Homework1

(Question1)

This is the development path of memory of PCs

How the memory of pc changes over the last 30 years
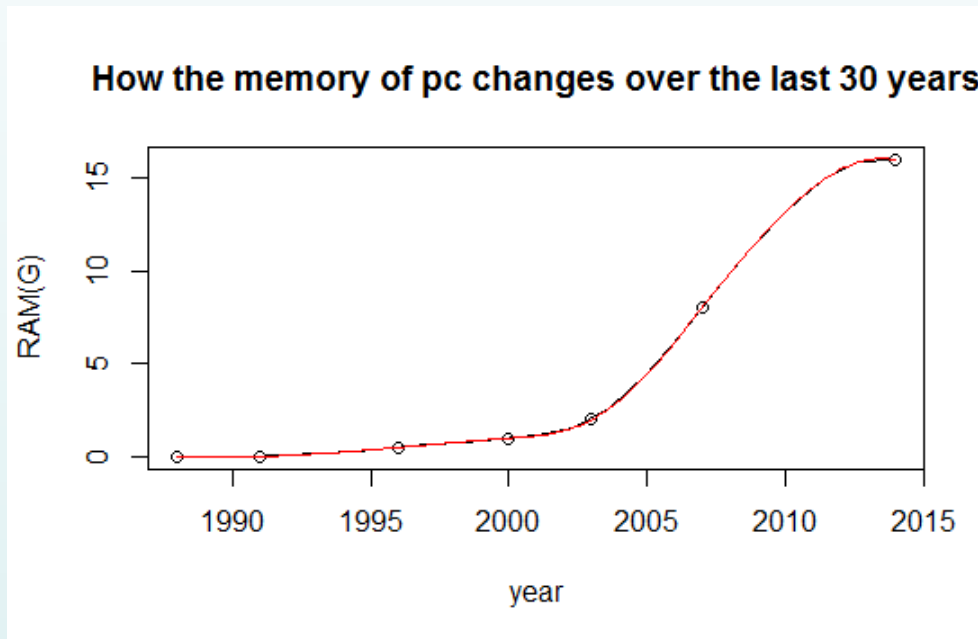
Corresponding code

```
year<c(1988,1991,1996,2000,2003,2007
,2014)
RAM<-c(0.002,0.004,0.5,1,2,8,16)
plot(year,RAM,ylab = "")
title(main ="How the memory of pc
changes over the last 30 years",ylab =
"RAM(G)")
lines(spline(year,RAM))
lines(spline(year,RAM, n = 201), col = 2)
```

Twenty years ago,the memory of pc is too small to support the analysis of FMRY

Homework1
(Question2)

Logistic regression:

In statistics, logistic regression, or logit regression, or logit model[1] is a regression model where the dependent variable (DV) is categorical. This article covers the case of a binary dependent variable—that is, where the output can take only two values, "0" and "1", which represent outcomes such as pass/fail, win/lose, alive/dead or healthy/sick. Cases where the dependent variable has more than two outcome categories may be analysed in multinomial logistic regression, or, if the multiple categories are ordered, in ordinal logistic regression.[2] In the terminology of economics, logistic regression is an example of a qualitative response/discrete choice model.

Homework1

(Question2)

A example:Probability of passing an exam versus hours of study

Suppose we wish to answer the following question:

A group of 20 students spend between 0 and 6 hours studying for an exam. How does the number of hours spent studying affect the probability that the student will pass the exam?

The reason for using logistic regression for this problem is that the dependent variable pass/fail represented by "1" and "0" are not cardinal numbers. If the problem was changed so that pass/fail was replaced with the grade 0–100 (cardinal numbers), then simple regression analysis could be used.

The table shows the number of hours each student spent studying, and whether they passed (1) or failed (0)

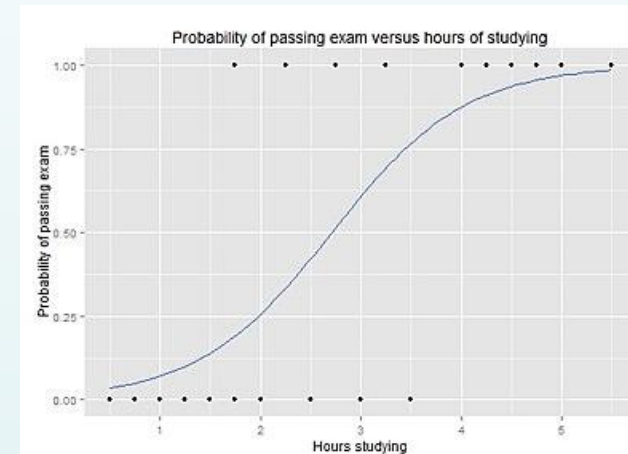| Hours	0.50	0.75	1.00	1.25	1.50	1.75	1.75	2.00	2.25	2.50	2.75	3.00	3.25	3.50	4.00	4.25	4.50	4.75	5.00	5.50
Pass	0	0	0	0	0	0	1	0	1	0	1	0	1	0	1	1	1	1	1	1

