homework 6

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Problem 1

```
(a)
```

```
library(lubridate)
##
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
       date, intersect, setdiff, union
##
Rday<-read.csv("DailyLogRet-intc9608.txt",sep = "",skip = 2)</pre>
day <- ymd(Rday$date)</pre>
a = 1
Rmon \leftarrow rep(0, 156)
count <- rep(0, 156)
for(i in 1996:2008){
for(j in 1:12){
   data = 0
   n = 0
    for(k in 1:3274){
      if(year(day[k])==i & month(day[k])==j){
        data = data+Rday$rtn[k]
        n = n+1
      Rmon[a] = data
      count[a] = n
    }
    a = a+1
 }
}
Rmon
##
     [1] -0.013699 0.065503 -0.027011 0.181176 0.111511 -0.025071 0.029423
##
     [8] 0.063809 0.182767 0.147855 0.148947 0.038207 0.219642 -0.129293
    [15] -0.015724  0.102468 -0.002666 -0.062963  0.266319  0.008622
##
                                                                         0.006631
##
     \begin{bmatrix} 22 \end{bmatrix} -0.163335 \quad 0.013446 \ -0.094221 \quad 0.147141 \quad 0.106189 \ -0.126723 \quad 0.040166 
##
   [29] -0.118845 0.044594 0.135982 -0.161752 0.196092 0.044883 0.194083
##
   [36] 0.107574 0.183240 -0.149892 -0.000643 0.039656 -0.113916
                                                                         0.108967
##
   [43] 0.155890 0.182175 -0.093192 0.054005 -0.002534 0.077845
  [50] 0.140807 0.164638 -0.015535 0.004544 0.078451 0.005738 0.121390
##
## [57] -0.543175 0.107835 -0.140987 -0.204036 0.223211 -0.244502 -0.051490
```

```
[64] 0.211793 -0.125551 0.088299 0.027776 -0.053142 -0.293604 0.193664
   [71] 0.299709 -0.032086 0.117700 -0.196893 0.071999 -0.054451 -0.019767
##
  [78] -0.377424 0.056101 -0.097383 -0.168568 0.259466 0.200794 -0.285841
## [85] 0.018604 0.099994 -0.048002 0.128772 0.132513 0.003538 0.186436
   [92] 0.142823 -0.033524 0.183653 0.021333 -0.041813 -0.046772 -0.039105
## [99] -0.065970 -0.052653 0.107360 -0.031368 -0.114865 -0.129804 -0.054459
## [106] 0.107763 0.009768 0.047804 -0.039428 0.072043 -0.030400 0.013990
## [113] 0.140886 -0.033736 0.044572 -0.049504 -0.040759 -0.046726 0.133255
## [120] -0.064905 -0.151963 -0.024828 -0.055966 0.027824 -0.095500
                                                                   0.056483
## [134] -0.047194 -0.036050 0.118772 0.037781 0.070853 -0.000574 0.094916
## [141] 0.005423 0.042601 -0.021447 0.025062 -0.213285 -0.044072 0.064082
## [148] 0.055255 0.050201 -0.071684 0.037863 0.039335 -0.183812 -0.123568
## [155] -0.118912 0.078969
 (b)
#find the ACF value at lag 1
acf(Rday$rtn,lag.max =1, plot = F)
##
## Autocorrelations of series 'Rday$rtn', by lag
##
##
       0
##
  1.000 -0.044
r1 = -0.044
vart < - rep(0, 156)
a = 1
for(i in 1: 156){
 sum = 0
 for( j in 1: count[i]){
   sum = sum+(Rday$rtn[a]-Rmon[i]/count[i])^2
   a = a + 1
 }
 vart[i] <- (1+ 2*r1)/(count[i]-1)*sum
}
vart
    [1] 1.065245e-03 2.532310e-04 5.842005e-04 4.556822e-04 2.602146e-04
##
    [6] 2.503194e-04 5.201839e-04 2.794737e-04 3.111962e-04 5.089541e-04
##
## [11] 4.422684e-04 6.148096e-04 3.392970e-04 4.445702e-04 3.732631e-04
##
   [16] 5.216405e-04 7.186794e-04 2.737350e-04 5.447446e-04 4.764906e-04
##
  [21] 4.199913e-04 1.456627e-03 5.471747e-04 4.672696e-04 3.742878e-04
