

Building GARCH Model to Estimate Volatilities

Overview

This is an example of using GARCH model to estimate volatilities of S&P500 index. I divide the analysis into following steps: 1. Data Preparation and Cleaning 2. Data Visualization 3. GARCH Modeling 4. Forecast and Estimate

Data Preparation and Cleaning

```
library(tidyverse)
library(tseries)
df = read.csv('sp500.csv', header = TRUE)

head(df)

##      date  price
## 1 7/18/2005 1221.13
## 2 7/19/2005 1229.35
## 3 7/20/2005 1235.20
## 4 7/21/2005 1227.04
## 5 7/22/2005 1233.68
## 6 7/25/2005 1229.03

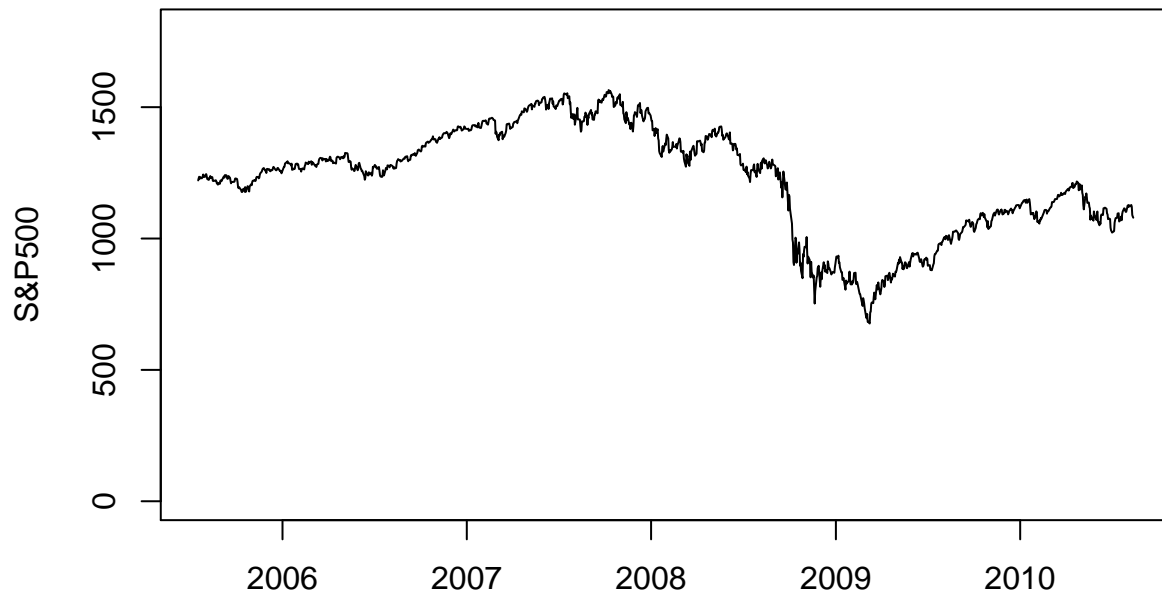
df$date = as.Date(df$date, format = "%m/%d/%Y")

n = nrow(df)
ret = df$price[-1]/df$price[-n] - 1
col3 = c(NA, ret)
df2 = data.frame(df, ret = col3)
```