Homework1
(Question2)

The graph shows the probability of passing the exam versus the number of hours studying, with the logistic regression curve fitted to the data. The logistic regression analysis gives the following output.

	Coefficient	Std. Error	z-value	P-value (Wald)
Intercept	−4.0777	1.7610	−2.316	0.0206
Hours	1.5046	0.6287	2.393	0.0167

Probability of passing exam versus hours of studying

Graph of a logistic regression curve showing probability of passing an exam versus hours studying

$$\text{Probability of passing exam} = \frac{1}{1 + \exp\left(-\left(1.5046 \cdot \text{Hours} - 4.0777\right)\right)}$$

Homework2
(Q1)

```
Code:
x = 6
n = 1000
lambda = 2
p = lambda / n
dbinom (x,2*n,p) # binomial probability mass function
dpois (x, 2*lambda ) # Poisson probability mass function
dpois (0, 5 )
```

Homework2
(Q2)

Code:
```
plot(year,number,type = "b",
    col="black",main = "The history of computer memory",
    sub = "This is the change in the number of types of computer memory",
    xlab = "year",ylab = "The number of memory")
barplot(number,
      xlab = "year",ylab ="The number of memory ")
```

Homework2
(Q3)

Code:
```
lambda=2
x=seq(0:6)
P<-data.frame(dpois(x,lambda))
sum<-(P[1,]+P[7,]+P[2,]+P[6,]+P[3,]+P[5,]+P[4,]+P[4,])
sum
lambda=5
x=0
dpois(x,lambda)
```

Homework3

(Q1)

```
library("digest")
# now do the hash code calculation
digest("I learn a lot from this class when I am proper listening to
the professor")
digest("I do not learn a lot from this class when I am absent and
playing on my Iphone")
```

What Is DSA (Digital Signature Algorithm)?

- Digital signatures are essential to **verify the sender of a document's identity.** The signature is computer using a set of rules and algorithm such that the identity of the person can be verified.
- The signature is generated by the use of **a private key** that known only to **the user.** The signature is verified when a public key is corresponds to the private key. With every user having a public/private key pair, this is an example of public-key cryptography.
- Public keys, which are known by everyone, can be used to verify the signature of a user. The private key, which is never shared, is used in signature generation, which can only be done by the user.

What can DSA do?