   [26] 4.112526e-04 9.545261e-04 4.758318e-04 3.888684e-04 6.634493e-04
   [31] 4.670564e-04 7.091810e-04 8.518386e-04 4.883052e-04 4.743127e-04
##
   [36] 9.297189e-04 9.809180e-04 1.050647e-03 6.829172e-04 9.639130e-04
##
  [41] 8.567854e-04 1.123824e-03 6.736237e-04 5.433252e-04 6.605664e-04
##
  [46] 1.171371e-03 6.251340e-04 6.203689e-04 1.994614e-03 6.872010e-04
##
    [51] 7.593204e-04 2.452774e-03 1.834458e-03 7.667027e-04 6.799972e-04
   [56] 4.993063e-04 3.222718e-03 2.450620e-03 2.251633e-03 2.903525e-03
##
   [61] 1.349846e-03 1.201201e-03 2.616614e-03 5.049441e-03 7.111077e-04
   [66] 7.798064e-04 8.077311e-04 8.405398e-04 2.148072e-03 1.267660e-03
##
    [71] 5.735516e-04 5.421576e-04 8.582765e-04 6.303721e-04 8.560743e-04
## [76] 5.641410e-04 1.308111e-03 2.842063e-03 2.473751e-03 1.820779e-03
```

```
## [81] 1.254773e-03 3.064689e-03 1.041480e-03 5.025200e-04 9.390189e-04
## [86] 3.813793e-04 9.495033e-04 7.061757e-04 5.423265e-04 3.662838e-04
## [91] 5.851532e-04 2.704361e-04 4.167494e-04 2.407281e-04 3.040823e-04
## [96] 3.078542e-04 2.012243e-04 3.767505e-04 4.049643e-04 2.595677e-04
## [101] 1.515378e-04 2.232387e-04 7.701786e-04 3.151487e-04 4.415835e-04
## [106] 3.153186e-04 2.381802e-04 3.197254e-04 1.487394e-04 2.033292e-04
## [111] 1.536916e-04 1.449262e-04 5.509603e-05 1.494313e-04 2.267442e-04
## [116] 1.066174e-04 1.534294e-04 9.076291e-05 2.367675e-04 1.480204e-04
## [121] 7.284860e-04 1.649761e-04 7.402076e-05 1.463542e-04 2.021607e-04
## [126] 3.032137e-04 6.259258e-04 1.826668e-04 2.165402e-04 1.640383e-04
## [131] 2.614879e-04 7.552764e-05 3.379458e-04 1.377233e-04 1.187753e-04
## [136] 1.577829e-04 1.258231e-04 2.551618e-04 4.109195e-04 3.085031e-04
## [141] 1.170915e-04 2.618829e-04 4.850049e-04 2.926694e-04 1.708315e-03
## [146] 4.391725e-04 4.993506e-04 4.789535e-04 3.367228e-04 3.920141e-04
## [151] 4.615271e-04 2.796273e-04 1.347975e-03 2.686054e-03 2.147038e-03
## [156] 1.583236e-03
 (c)
summary(Rday$rtn) # day summary
                 1st Qu.
                             Median
        Min.
                                          Mean
                                                  3rd Qu.
## -0.2203300 -0.0147847
                         0.0003130 0.0006583 0.0152928 0.2012290
summary(Rmon) # month summary
##
       Min. 1st Qu.
                       Median
                                  Mean
                                        3rd Qu.
                                                    Max.
## -0.54317 -0.05000 0.01630 0.01382 0.10340 0.29971
lnvart <- log(vart)</pre>
summary(vart)
##
                          Median
               1st Qu.
                                      Mean
                                             3rd Qu.
                                                          Max.
## 0.0000551 0.0002780 0.0004867 0.0007387 0.0008529 0.0050494
summary(lnvart)
     Min. 1st Qu. Median
                             Mean 3rd Qu.
##
## -9.806 -8.188 -7.628 -7.596 -7.067 -5.288
## N = 156 for var and ln \ var
 (e)
library(TSA)
##
## Attaching package: 'TSA'
## The following objects are masked from 'package:stats':
##
##
       acf, arima
## The following object is masked from 'package:utils':
##
##
       tar
eacf(lnvart, ar.max = 10, ma.max = 10)
```