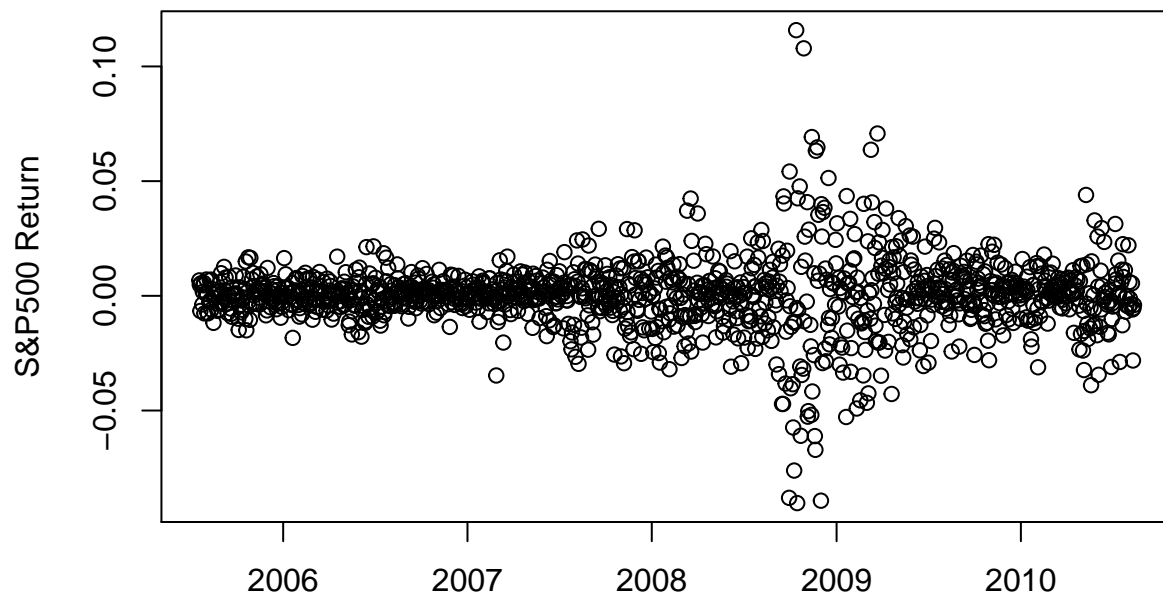
Data Visualization

```
plot(price~date, df, type = 'l',ylim = c(0,1800), ylab = "S&P500", xlab = "", main = "S&P500 Index Trend")
```

S&P500 Index Trend

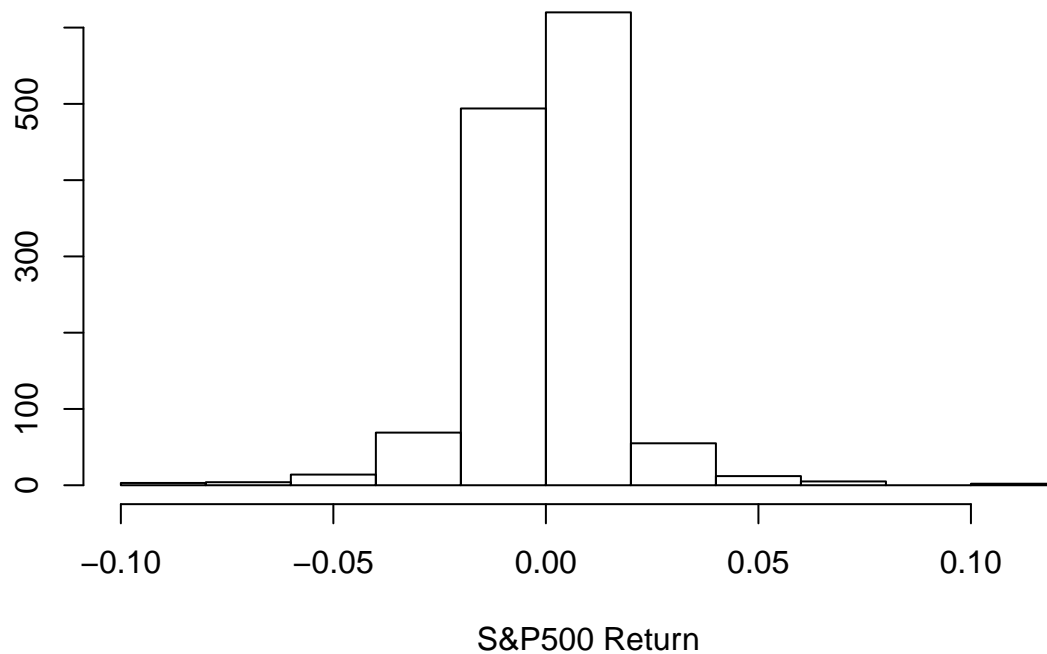


```
plot(ret~date, df2, type = 'p', ylab = "S&P500 Return", xlab = "")
```



```
hist(df2$ret, xlab = "S&P500 Return",main = "Distribution of S&P500 Return",ylab = "")
```