- Digital signatures are used to detect unauthorized modifications to data. Also, the recipient of a digitally signed document in proving to a third party that the document was indeed signed by the person who it is claimed to be signed by. This is known as nonrepudiation, because the person who signed the document cannot repudiate the signature at a later time.
- Digital signature algorithms can be used in e-mails, electronic funds transfer, electronic data interchange, software distribution, data storage, and just about any application that would need to assure the integrity and originality of dat
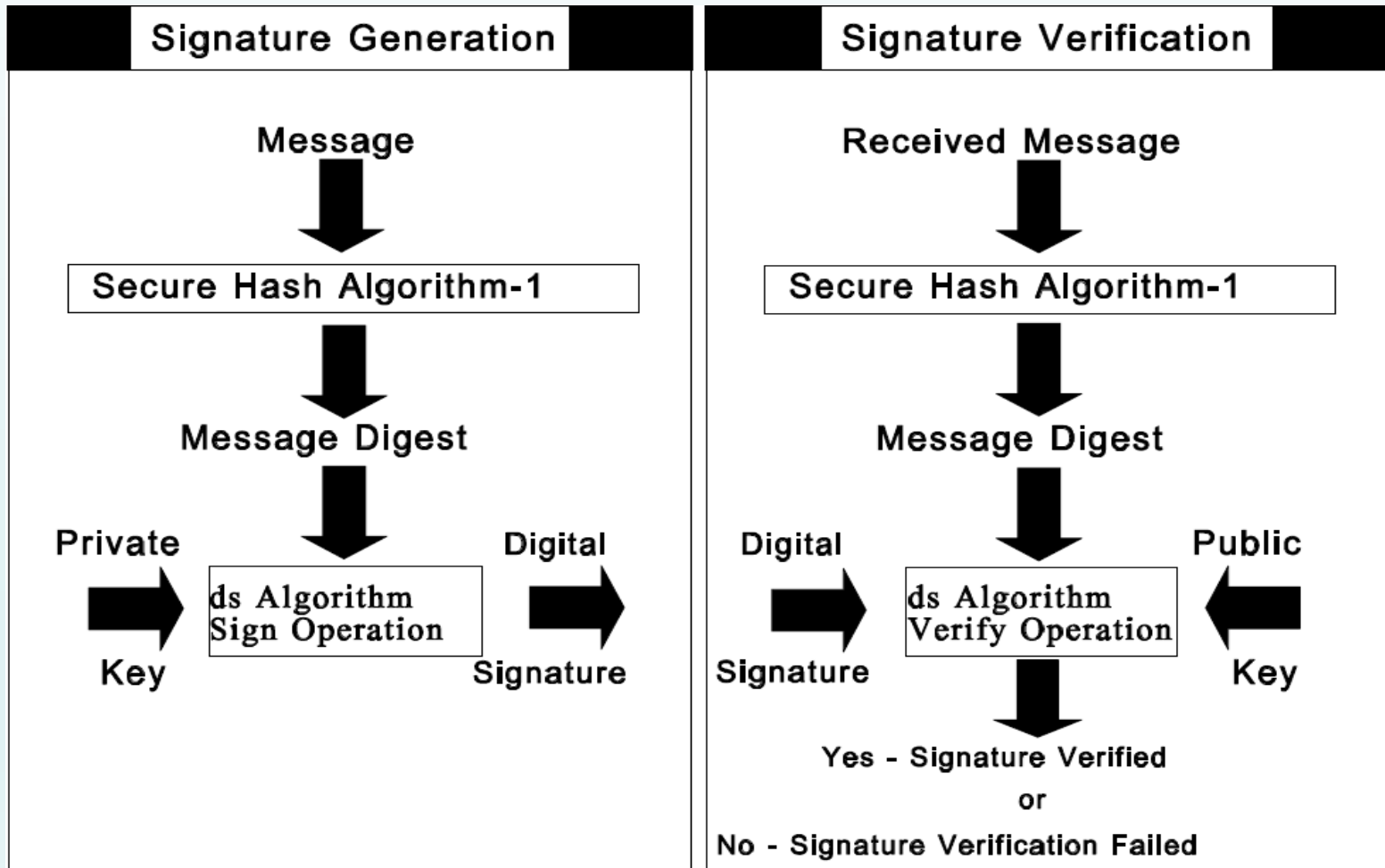
Homewor3
(Q2)

The first part of the DSA algorithm is the public key and private key generation

- Choose a prime number q, which is called the prime divisor.

- Choose another primer number p, such that p-1 mod q = 0. p is called the prime modulus.

- Choose an integer g, such that $1 < g < p$, $g^{**}q$ mod p = 1 and g = $h^{**}((p-1)/q)$ mod p. q is also called g's multiplicative order modulo p.

- Choose an integer, such that $0 < x < q$.

- Compute y as $g^{**}x$ mod p.

- Package the public key as {p,q,g,y}.

- Package the private key as {p,q,g,x}.

Homewor3
(Q3)

R-code:

```
>library(RJSONIO)
> letter<-LETTERS[1:10]
>country<-c("China","the US","the UK","Russia",
            "Korea","Japan","Italy","Brazil","India","Germany")
> data<-data.frame(letter,country)
> da<-as.matrix(data)
>cat(toJSON(da))
```

Homewor3
(Q3)

```
[{
 "letter": "A",
 "country": "China"
},

{
 "letter": "B",
 "country": "the US"
},

{
 "letter": "C",
 "country": "the UK"
},

{
 "letter": "D",
 "country": "Russia"
},

{
 "letter": "E",
 "country": "Korea"
},
```

```
{
 "letter": "F",
 "country": "Japan"
},

{
 "letter": "G",
 "country": "Italy"
},

{
 "letter": "H",
 "country": "Brazil"
},

{
 "letter": "I",
 "country": "India"
},

{
 "letter": "J",
 "country": "Germany"
}]
```

