ba2500_TS_HW6

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```
getSymbols("VALE", from="2006-01-01")

## As of 0.4-0, 'getSymbols' uses env=parent.frame() and
## auto.assign=TRUE by default.
```

```
## 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] "VALE"
```

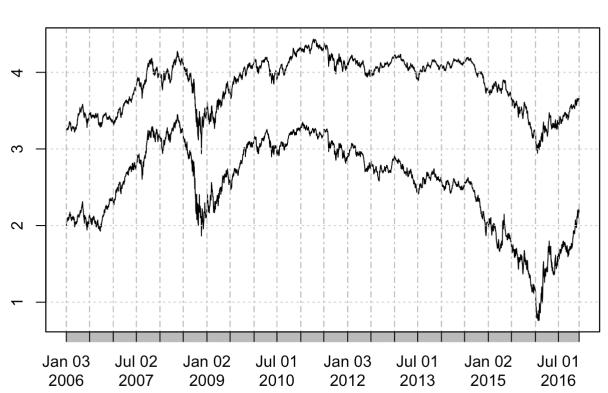
```
getSymbols("BHP", from="2006-01-01")
```

```
## [1] "BHP"
```

```
y1 = log(BHP[,6])
y2 = log(VALE[,6])

plot(y1, ylim=range(c(y1,y2)), ylab = "")
#second plot EDIT: needs to have same ylim
par(new = TRUE)
plot(y2, ylim=range(c(y1,y2)), axes = FALSE, xlab = "", ylab = "")
```





```
data.hw6 = data.frame(log(BHP[,6]), log(VALE[,6])) \#p1=log(BHP[,6]) and p2=log(VALE[,6])
n = nrow(data.hw6)
t = 200
coeff = NULL
error = matrix(0, ncol=(n-t), nrow=t)
a = numeric(ncol(error))
h = numeric(length(a))
s = NULL
r Ts = NULL
sigma = NULL
delta = NULL
for(i in 1:(n-t)){
    temp1 = data.hw6[i:(i+t-1),]
   obj = lm(temp1[,1] \sim temp1[,2], data = temp1)
   coeff = rbind.data.frame(coeff, coef(obj))
    names(coeff) = c("alpha", "beta")
   error[,i] = obj$residuals
    a[i] = coef(lm(error[2:t, i] ~ 0 + error[1:(t-1), i]))
   h[i] = log(0.5)/abs(a[i])
    s_h = ifelse(h[i] >= 1, floor(h[i]), 1)
    error T = tail(error[,i], 1)
    if(error T>0){
      temp2 = -0.5*(data.hw6[(i+t-1+s_h),1] - data.hw6[(i+t-1),1]) +
               0.5*(data.hw6[(i+t-1+s_h),2] - data.hw6[(i+t-1),2])
    } else {
      temp2 = 0.5*(data.hw6[(i+t-1+s h),1] - data.hw6[(i+t-1),1]) -
              0.5*(data.hw6[(i+t-1+s h),2] - data.hw6[(i+t-1),2])
    s = c(s, s h)
    r Ts = c(r Ts, temp2)
   temp3 = var(error[,i])
    sigma = c(sigma, temp3)
    temp4 = abs(tail(error[,i], 1))/sqrt(temp3)
   delta = c(delta, temp4)
   rm(temp1, obj, temp2, error_T, s_h, temp3, temp4)
}
data.resume = cbind.data.frame(coeff, a, h, s, r Ts, sigma, delta)
head(data.resume)
```

```
##
        alpha
                   beta
                                           h s
                                                         r_Ts
                                                                    sigma
## 1 1.945640 0.6898287 0.9745318 -0.7112617 1 0.0021221440 0.005161027
## 2 1.960928 0.6827064 0.9746657 -0.7111641 1 -0.0030017683 0.005117595
## 3 1.969618 0.6787762 0.9722651 -0.7129199 1 -0.0004944919 0.005078688
## 4 1.980400 0.6738983 0.9729582 -0.7124121 1 -0.0015535618 0.005013557
## 5 1.986213 0.6714032 0.9734354 -0.7120628 1 0.0279978466 0.004959784
## 6 1.996056 0.6668561 0.9710257 -0.7138299 1 -0.0002435239 0.004913481
##
          delta
## 1 0.08011858
## 2 0.04769191
## 3 0.19974860
## 4 0.17699591
## 5 0.26917266
## 6 0.32108152
```

tail(data.resume)