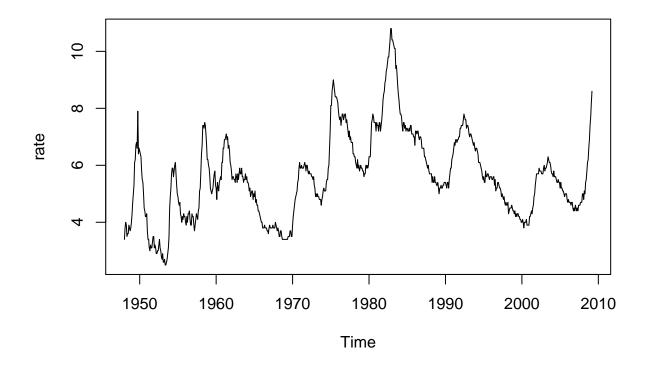
```
## AR/MA
##
     0 1 2 3 4 5 6 7 8 9 10
## 0 x x x x x x x x x x
## 1 x o o o o o o o o
## 2 x x o o o o o x o o o
## 3 x x o o o o o x o o o
## 4 x x x x o o o o o o
## 5 x x o x o o o o o o
## 6 x x o x o x o o o o
## 7 x x o x o o o o o o
## 8 x x o x x o o o o o
## 9 x o o x o o o x x o o
## 10 x o o x o o o x x o o
est <- arima(lnvart, order=c(13,0,1), include.mean = T)
##
## Call:
## arima(x = lnvart, order = c(13, 0, 1), include.mean = T)
## Coefficients:
##
                                                            ar7
                   ar2
                           ar3
                                   ar4
                                           ar5
                                                   ar6
                                                                     ar8
                                                                             ar9
           ar1
##
        0.4306 0.1408 0.1260 0.0073 0.1430 0.0954
                                                        -0.0707 -0.0911
                                                                         0.1127
## s.e.
       0.4137 0.1764 0.1094 0.1037 0.0904 0.1149
                                                         0.1024
                                                                  0.0972 0.1049
##
                  ar11
                          ar12
                                   ar13
                                             ma1
                                                 intercept
        0.0387 0.0305 0.1071
                                -0.1860
                                                    -7.4812
##
                                        -0.0442
## s.e. 0.0989 0.0891 0.0906
                                0.0854
                                          0.4206
                                                     0.3360
##
## sigma^2 estimated as 0.3011: log likelihood = -128.75, aic = 287.49
T = length(lnvart)
(pmax = floor(12 * (T/100)^(1/4)))
## [1] 13
estt = (0.9553-1)/0.028
estlnvart \leftarrow rep(0,142)
for (t in 15: 156){
 sum = 0
 for(i in 1:pmax){
    sum=sum+0.9553*(lnvart[t-i]-lnvart[t-i-1])
estlnvart[t-14] = -0.7596 + (0.9553-1)*lnvart[t-1] + sum + est$residuals[t]
}
estlnvart
    [1] -1.49214549 0.18192313 -0.05278762 -0.66450929 -0.13846083 0.40979482
    [7] -0.70210896 0.96111004 0.76245902 -0.53499277 -0.71343147 -1.10308754
##
  [13] 0.45845291 -0.15371806 -0.46274521 -0.09476048 -0.82401791 0.31129929
## [19] 0.05679269 -0.23355248 -0.20151057 -0.99526245 0.43867364 0.52225958
   [25] 0.19500560 0.55937957 -0.38214121 0.23753132 0.23758758 -0.74932591
##
   [31] -0.28162860 0.11552711 -0.55512658 -0.44413489 0.95253871 -0.06175974
## [37] -0.88689020 0.24260260 1.31874190 -0.41745226 -0.97879992 -1.30867042
## [43] 0.82810707 1.50751124 1.32120936 0.86943941 0.60340995 0.04528697
```

```
[49] -0.57225749 1.70571302 0.33836491 -2.00365490 -1.51584191 -0.54815287
##
   [61] -1.17485913 -0.97392403 -0.63732073 -0.64605976 1.34494995 0.74200790
##
   ##
##
   [73] -0.92524659 -0.31662203 -0.97246300 -1.04749132 -1.51159408 -2.63965438
   [79] -2.46752007 -2.55643595 -1.70715881 -2.64834673 -2.17986371 -0.87784866
##
   [85] -1.17785541 -0.62302625 -2.31805392 -1.88979686 -0.13538550 0.06017379
##
   ##
   [97] -0.74178031 -1.55246551 -2.60194023 -1.79523804 -0.16864911 -0.75894647
  [103] -2.37094438 -1.47504557 -1.18969150 -1.06535621
                                              0.64721204 -0.23818867
  [109] -1.24448789 -1.48339692 -0.13978437 0.03841552
                                              2.07215318
                                                        0.55098700
  [115] -0.60655371 -0.35653339 -0.46693629 -0.66469255 -0.85329391
                                                        0.21573742
## [121] -2.24326787 -1.13989336 -0.02981488 -0.15865522
                                               0.37820507
                                                        0.18815819
                                     0.58032871
## [127] -1.84604806 -0.56277649
                          0.39658895
                                               1.08172923
                                                        2.46412254
## [133] 0.02732161 0.49597977
                           0.43427525
                                     0.03893011
                                               0.63649590 -0.18935800
## [139] 0.75230860 2.07661394 3.13719293
                                     1.72914908
```