Homewor3(Q4)

```
Answer 4
rm(list = ls(all = TRUE))
graphics.off()
# install and load packages #
libraries = c("zoo", "tseries")
lapply(libraries, function(x) if (!(x %in% installed.packages())) {install.packages(x)})
lapply(libraries, library, quietly = TRUE, character.only = TRUE)
# load dataset #
load(file = "C:/Users/xiumei/Desktop/big data/crix.RData")
ret = diff(log(crix))
# d order #
Box.test(ret, type = "Ljung-Box", lag = 20)
# stationary test #
adf.test(ret, alternative = "stationary")
kpss.test(ret, null = "Trend")
par(mfrow = c(1, 2))
# acf plot #
autocorr = acf(ret, lag.max = 20, ylab = "Sample Autocorrelation", main = NA, lwd = 2, ylim = c(-0.3, 1))
# LB test of linear dependence #
print(cbind(autocorr$lag, autocorr$acf))
Box.test(ret, type = "Ljung-Box", lag = 1, fitdf = 0)
Box.test(autocorr$acf, type = "Ljung-Box")
 # plot of pacf #
autopcorr = pacf(ret, lag.max = 20, ylab = "Sample Partial Autocorrelation", main = NA, ylim = c(-0.3, 0.3), lwd = 2)
print(cbind(autopcorr$lag, autopcorr$acf))
# arima model#
par(mfrow = c(1, 1))
auto.arima(ret)
fit1 = arima(ret, order = c(1, 0, 1))
tsdiag(fit1)
Box.test(fit1$residuals, lag = 1)
# aic#
aic = matrix(NA, 6, 6)
for (p in 0:4) {
  for (q in 0:3) {
    a.p.q = arima(ret, order = c(p, 0, q))
    aic.p.q = a.p.q$aic
    aic[p + 1, q + 1] = aic.p.q
  }
}
```

```
bic
# select p and q order of ARIMA model
fit4 = arima(ret, order = c(2, 0, 3))
tsdiag(fit4)
Box.test(fit4$residuals, lag = 1)

fitr4 = arima(ret, order = c(2, 1, 3))
tsdiag(fitr4)
Box.test(fitr4$residuals, lag = 1)
# to conclude, 202 is better than 213
fit202 = arima(ret, order = c(2, 0, 2))
tsdiag(fit202)
tsdiag(fit4)
tsdiag(fitr4)

AIC(fit202, k = log(length(ret)))
AIC(fit4, k = log(length(ret)))
AIC(fitr4, k = log(length(ret)))
fit202$aic
fit4$aic
fitr4$aic
# arima202 predict
fit202 = arima(ret, order = c(2, 0, 2))
crpre = predict(fit202, n.ahead = 30)

dates = seq(as.Date("02/08/2014", format = "%d/%m/%Y"), by = "days", length = length(ret))

plot(ret, type = "1", xlim = c(0, 644), ylab = "log return", xlab = "days",
    lwd = 1.5)
lines(crpre$pred, col = "red", lwd = 3)
lines(crpre$pred + 2 * crpre$se, col = "red", lty = 3, lwd = 3)
lines(crpre$pred - 2 * crpre$se, col = "red", lty = 3, lwd = 3)

# Produces GARCH estimation results using ARIMA model residuals
rm(list = ls(all = TRUE))
graphics.off()
```

```
# install and load packages
libraries = c("FinTS", "tseries", "forecast", "fGarch")
lapply(libraries, function(x) if (!(x %in% installed.packages())) {
  install.packages(x)
})
lapply(libraries, library, quietly = TRUE, character.only = TRUE)

# load dataset
load(file = "C:/Users/xiumei/Desktop/big data/crix.RData")
ret = diff(log(crix1))

# vol cluster
fit202 = arima(ret, order = c(2, 0, 2))
par(mfrow = c(1, 1))
res = fit202$residuals
res2 = fit202$residuals^2

# different garch model
fg11 = garchFit(data = res, data ~ garch(1, 1))
summary(fg11)
fg12 = garchFit(data = res, data ~ garch(1, 2))
summary(fg12)
fg21 = garchFit(data = res, data ~ garch(2, 1))
summary(fg21)
fg22 = garchFit(data = res, data ~ garch(2, 2))
summary(fg22)
```

```
# residual plot
reszo = zoo(fg11@residuals, order.by = index(crix1))
plot(reszo, ylab = NA, lwd = 2)

par(mfrow = c(1, 2))
fg11res2 = fg11@residuals
acfres2  = acf(fg11res2, lag.max = 20, ylab = "Sample Autocorrelation",
               main = NA, lwd = 2)
pacfres2 = pacf(fg11res2, lag.max = 20, ylab = "Sample Partial Autocorrelation",
               main = NA, lwd = 2, ylim = c(-0.5, 0.5))

fg12res2 = fg12@residuals
acfres2  = acf(fg12res2, lag.max = 20, ylab = "Sample Autocorrelation",
               main = NA, lwd = 2)
pacfres2 = pacf(fg12res2, lag.max = 20, ylab = "Sample Partial Autocorrelation",
               main = NA, lwd = 2, ylim = c(-0.5, 0.5))

# qq plot
par(mfrow = c(1, 1))
plot(fg11, which = 13)   #9, 10, 11, 13

# kp test
set.seed(100)
x = rnorm(200)

# Do x and y come from the same distribution?
ks.test(x, fg11@residuals)
```

