametric. To implement the above unit root tests, first, we load the package 'tseries' by using the command `library(tseries)`, and then we perform the unit root test under consideration by using one of the following commands: `adf.test(y, k=0)` for the DF test, `adf.test(y)` for the ADF test, or the command `pp.test(y)` for the PP test. For example:

```
library(tseries)
```

```
adf.test(y,k=4)
```

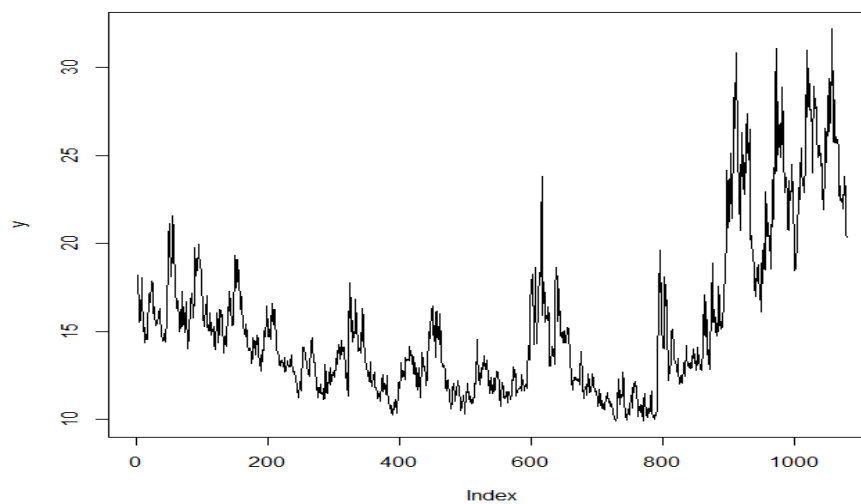
Augmented Dickey-Fuller Test, data:dlj,

Dickey-Fuller = -4.5649, Lag order = 4, p-value = 0.01, alternative hypothesis: stationarity.

Example 2: Unit root testing to the VIX time series

Consider the VIX of CBOE from 2004 to 2008. The data are obtained from the CBOE web site. First, we read the data, and obtain a time series plot:

```
data<- read.table("E:/Loukia/Teaching/TimeSeries/Notes-R/vix08.txt",header=T)
dim(data)
data[1,]
y=data[,7]
plot(y,type="l")
```