Problem 2

(a)

```
unem<-read.csv("m-unrate.txt",sep = "")
unts <- ts(unem$Rate, start=c(1948,1), freq = 12)
plot(unts, ylab='rate') #ts plot</pre>
```

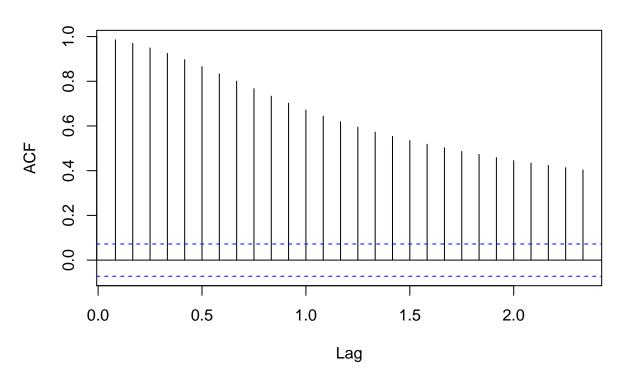


```
summary(unts) #summary
```

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 2.500 4.500 5.500 5.607 6.600 10.800
```

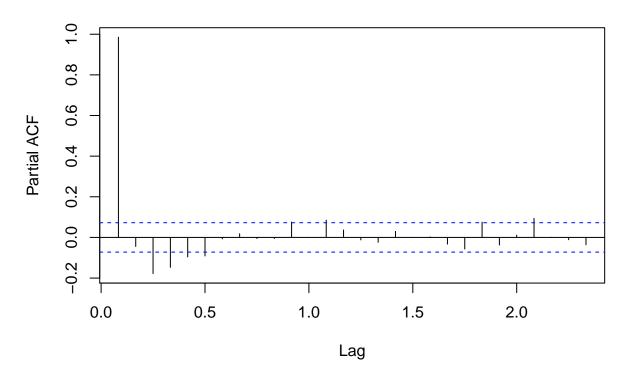
acf(unts) # acf plot

Series unts



pacf(unts)# pacf plot

Series unts



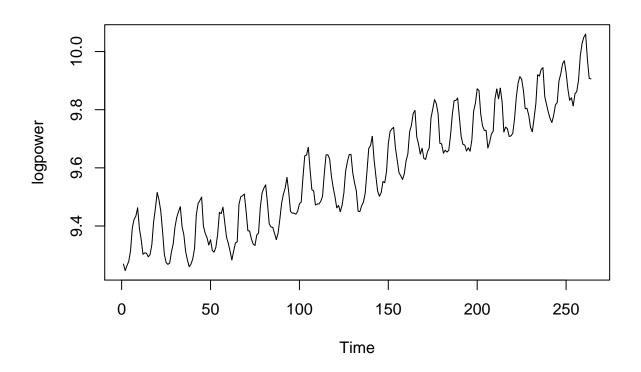
```
eacf(unts, ar.max = 10, ma.max = 10)
## AR/MA
##
     0 1 2 3 4 5 6 7 8 9 10
## 0 x x x x x x x x x x x
     x x x x x x x x x o x
     x x o o x o o o x o o
## 3
     x x o o x o x o o o
## 4
     x x o x x o o o o o
     x x o x x o o o o o
     o x o x x x o o o o o
     o x x o x x o o o o o
## 9 x o x x x o x x o o o
## 10 o o x x x o x x o o
 (b)
est1 <-arima(unts, order=c(2,0,2), include.mean = T)</pre>
est2 <-arima(unts, order=c(3,0,2), include.mean = T)#ARMA(3,2) perfered
est3 <-arima(unts, order=c(2,0,3), include.mean = T)</pre>
est1
##
## Call:
## arima(x = unts, order = c(2, 0, 2), include.mean = T)
## Coefficients:
```

