Building GARCH Model to Estimate Volatilites

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 Prearation and Cleaning 2. Data Visualization 3. Variable Testing 4. GARCH Modeling

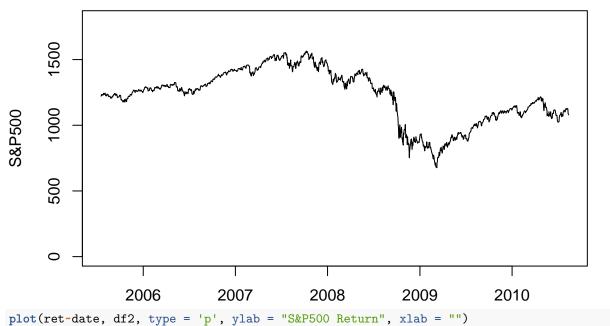
Data Preparation and Cleaning

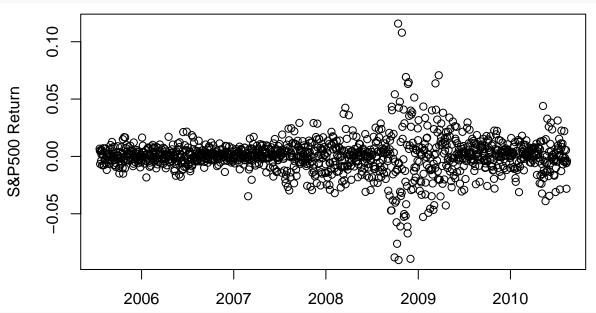
```
library(tidyverse)
library(tseries)
library(TSA)
df = read.csv('sp500.csv', header = TRUE)
head(df)
##
          date
               price
## 1 7/18/2005 1221.13
## 2 7/19/2005 1229.35
## 3 7/20/2005 1235.20
## 4 7/21/2005 1227.04
## 5 7/22/2005 1233.68
## 6 7/25/2005 1229.03
df$date = as.Date(df$date, 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 Tren
```

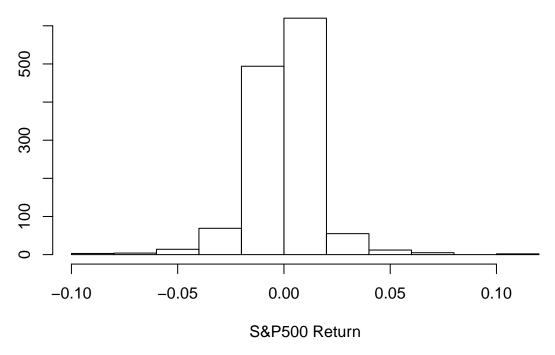
S&P500 Index Trend





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

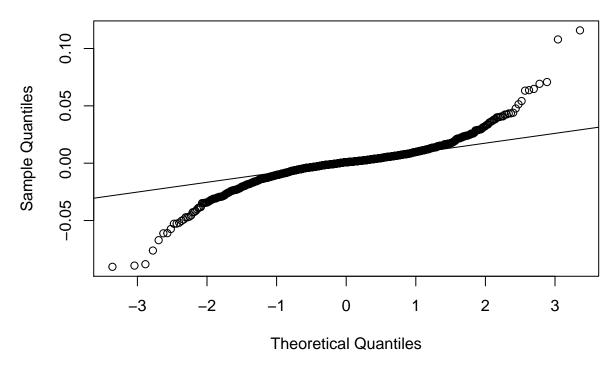
Distribution of S&P500 Return



Variable Testing

qqnorm(df2\$ret)
qqline(df2\$ret)

Normal Q-Q Plot



Shapiro-Wilk Test of Normality The result shows that return of S&P500 does not follow normal distribution.

shapiro.test(df2\$ret) ## ## Shapiro-Wilk normality test ## ## data: df2\$ret ## W = 0.87371, p-value < 2.2e-16 Use acf and pacf to identify time-correlation of daily returns/squared returns. $ret2 = (df2\$ret[-1])^2$ par(mfrow= c(2,2)) acf(ret, lag.max = 15)acf(ret2, lag.max = 15)pacf(ret, lag.max = 15) pacf(ret2, lag.max = 15) Series ret Series ret2 0.05 0.3 ACF -0.10 0.0 2 4 6 8 10 12 14 2 4 6 8 10 12 14 Lag Lag Series ret2 Series ret 0.05 Partial ACF Partial ACF 0.2 -0.10 -0.1 2 6 8 2 6 4 10 12 14 4 8 10 12 14

Use t-test on whether mean return is equal to zero. Since p-value is larger than 0.05, we fail to reject null hypothesis that mean return is equal to zero.

Lag

```
##
## One Sample t-test
##
## data: ret
## t = 0.055256, df = 1277, p-value = 0.9559
## alternative hypothesis: true mean is not equal to 0
## 95 percent confidence interval:
## -0.0008283061 0.0008763181
## sample estimates:
```

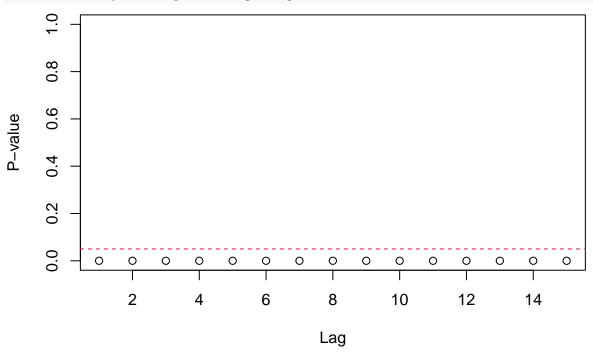
Lag

t.test(ret,mu=0)

```
## mean of x
## 2.400598e-05
```

Use McLeod-Li Test on whether daily volatility is constant. p-value for different lags are all less than 0.05, therefore we have confidence to reject null hypothesis that daily volatility is constant.