Distribution of S&P500 Return



GARCH Modeling

```
# build garch model
```

```
ml = garch(ret)
```

```
# Extract parameters
```

```
w = coef(ml)[1] #gamma
```

```
a = coef(ml)[2] #alpha
```

```
b = coef(ml)[3] #beta
```

```
varlong = w/(1-a-b) # long term variance
```

```
summary(ml)
```

```
##
```

```
## Call:
```

```
## garch(x = ret)
```

```
##
```

```
## Model:
```

```
## GARCH(1,1)
```

```
##
```

```
## Residuals:
```

```
##      Min      1Q  Median      3Q      Max
```

```
## -6.4131 -0.5129  0.0761  0.5601  2.9629
```

```
##
```

```
## Coefficient(s):
```

```
##      Estimate Std. Error  t value Pr(>|t|)
```

```
## a0 1.552e-06  3.201e-07   4.849 1.24e-06 ***
```

```
## a1 9.287e-02  1.262e-02   7.361 1.82e-13 ***
```

```
## b1 8.998e-01    1.282e-02    70.206    < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Diagnostic Tests:
##  Jarque Bera Test
##
## data:  Residuals
## X-squared = 229.56, df = 2, p-value < 2.2e-16
##
##
##  Box-Ljung test
##
## data:  Squared.Residuals
## X-squared = 5.3351, df = 1, p-value = 0.0209
# display the attributes of ml
str(ml)