```
##
                    ar2
                                     ma2 intercept
           ar1
                             ma1
##
        1.8351 -0.8424 -0.8517 0.2401
                                             5.6678
                                             0.3788
## s.e. 0.0361 0.0358 0.0494 0.0367
## sigma^2 estimated as 0.03958: log likelihood = 141.6, aic = -273.19
est2
##
## Call:
## arima(x = unts, order = c(3, 0, 2), include.mean = T)
## Coefficients:
                            ar3
           ar1
                    ar2
                                             ma2 intercept
                                     ma1
        2.6172 -2.3549 0.7345 -1.6062 0.8235
##
                                                     5.6520
## s.e. 0.0606 0.1273 0.0680 0.0624 0.0667
                                                     0.4862
## sigma^2 estimated as 0.03946: log likelihood = 142.55, aic = -273.1
##
## Call:
## arima(x = unts, order = c(2, 0, 3), include.mean = T)
##
## Coefficients:
##
           ar1
                   ar2
                           ma1
                                   ma2
                                           ma3 intercept
        0.0382 0.9388 1.0027 0.2691 0.2567
                                                   5.8949
## s.e. 0.0175 0.0174 0.0409 0.0465 0.0325
                                                   0.7940
## sigma^2 estimated as 0.04241: log likelihood = 115.9, aic = -219.8
Box.test(est1$residuals, lag = 13, type = "Ljung-Box", fitdf=2+2)
##
## Box-Ljung test
##
## data: est1$residuals
## X-squared = 28.365, df = 9, p-value = 0.0008286
Box.test(est2$residuals, lag = 13, type = "Ljung-Box", fitdf=2+3)
##
## Box-Ljung test
##
## data: est2$residuals
## X-squared = 38.088, df = 8, p-value = 7.252e-06
Box.test(est3$residuals, lag = 13, type = "Ljung-Box", fitdf=3+2)
## Box-Ljung test
## data: est3$residuals
## X-squared = 78.865, df = 8, p-value = 8.271e-14
 (c)
```

```
pred.dat <- predict(est2, n.ahead=4)</pre>
pred.dat
## $pred
##
             Apr
                       May
                                Jun
                                          Jul
## 2009 8.770181 8.934577 9.065678 9.146673
##
## $se
##
                         May
                                              Jul
                                   Jun
              Apr
## 2009 0.1986400 0.2824785 0.3589321 0.4388648
 (d) yes, since he AR polynomial relate to the behavior of stochastic cycles.
 (e)
library(FinTS)
## Loading required package: zoo
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
dif12 = diff(unts, lag=12)
dif = diff(unts); ddif12 = diff(dif, lag=12)
fit <- ARIMA(dif, order=c(3,0,2), seasonal=list(order=c(0,0,2), period=6), Box.test.lag=15)
fit$Box.test
##
##
    Box-Ljung test (lag = 15)
## data: fit$resid
## X-squared = 11.227, df = 8, p-value = 0.1892
## the corresponding period is 6
```

Problem 3

```
ele<-read.csv("power6.txt")
elets <- ts(ele$power)
plot(elets, ylab='logpower')</pre>
```



```
summary(elets)
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                               Max.
             9.440
                     9.576
                             9.590
                                      9.741
                                            10.060
eacf(elets, ar.max = 10, ma.max = 10)
## AR/MA
      0 1 2 3 4 5 6 7 8 9 10
     x x x x x x x x x x x
     x x o x x x x x o x x
     x x o x o o x x o o o
     x o x x o o o x x o o
      x x x x o o o x x o o
## 5
     x o x x o o o x o o o
     x \circ o x x x x x o x o
     x x o x x o o x o o o
     x o x x x o x x o o o
## 9 x x x x x x x x o x o
## 10 x x o x x x x x o x o
est <-arima(unts, order=c(2,0,2), include.mean = T)
est
##
## Call:
## arima(x = unts, order = c(2, 0, 2), include.mean = T)
```

```
## Coefficients:
       ar1 ar2 ma1 ma2 intercept
##
       1.8351 -0.8424 -0.8517 0.2401 5.6678
## s.e. 0.0361 0.0358 0.0494 0.0367
                                           0.3788
## sigma^2 estimated as 0.03958: log likelihood = 141.6, aic = -273.19
Box.test(est$residuals, lag = 13, type = "Ljung-Box", fitdf=2+2)
##
## Box-Ljung test
##
## data: est$residuals
## X-squared = 28.365, df = 9, p-value = 0.0008286
fit <- ARIMA(dif, order=c(2,0,2),seasonal=list(order=c(0,1,1), period=11), Box.test.lag=14)
fit$Box.test
##
## Box-Ljung test (lag = 14)
##
## data: fit$resid
## X-squared = 31.864, df = 9, p-value = 0.0002102
```