Homewor4

(Q1)

Q1. Improve the R quantlets on **GH** (from **CRIX** directory on quantlet.de) and make excellent graphics that follow **Fig 3,4,5,6** of the "Econometrics of **CRIX**" paper.
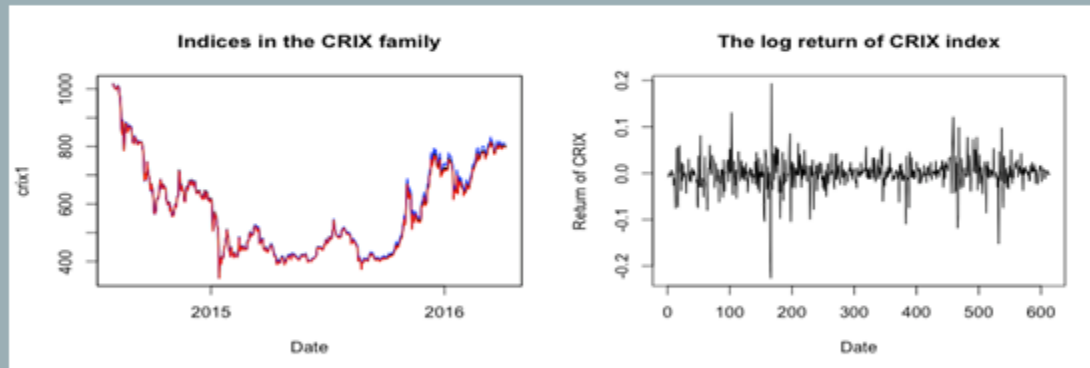
Figure 3: The daily value of indices in the CRIX family

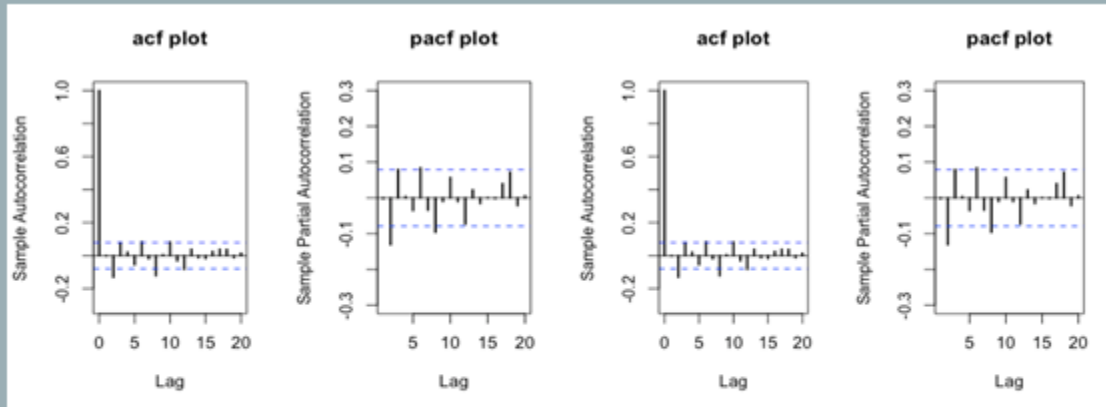Figure 4: The log returns of CRIX index

Homewor4

(Q1)

Figure 5: Histogram and QQ plot of CRIX returns   Figure 6: The sample ACF and PACF of CRIX returns

```
rm(list = ls(all = TRUE))
graphics.off()
# install and load packages
libraries = c("zoo", "tseries", "xts", "ccgarch")
lapply(libraries, function(x) if (!(x %in% installed.packages())) { install.packages(x)}
```

