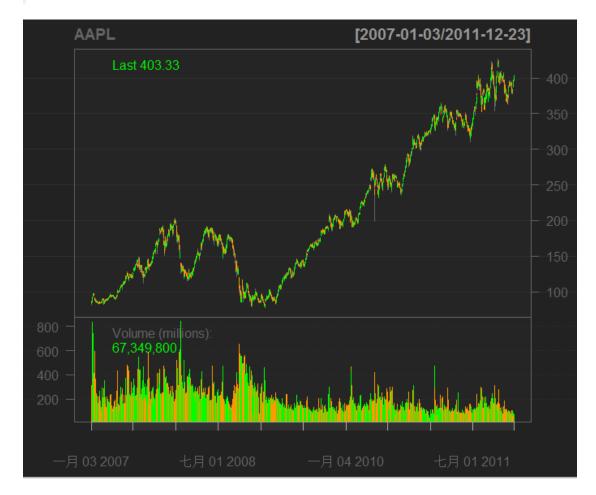
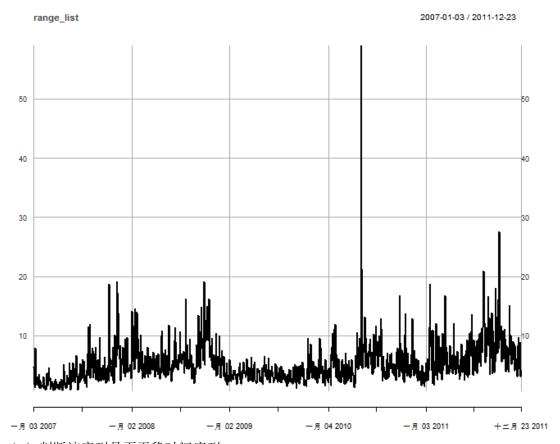
- 1. 考虑在 2007 年 1 月 2 日到 2011 年 12 月 23 日期间,苹果公司股票每天的股价波动幅度(即当天的最高价减去最低价)。这个数据可以利用 quantmod 包从雅虎财经获得(股票代码 AAPL)。
- (1)判断该序列是否平稳时间序列
- (2) 寻找适当的 ARIMA 模型拟合该序列,并做超前 1~12 步预测
- (3)对拟合的模型做残差检验

#获取整理数据

```
library(quantmod)
getSymbols("AAPL",src="yahoo",from="2007-01-02",to="2011-12-23")
chartSeries(AAPL,subset="2007-01-02::2011-12-23")
range_list=Hi(AAPL)-Lo(AAPL)
head(range_list)
tail(range_list)
plot(range_list)|
```



```
> head(range_list)
          AAPL. High
2007-01-03
               4.68
2007-01-04
                2.13
2007-01-05
                1.80
2007-01-08
                1.25
2007-01-09
                7.83
2007-01-10
                4.35
> tail(range_list)
          AAPL. High
2011-12-16
               4.58
2011-12-19
                4.37
2011-12-20
                8.84
2011-12-21
                5.29
2011-12-22
                3.03
2011-12-23
                4.10
```