## List of 10
##  $ order      : Named num [1:2] 1 1
##    ..- attr(*, "names")= chr [1:2] "p" "q"
##  $ coef       : Named num [1:3] 1.55e-06 9.29e-02 9.00e-01
##    ..- attr(*, "names")= chr [1:3] "a0" "a1" "b1"
##  $ n.likeli    : num -5108
##  $ n.used      : int 1278
##  $ residuals   : num [1:1278] NA 0.34 -0.492 0.418 -0.303 ...
##  $ fitted.values: num [1:1278, 1:2] NA 0.014 0.0134 0.0129 0.0125 ...
##    ..- attr(*, "dimnames")=List of 2
##      .. ..$ : NULL
##      .. ..$ : chr [1:2] "sigt" "-sigt"
##  $ series      : chr "ret"
##  $ frequency    : num 1
##  $ call        : language garch(x = ret)
##  $ vcov        : num [1:3, 1:3] 1.02e-13 2.72e-09 -3.52e-09 2.72e-09 1.59e-04 ...
##    ..- attr(*, "dimnames")=List of 2
##      .. ..$ : chr [1:3] "a0" "a1" "b1"
##      .. ..$ : chr [1:3] "a0" "a1" "b1"
##  - attr(*, "class")= chr "garch"

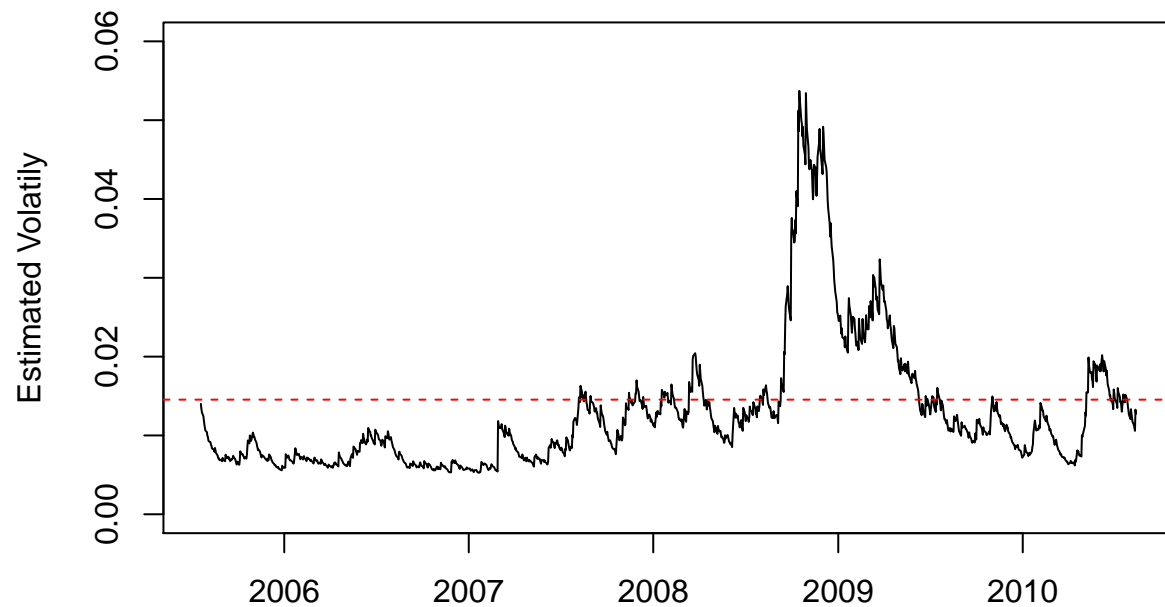
# fitted daily volatilities
fit = ml$fitted.values[-1,1]
head(fit)

## [1] 0.01399930 0.01341628 0.01294467 0.01245169 0.01193228 0.01139924

dates = df$date[-c(1,2)]

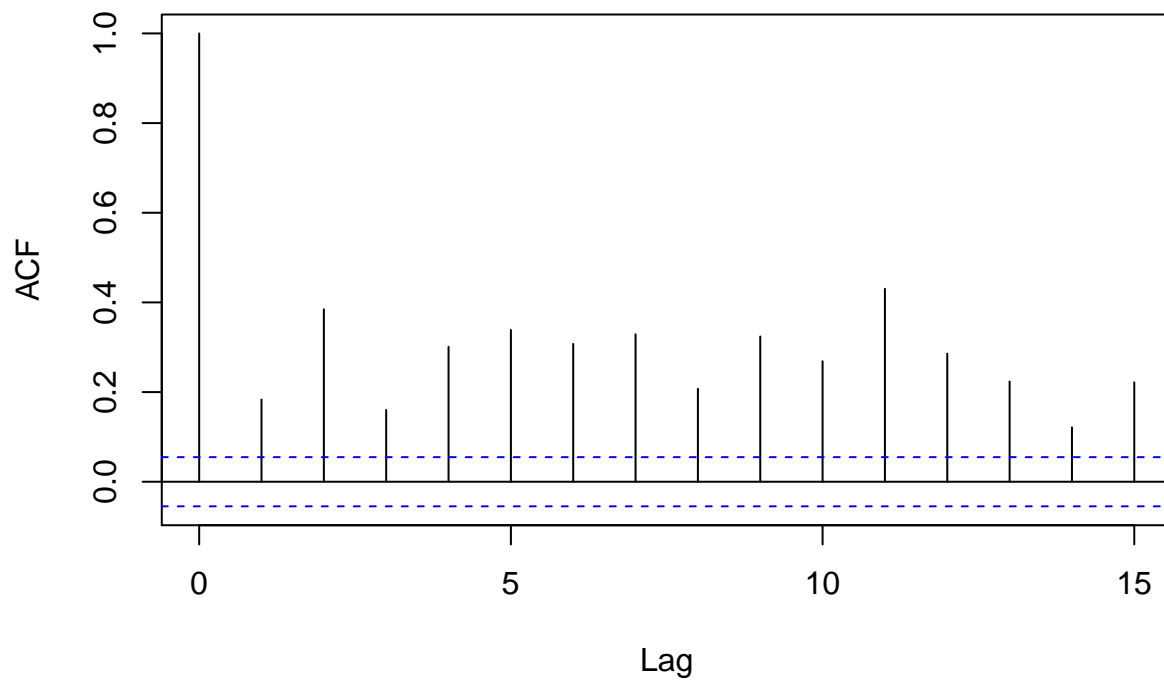
plot(fit~dates, type = 'l', ylim = c(0,0.06), xlab="", ylab = "Estimated Volatility", main = "Estimated S&P
abline(h = sqrt(varlong), lty = 2, col = 'red')
```