Fit an AR model and run the Augmented Dickey-Fuller test:

```
m=ar(y)
m
m$order
m1=ur.df(y,type="drift",lags=10)
```

```
m1
```

```
summary(m1)
```

```
m1
```

```
#####
# Augmented Dickey-Fuller Test Unit Root / Cointegration Test #
#####
The value of the test statistic is: -2.2726 2.6163
```

```
summary(m1)
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
Test regression drift
Call:
lm(formula = z.diff ~ z.lag.1 + 1 + z.diff.lag)
Residuals:
    Min     1Q  Median     3Q    Max
-5.6122 -0.5330 -0.1247  0.4036  7.1974
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.274964  0.121815  2.257  0.0242 *
z.lag.1      -0.017346  0.007633 -2.273  0.0233 *
z.diff.lag1  -0.193620  0.031144 -6.217 7.29e-10 ***
z.diff.lag2  -0.108027  0.031777 -3.400  0.0007 ***
z.diff.lag3  -0.011997  0.031538 -0.380  0.7037
z.diff.lag4  -0.051623  0.031477 -1.640  0.1013
z.diff.lag5  -0.027070  0.031422 -0.862  0.3892
z.diff.lag6  -0.064092  0.031399 -2.041  0.0415 *
z.diff.lag7  -0.043289  0.031373 -1.380  0.1679
z.diff.lag8  -0.145963  0.031360 -4.654 3.66e-06 ***
z.diff.lag9   0.017763  0.031417  0.565  0.5719
z.diff.lag10  0.061834  0.030741  2.011  0.0445 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.089 on 1057 degrees of freedom
Multiple R-squared:  0.08152,    Adjusted R-squared:  0.07196
F-statistic: 8.529 on 11 and 1057 DF, p-value: 1.48e-14
Value of test-statistic is: -2.2726 2.6163
Critical values for test statistics:
    1pct 5pct 10pct
tau2 -3.43 -2.86 -2.57
phi1  6.43  4.59  3.78
```

```
m2=ur.df(y,type="trend",lags=10)
```

```
m2
```

```
summary(m2)
```

```
m2
```

```
#####
# Augmented Dickey-Fuller Test Unit Root / Cointegration Test #
#####
The value of the test statistic is: -2.7422 2.5997 3.8655
```

```
summary(m2)
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
Test regression trend
Call:
lm(formula = z.diff ~ z.lag.1 + 1 + tt + z.diff.lag)
Residuals:
    Min     1Q  Median     3Q    Max
-5.6281 -0.5331 -0.1150  0.4262  7.1234
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept)  0.2603461  0.1220682   2.133 0.033172 *
z.lag.1      -0.0231909  0.0084572  -2.742 0.006207 **
tt           0.0001916  0.0001198   1.600 0.109987
z.diff.lag1  -0.1902722  0.0311915  -6.100 1.49e-09 ***
z.diff.lag2  -0.1054542  0.0317940  -3.317 0.000942 ***
z.diff.lag3  -0.0100074  0.0315391  -0.317 0.751078
z.diff.lag4  -0.0496545  0.0314776  -1.577 0.114991
z.diff.lag5  -0.0252114  0.0314203  -0.802 0.422507
z.diff.lag6  -0.0622962  0.0313956  -1.984 0.047488 *
z.diff.lag7  -0.0417138  0.0313650  -1.330 0.183823
z.diff.lag8  -0.1446717  0.0313477  -4.615 4.41e-06 ***
z.diff.lag9   0.0184760  0.0313970   0.588 0.556347
z.diff.lag10  0.0622955  0.0307192   2.028 0.042821 *
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.088 on 1056 degrees of freedom
Multiple R-squared:  0.08374,    Adjusted R-squared:  0.07333
F-statistic: 8.043 on 12 and 1056 DF, p-value: 1.422e-14
Value of test-statistic is: -2.7422 2.5997 3.8655
Critical values for test statistics:
    1pct  5pct 10pct
tau3 -3.96 -3.41 -3.12
phi2  6.09  4.68  4.03
phi3  8.27  6.25  5.34
```

```
m3=ur.df(y,type="none",lags=10)
```

```
m3
```

```
summary(m3)
```

```
m3
```

```
#####
# Augmented Dickey-Fuller Test Unit Root / Cointegration Test #
#####
The value of the test statistic is: -0.3701
```

```
summary(m3)
```

```
#####
# Augmented Dickey-Fuller Test Unit Root Test #
#####
Test regression none
Call:
lm(formula = z.diff ~ z.lag.1 - 1 + z.diff.lag)
Residuals:
  Min   1Q Median   3Q   Max
-5.7480 -0.4964 -0.0723  0.4372  7.2926
Coefficients:
      Estimate Std. Error t value Pr(>|t|)
z.lag.1  -0.000774  0.002091  -0.370 0.711372
z.diff.lag1 -0.205749  0.030736  -6.694 3.52e-11 ***
z.diff.lag2 -0.119423  0.031434  -3.799 0.000153 ***
z.diff.lag3 -0.022077  0.031280  -0.706 0.480471
z.diff.lag4 -0.061325  0.031242  -1.963 0.049922 *
z.diff.lag5 -0.035995  0.031232  -1.152 0.249386
z.diff.lag6 -0.072692  0.031227  -2.328 0.020107 *
z.diff.lag7 -0.051163  0.031238  -1.638 0.101756
z.diff.lag8 -0.153488  0.031243  -4.913 1.04e-06 ***
z.diff.lag9  0.011601  0.031359   0.370 0.711506
z.diff.lag10 0.056654  0.030714   1.845 0.065378 .
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.091 on 1058 degrees of freedom
Multiple R-squared:  0.07711,    Adjusted R-squared:  0.06752
F-statistic: 8.037 on 11 and 1058 DF, p-value: 1.406e-13
Value of test-statistic is: -0.3701
Critical values for test statistics:
  1pct 5pct 10pct
tau1 -2.58 -1.95 -1.62
```