#(1)判断该序列是否平稳时间序列

- > library(urca)
- > summary(ur.df(range_list,lags=6,type='drift'))

```
# Augmented Dickey-Fuller Test Unit Root Test #
 *************************************
Test regression drift
call:
 lm(formula = z.diff \sim z.lag.1 + 1 + z.diff.lag)
 Residuals:
             1Q Median
                            3Q
    Min
                                   Max
        -1.502 -0.549
                        1.007 51.637
 -10.064
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
                                6.323 3.56e-10 ***
 (Intercept) 1.20857
                       0.19112
                       0.03279 -6.972 5.05e-12 ***
 z.lag.1
            -0.22863
                       0.03863 -11.286 < 2e-16 ***
 z.diff.lag1 -0.43597
 z.diff.lag2 -0.34883
                       0.03953 -8.824 < 2e-16 ***
                       0.03890 -5.890 4.95e-09 ***
 z.diff.lag3 -0.22917
 z.diff.lag4 -0.19745
                       0.03724 -5.302 1.36e-07 ***
 z.diff.lag5 -0.09407
                       0.03391 -2.774 0.00562 **
z.diff.lag6 -0.05124
                       0.02832 -1.810 0.07060 .
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 2.882 on 1241 degrees of freedom
Multiple R-squared: 0.3192, Adjusted R-squared: 0.3154
F-statistic: 83.14 on 7 and 1241 DF, p-value: < 2.2e-16
Value of test-statistic is: -6.9724 24.3077
Critical values for test statistics:
1pct 5pct 10pct
tau<mark>2 -3.43 -2.86 -2.57</mark>
 phi1 6.43 4.59 3.78
10 library(tserise)
11 adf.test(range_list)
 > adf.test(range_list)
  Augmented Dickey-Fuller Test
 data:
         range_list
 Dickey-Fuller = -5.4855, Lag order = 10, p-value = 0.01
 alternative hypothesis: stationary
 Warning message:
 In adf.test(range_list) : p-value smaller than printed p-value
#-6.9724 小于-3.43、-2.86、-2.57,且 p 值足够小,所以可以拒绝 H0 假设,是平稳的时间序
#(2) 寻找适当的 ARIMA 模型拟合该序列,并做超前 1~12 步预测
12 library(forecast)
13 range_list.arma<-auto.arima(range_list,stationary=TRUE,seasonal=FALSE,ic="aic")</pre>
14 summary(range_list.arma)
15 confint(range_list.arma)
```

```
> summary(range_list.arma)
 Series: range_list
 ARIMA(2,0,3) with non-zero mean
 Coefficients:
         ar1
                 ar2
                        ma1
                                ma2
                                        ma3 intercept
      0.0233 0.9108 0.2956
                            -0.7672
                                     -0.1208
                                                5.2201
 s.e. 0.0418
             0.0414 0.0509
                             0.0389
                                     0.0335
                                                0.4890
 sigma^2 estimated as 8.136: log likelihood=-3099.04
 AIC=6212.09
             AICC=6212.18 BIC=6248.04
 Training set error measures:
                           RMSE
                    ME
                                    MAE
                                             MPF
 Training set 0.01000129 2.852327 1.761872 -20.55835 40.00176
                 MASE
                            ACF1
 Training set 0.8098619 0.003593337
 > confint(range_list.arma)
                          97.5 %
               2.5 %
          -0.05850686 0.10517226
 ar1
           0.82966632 0.99190966
 ar2
           0.19588139 0.39524392
 ma1
          -0.84338509 -0.69091776
 ma2
          -0.18647357 -0.05512852
 ma3
 intercept 4.26176420 6.17849484
 > pred.last12<-t(predict(range_list.arma,n.ahead=12)$pred[1:12])</pre>
 > pred.last12
           [,1]
                      [,2]
                                 [,3]
                                            [,4]
                                                      [,5]
                                                                 [,6]
                                                                            [,7]
  [1,] 5.10333 5.247323 5.229492 5.245115 5.22924 5.243099 5.228963
                       [,9] [,10] [,11]
  [1,] 5.241256 5.228668 5.23957 5.22836 5.238028
  > |
#(3)对拟合的模型做残差检验
 > accuracy(range_list.arma)
                           ME
                                    RMSE
                                                MAE
                                                             MPE
 Training set 0.01000129 2.852327 1.761872 -20.55835 40.00176
                       MASE
                                      ACF1
 Training set 0.8098619 0.003593337
 > tsdiag(range_list.arma)
                                    Standardized Residuals
   5
                  200
                              400
                                          600
                                                     800
                                                                1000
                                                                            1200
                                      ACF of Residuals
   8.
ACF
0.4
                               10
                                           15
                                                       20
                                                                               30
                                           Lag
                                  p values for Ljung-Box statistic
  8.0
  4.0
                2
                                                6
                                                                8
                                                                                10
```