Homewor4
(Q1)

```
lapply(libraries, library, quietly = TRUE, character.only = TRUE)

# load dataset
load(file.choose())
load(file.choose())
load(file.choose())

# three indices return
ecrix1 = zoo(ecrix, order.by = index(crix1))
efcrix1 = zoo(efcrix, order.by = index(crix1))

# plot with different x-axis scales with zoo
my.panel <- function(x, ...) {
  lines(x, ...)
  lines(ecrix1, col = "blue")
  lines(efcrix1, col = "red")
}
plot.zoo(crix1, plot.type = "multiple", type = "l", lwd = 1.5, panel = my.panel,
      main = "Indices in the CRIX family", xlab = "Date")
```

Homewor4
(Q1)

```r
# plot of crix
# plot(as.xts(crix), type="l", auto.grid=FALSE, main = NA)
plot(crix1, ylab = "Price of CRIX", xlab = "Date")

# plot of crix return
ret  = diff(log(crix1))
# plot(as.xts(ret), type="l", auto.grid=FALSE, main = NA)
plot(ret, ylab = "Return of CRIX", xlab = "Date")

# stationary test
adf.test(ret, alternative = "stationary")
kpss.test(ret, null = "Trend")

par(mfrow = c(1, 2))
# histogram of returns
hist(ret, col = "grey", breaks = 20, freq = FALSE, ylim = c(0, 25), xlab = "Return of CRIX")
lines(density(ret), lwd = 2)
mu = mean(ret)
sigma = sd(ret)
x = seq(-4, 4, length = 100)
curve(dnorm(x, mean = mean(ret), sd = sd(ret)), add = TRUE, col = "red",
      lwd = 2)
```

Homewor4
(Q1)

```
# qq-plot
qqnorm(ret)
qqline(ret, col = "blue", lwd = 3)

# acf plot
autocorr = acf(ret, lag.max = 20, ylab = "Sample Autocorrelation", main = "acf plot",
        lwd = 2, ylim = c(-0.3, 1))

# pacf plot
autopcorr = pacf(ret, lag.max = 20, ylab = "Sample Partial Autocorrelation",
        main = "pacf plot", ylim = c(-0.3, 0.3), lwd = 2)
```

Homewor4
(Q2)

Q2. Make your R code perfect as in the R examples on quantlet.de i.e. make sure that the code is "time independent" by using actual dimensions of the data that you are collecting from crix.hu-berlin.de Recreate Fig 7 from "Econometrics of CRIX".
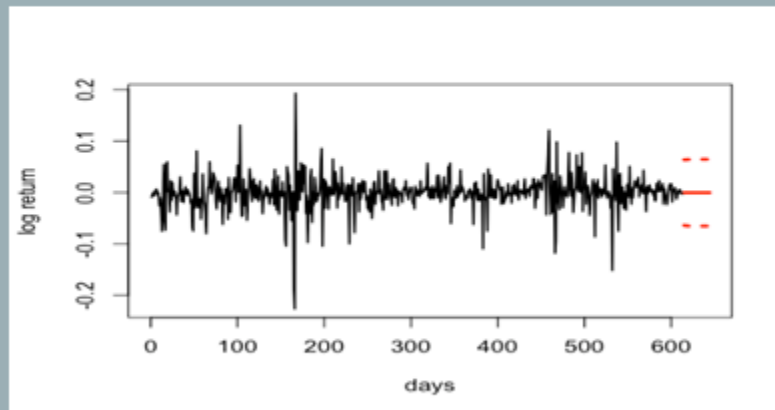
Figure 7: CRIX returns and predicted values.

Homewor4
(Q2)

```
Codes:

#arima model
par(mfrow = c(1, 1))
fit1 = arima(ret, order = c(1, 0, 1))
tsdiag(fit1)
Box.test(fit1$residuals, lag = 1)

#aic
aic = matrix(NA, 6, 6)
for (p in 0:4) {
  for (q in 0:3) {
    a.p.q = arima(ret, order = c(p, 0, q))
    aic.p.q = a.p.q$aic
    aic[p + 1, q + 1] = aic.p.q
  }
}

#bic
bic = matrix(NA, 6, 6)
for (p in 0:4) {
  for (q in 0:3) {
    b.p.q = arima(ret, order = c(p, 0, q))
    bic.p.q = AIC(b.p.q, k =
log(length(ret)))
    bic[p + 1, q + 1] = bic.p.q
  }
}

#select p and q order of ARIMA
model
fit4 = arima(ret, order = c(2, 0, 3))
tsdiag(fit4)
Box.test(fit4$residuals, lag = 1)

fitr4 = arima(ret, order = c(2, 1, 3))
tsdiag(fitr4)
Box.test(fitr4$residuals, lag = 1)
```