```
##
           alpha
                      beta
                                              h s
                                                           r Ts
## 2550 2.522120 0.5337199 0.9390921 -0.7381035 1 -0.013906169 0.002433588
## 2551 2.517771 0.5361183 0.9405981 -0.7369218 1 -0.003486021 0.002416522
## 2552 2.515469 0.5373167 0.9410697 -0.7365524 1 -0.013775845 0.002406297
## 2553 2.515072 0.5373245 0.9435148 -0.7346437 1 -0.011972135 0.002405859
## 2554 2.517251 0.5358255 0.9432576 -0.7348440 1 0.011898343 0.002413906
## 2555 2.515155 0.5368734 0.9415400 -0.7361845 1 0.003601753 0.002409673
##
            delta
## 2550 0.2084866
## 2551 0.4717257
## 2552 0.3773384
## 2553 0.7693311
## 2554 0.8653116
## 2555 0.6152449
```

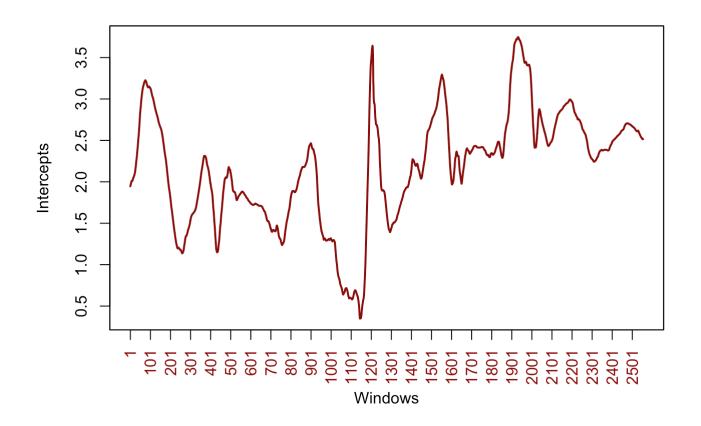
```
error = as.data.frame(error)
names(error) = paste("Residuals W", 1:ncol(error), sep="_")
head(error[1:6])
```

```
Residuals W 1 Residuals W 2 Residuals W 3 Residuals W 4 Residuals W 5
##
## 1
       -0.09251079
                     -0.08879261
                                   -0.11411946
                                                 -0.10485434
                                                               -0.09836423
## 2
       -0.08792787
                     -0.11341171
                                   -0.10404710
                                                 -0.09768665
                                                               -0.11130693
## 3
      -0.11258882
                     -0.10339387
                                   -0.09694401
                                                 -0.11062388
                                                               -0.10906963
## 4
       -0.10266979
                     -0.09634285
                                   -0.10987053
                                                 -0.10842467
                                                               -0.14734012
## 5
       -0.09571310
                    -0.10926073
                                  -0.10774579
                                                -0.14676807
                                                               -0.13680250
## 6
       -0.10861535
                     -0.10719600
                                   -0.14623174
                                                 -0.13621731
                                                               -0.11138095
##
    Residuals W 6
## 1
       -0.1118010
## 2
       -0.1094943
## 3
       -0.1476319
       -0.1371183
## 4
## 5
        -0.1117305
## 6
        -0.1171587
```

```
tail(error[1:6])
```

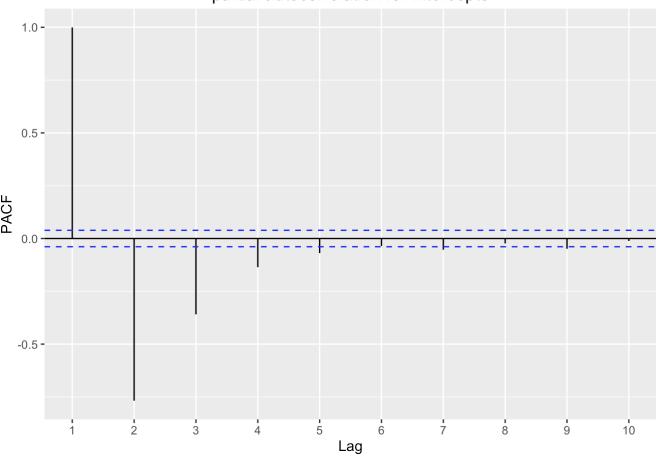
```
##
      Residuals W_1 Residuals W_2 Residuals W_3 Residuals W_4 Residuals W_5
## 195 -0.034210684 -0.044608875
                                  -0.027836617 -0.009864911
                                                              0.012444030
## 196 -0.044240608 -0.027467798 -0.009396344
                                                 0.012926572
                                                              0.004170778
## 197 -0.027259117 -0.009016007
                                   0.013287931
                                                 0.004703978
                                                              0.002119857
## 198 -0.008786453
                      0.013581885
                                   0.005164371
                                                 0.002636403
                                                              0.013409291
## 199
        0.013654900
                      0.005538122
                                   0.003064237
                                                 0.013885699
                                                              0.012038126
## 200
        0.005755741
                      0.003411754
                                   0.014235068
                                                 0.012532456
                                                              0.018956682
##
      Residuals W_6
## 195
        0.003949772
## 196
        0.001929203
## 197
        0.013291782
## 198  0.011887959
## 199 0.018859257
## 200 -0.022506603
```