lag

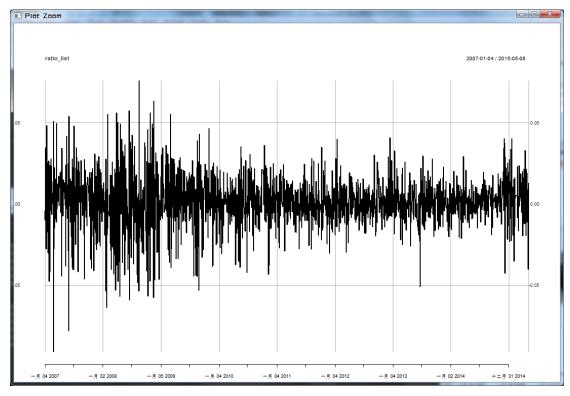
2. 自己随意找一组时间序列,按照时间序列分析的步骤对该序列做分析并检验模型

```
> library(quantmod)
> getSymbols("ASSEC")
    As of 0.4-0, 'getSymbols' uses env=parent.frame() and
    auto.assign=TRUE by default.
This behavior will be phased out in 0.5-0 when the call will
    default to use auto.assign=FALSE. getOption("getSymbols.env") and
    getOptions("getSymbols.auto.assign") are now checked for alternate defaults

This message is shown once per session and may be disabled by setting
    options("getSymbols.warning4.0"=FALSE). See ?getSymbols for more details.
[1] "SSEC"
Warning message:
In download.file(paste(yahoo.URL, "s=", Symbols.name, "&a=", from.m, :
    downloaded length 145479 != reported length 200
> ratio_list = (Cl(SSEC) - op(SSEC))/Hi(SSEC)
```



```
> #指熱变化室
> plot(ratio_list)
```



```
> ratio_list.numeric = as.numeric(ratio_list$SSEC.Close)
> ar(ratio_list.numeric,method='mle')$order #base on AIC准则
[1] 4
```

```
> library(fUnitRoots)
> adfTest(ratio_list.numeric,lags=4,type="c")

Title:
Augmented Dickey-Fuller Test

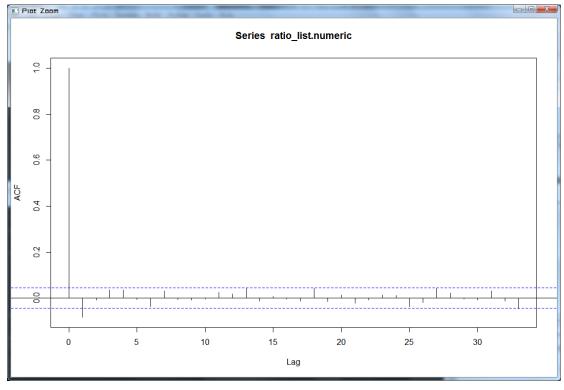
Test Results:
    PARAMETER:
    Lag Order: 4
    STATISTIC:
    Dickey-Fuller: -19.4049
    P VALUE:
    0.01

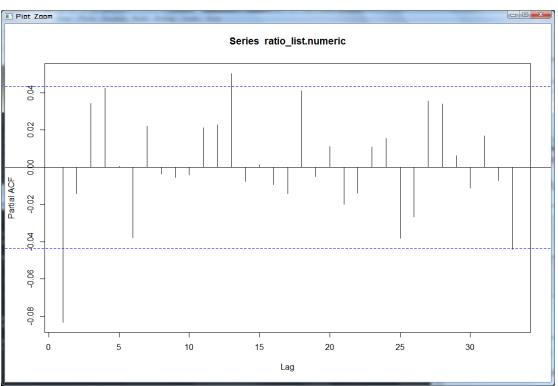
Description:
    Mon May 11 20:35:35 2015 by user: Administrator

Warning message:
In adfTest(ratio_list.numeric, lags = 4, type = "c") :
    p-value smaller than printed p-value
```

p 值=0.01 很小,拒绝 H0 假设,证明是一个平稳序列

```
> acf(ratio_list.numeric)
> pacf(ratio_list.numeric)
```



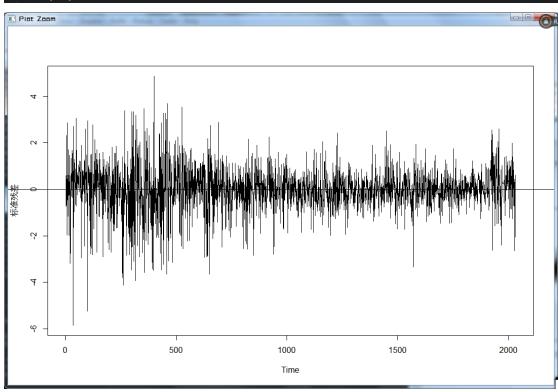


因此不需要差分,原时间序列已经是一个平稳时间序列

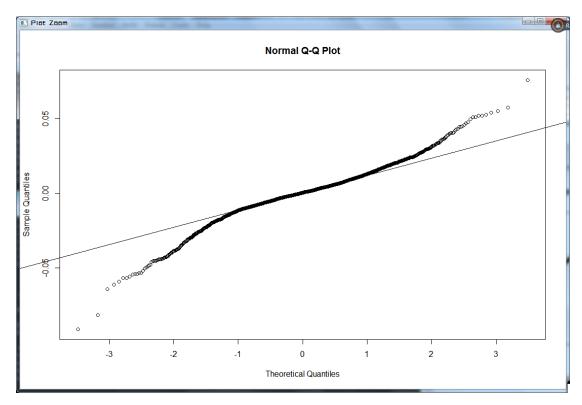
根据结果我们最终选择 p=0,q=1,即确定 ARIMA(0,0,1)

检查残差是否为白噪声序列

```
> plot(rstandard(m2),ylab='标准残差',type='l')
> abline(h=0)
```



```
> qqnorm(residuals(m2))
> qqline(residuals(m2))
```



```
> predict(m2,12)$pred

Time Series:

start = 2033

End = 2044

Frequency = 1

[1] 0.0002930005 0.0010943606 0.0010943606 0.0010943606 0.0010943606 0.0010943606

[8] 0.0010943606 0.0010943606 0.0010943606 0.0010943606
```

通过预测当然不能得到准确的未来大盘指数变化率。但是可以看到市场已经进入一个不稳定状态并且收益率待定状态。