Homewor4
(Q2)

```r
# to conclude, 202 is better than 213
fit202 = arima(ret, order = c(2, 0, 2))

AIC(fit202, k = log(length(ret)))
AIC(fit4, k = log(length(ret)))
AIC(fitr4, k = log(length(ret)))
fit202$aic
fit4$aic
fitr4$aic

# arima202 predict
predict_num = 30
fit202 = arima(ret, order = c(2, 0, 2))
crpre = predict(fit202, n.ahead = predict_num)

dates = seq(as.Date("02/08/2014", format = "%d/%m/%Y"), by = "days", length = length(ret))
plot(ret, type = "l", xlim = c(0, length(ret)+predict_num), ylab = "log return", xlab = "days",
    lwd = 1.5, col = "black")
lines(crpre$pred, col = "red", lwd = 3)
lines(crpre$pred + 2 * crpre$se, col = "red", lty = 3, lwd = 3)
lines(crpre$pred - 2 * crpre$se, col = "red", lty = 3, lwd = 3)
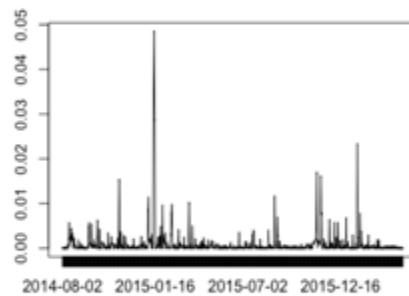```

# Homewor4
(Q3)



**Q3. Redo as many figures as you can.**

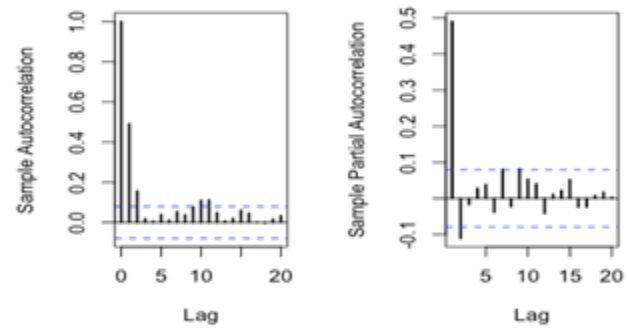Figure 8: The squared ARIMA(2,0,2) residuals of CRIX returns.

Figure 9: The ACF and PACF of squared ARIMA(2,0,2) residuals

# Homewor4
(Q3)

```
Codes:
rm(list = ls(all = TRUE))
graphics.off()

# install and load packages
libraries = c("tseries")
lapply(libraries, function(x) if (!(x %in% installed.packages())) {
  install.packages(x)
})
lapply(libraries, library, quietly = TRUE, character.only = TRUE)

# please change your working directory
setwd()
load(file.choose())
Pr = as.numeric(crix)
Da = factor(date1)
crx = data.frame(Da, Pr)
# plot of crix return
ret = diff(log(crx$Pr))
Dare = factor(date1[-1])
retts = data.frame(Dare, ret)
# arima202 predict
fit202 = arima(ret, order = c(2, 0, 2))

# vola cluster
par(mfrow = c(1, 1))
res = fit202$residuals
res2 = fit202$residuals^2
tsres202 = data.frame(Dare, res2)
plot(tsres202$Dare,
tsres202$res2, type = "o", ylab =
NA)
lines(tsres202$res2)