Estimated S&P500 Volatility v.s. Long-term Volatility



```
# autocorrelation for squared returns  
ret2 = ret^2  
m2 = acf(ret2, lag.max = 15)
```

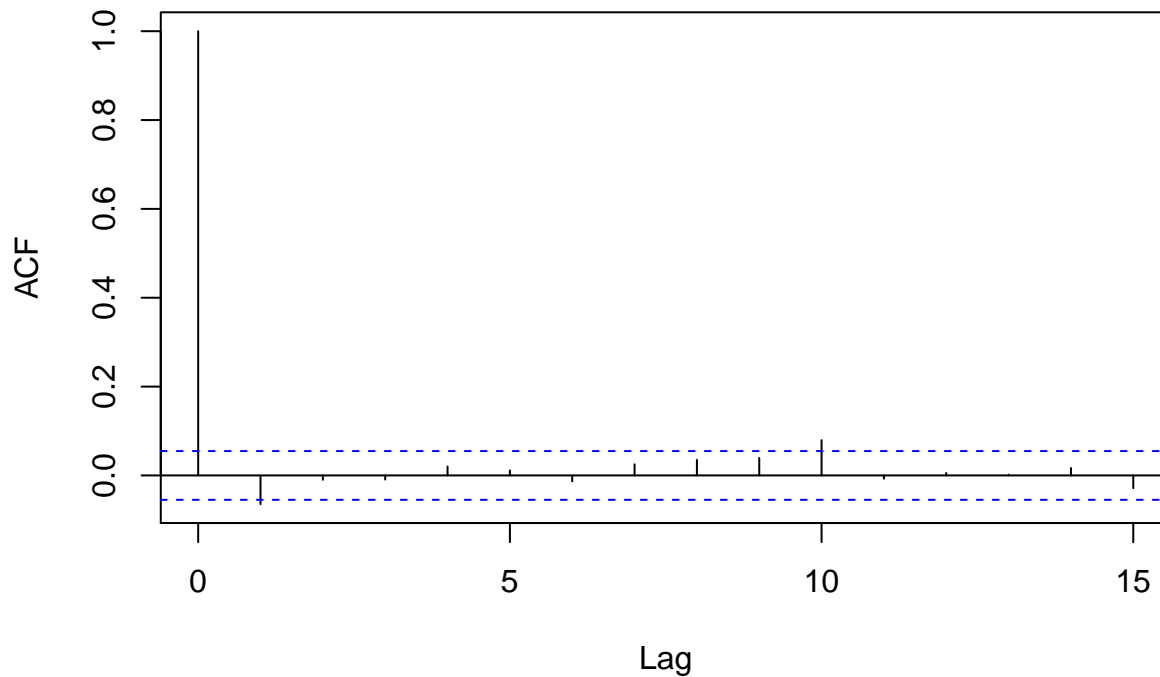
Series ret2



```
# if lag is within blue band, it is white noise  
col1 = m2$acf
```

```
# autocorrelation for the ratio of squared return over fitted daily variance
ratio = ret2[-1]/fit^2
m3 = acf(ratio,15)
```

Series ratio



```
col2 = m3$acf[-1]
```

```
# table
col1 = m2$acf[-1]
col2 = m3$acf[-1]
df3 = data.frame(ret2 = col1, ratio = col2)
print(df3)
```

##	ret2	ratio
## 1	0.1833666	-0.064560630
## 2	0.3847384	-0.009807882
## 3	0.1601414	-0.009358168
## 4	0.3010945	0.020028486
## 5	0.3386284	0.011479485
## 6	0.3074964	-0.013356350
## 7	0.3289945	0.025121788
## 8	0.2073292	0.035024266
## 9	0.3239728	0.039379281
## 10	0.2688047	0.079414876
## 11	0.4303820	-0.007233287
## 12	0.2857194	0.005529120
## 13	0.2234592	0.001819274
## 14	0.1213700	0.016635141
## 15	0.2218163	-0.028807675

```

#ljung box test for squared return
Box.test(ret2, lag = 15, type = "Ljung-Box")

##
## Box-Ljung test
##
## data:  ret2
## X-squared = 1566.3, df = 15, p-value < 2.2e-16
# ljung box test for ratio
Box.test(ratio, lag = 15, type = 'Ljung-Box')

##
## Box-Ljung test
##
## data:  ratio
## X-squared = 20.545, df = 15, p-value = 0.152

```