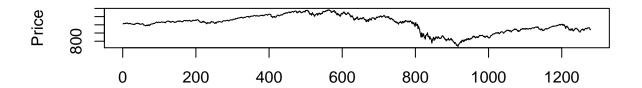
```
McLeod.Li.test(y = ret, plot = T, gof.lag=15)
```

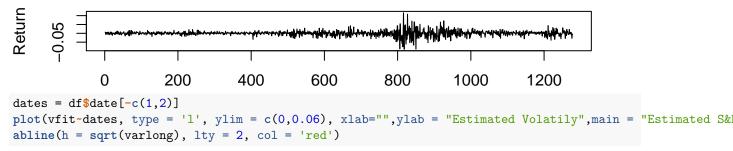


GARCH Modeling

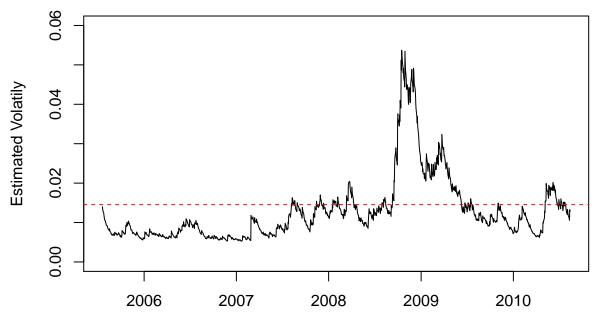
```
# build garch model
ml = garch(ret)
# Extract parameters
w = coef(ml)[1]
                   #gamma
a = coef(m1)[2]
                   #alpha
b = coef(m1)[3]
                  \#beta
varlong = w/(1-a-b) # long term variance
summary(ml)
##
## Call:
## garch(x = ret)
##
## Model:
## GARCH(1,1)
##
## Residuals:
##
       Min
                1Q
                    Median
                                 ЗQ
                                        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 '**' 0.05 '.' 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 atrributes 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"
                : chr "ret"
## $ series
## $ 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
vfit = ml$fitted.values[-1,1]
head(vfit)
## [1] 0.01399930 0.01341628 0.01294467 0.01245169 0.01193228 0.01139924
par(mfrow = c(2,1))
plot(df2$price, type="l",ylab = "Price", xlab = "")
plot(ret, type = "l", ylab = "Return", xlab = "")
```



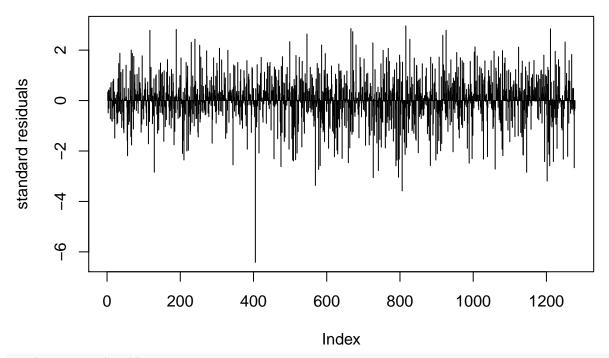


Estimated S&P500 Volatility v.s. Long-term Volatility



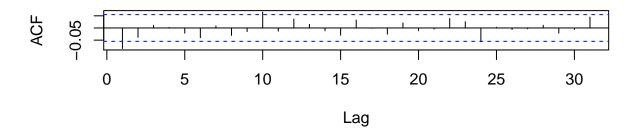
Test of Residual

plot(residuals(ml),type = "h", ylab = "standard residuals")

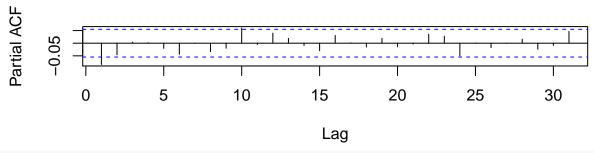


par(mfrow = c(2,1))
acf(residuals(ml)[-1])
pacf(residuals(ml)[-1])

Series residuals(ml)[-1]

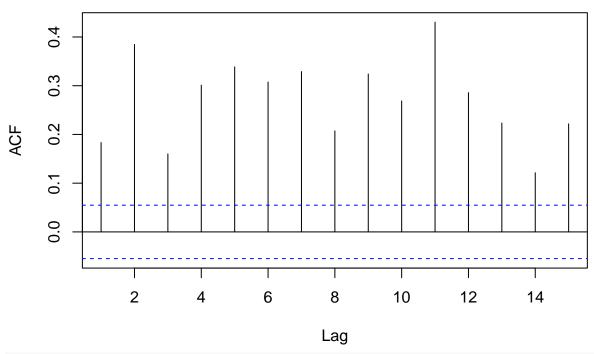


Series residuals(ml)[-1]



```
# autrocorrelation 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]/vfit^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
## 1 0.3847384 -0.009807882
## 2 0.1601414 -0.009358168
## 3 0.3010945 0.020028486
## 4 0.3386284 0.011479485
## 5 0.3074964 -0.013356350
## 6 0.3289945 0.025121788
## 7 0.2073292 0.035024266
## 8 0.3239728 0.039379281
## 9 0.2688047 0.079414876
## 10 0.4303820 -0.007233287
## 11 0.2857194 0.005529120
## 12 0.2234592 0.001819274
## 13 0.1213700 0.016635141
## 14 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</pre>
```