plot(res2, ylab='Squared
residuals', main=NA)
par(mfrow = c(1, 2))
acfres2 = acf(res2, main = NA,
lag.max = 20, ylab = "Sample
Autocorrelation", lwd = 2)
pacfres2 = pacf(res2, lag.max = 20,
ylab = "Sample Partial
Autocorrelation", lwd = 2, main =
NA)
```
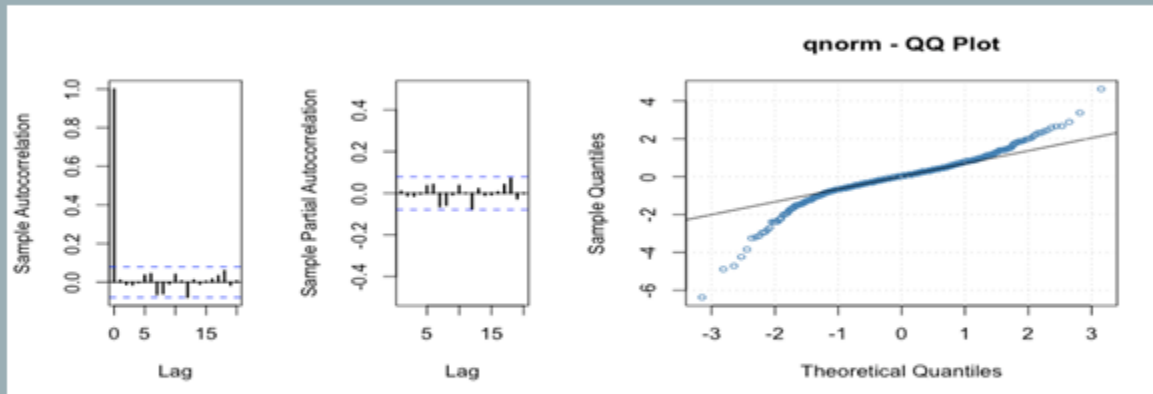
# Homewor4

(Q3)



Figure 10: The ACF and PACF of squared ARIMA(2,0,2) residuals

Figure 11: The QQ plots of model residuals of ARIMA-GARCH process.

# Homewor4
(Q3)

```
Codes:
rm(list = ls(all = TRUE))
graphics.off()

# install and load packages
libraries = c("tseries")
lapply(libraries, function(x) if (!(x %in% installed.packages())) {
  install.packages(x)
})
lapply(libraries, library, quietly = TRUE, character.only = TRUE)

# please change your working directory
setwd()
load(file.choose())
Pr = as.numeric(crix)
Da = factor(date1)
crx = data.frame(Da, Pr)
# plot of crix return
ret = diff(log(crx$Pr))
Dare = factor(date1[-1])
retts = data.frame(Dare, ret)
# arima202 predict
fit202 = arima(ret, order = c(2, 0, 2))
```

```
# vola cluster
par(mfrow = c(1, 1))
res = fit202$residuals
res2 = fit202$residuals^2
tsres202 = data.frame(Dare, res2)
plot(tsres202$Dare,
tsres202$res2, type = "o", ylab =
NA)
lines(tsres202$res2)

# plot(res2, ylab='Squared
residuals', main=NA)
par(mfrow = c(1, 2))
acfres2 = acf(res2, main = NA,
lag.max = 20, ylab = "Sample
Autocorrelation", lwd = 2)
pacfres2 = pacf(res2, lag.max = 20,
ylab = "Sample Partial
Autocorrelation", lwd = 2, main =
NA)
```
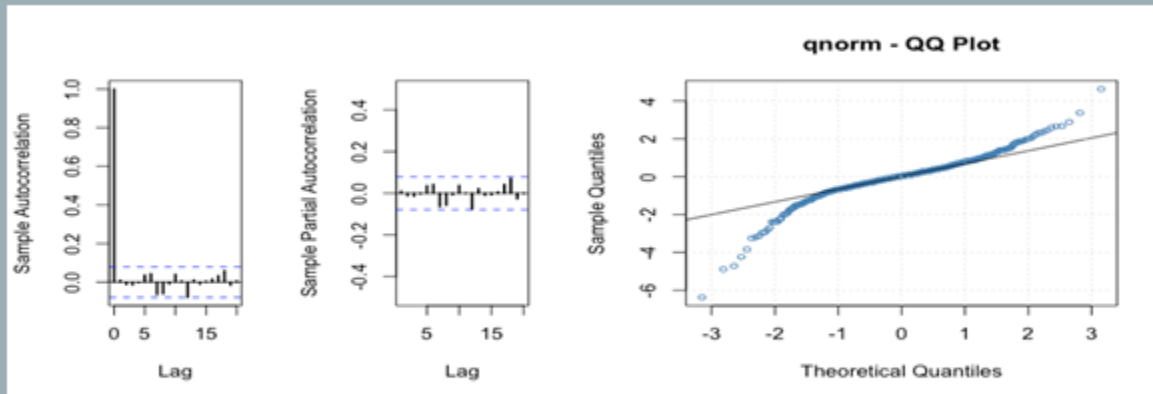
Homewor4

(Q3)

Figure 10: The ACF and PACF of squared ARIMA(2,0,2) residuals

Figure 11: The QQ plots of model residuals of ARIMA-GARCH process.