```
data.resume = zoo(data.resume)
plot(data.resume$alpha, xaxt="n", xlab="Windows", ylab="Intercepts", lwd=2, col="darkre
d")
x = seq(1, (n-t), by=100)
axis(1, at=x, labels=x, col.axis="darkred", las=2)
```

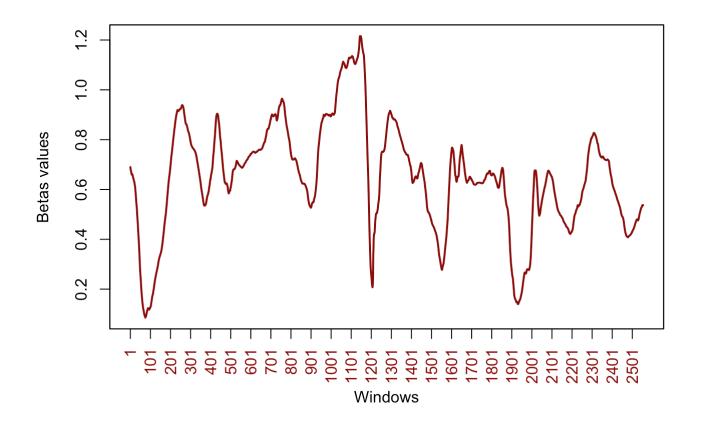


#fBasics::pacfPlot(data.resume\$alpha, lag.max=10)
forecast::ggPacf(data.resume\$alpha, lag.max=10, main="partial autocorrelation for Interc
epts")

partial autocorrelation for Intercepts

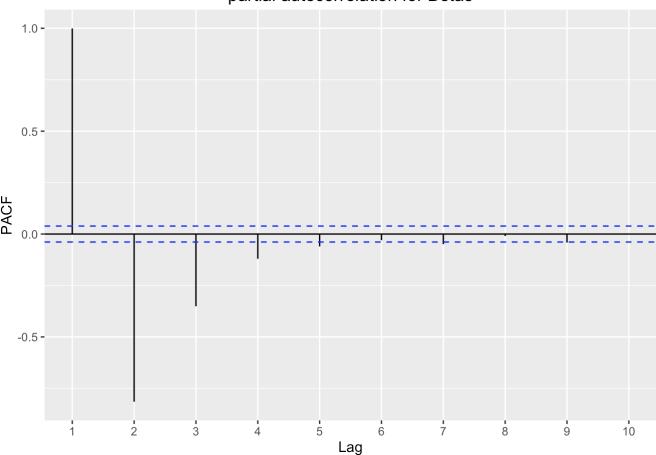


```
plot(data.resume$beta, xaxt="n", xlab="Windows", ylab="Betas values", lwd=2, col="darkre
d")
x = seq(1, (n-t), by=100)
axis(1, at=x, labels=x, col.axis="darkred", las=2)
```

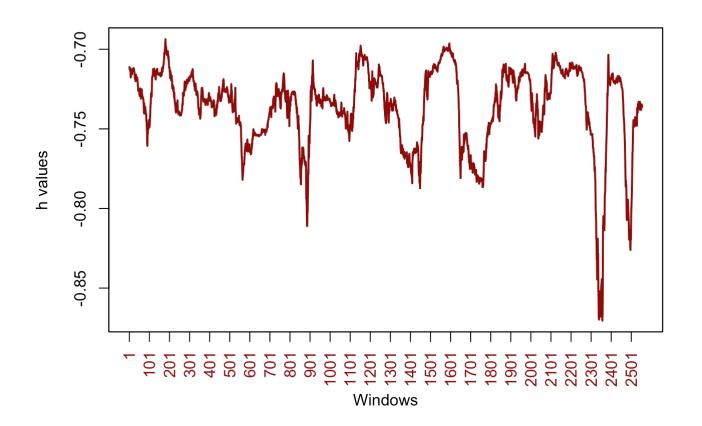


#fBasics::pacfPlot(coeff\$beta, lag.max=10)
forecast::ggPacf(data.resume\$beta, lag.max=10, main="partial autocorrelation for Betas")

partial autocorrelation for Betas

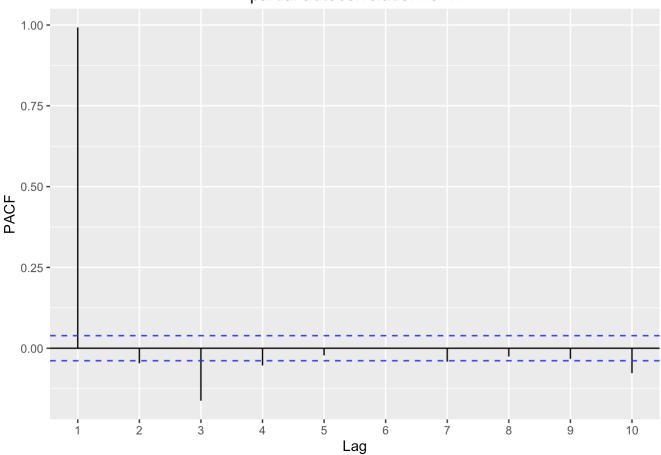


```
plot(data.resume$h, xaxt="n", xlab="Windows", ylab="h values", lwd=2, col="darkred")
x = seq(1, (n-t), by=100)
axis(1, at=x, labels=x, col.axis="darkred", las=2)
```

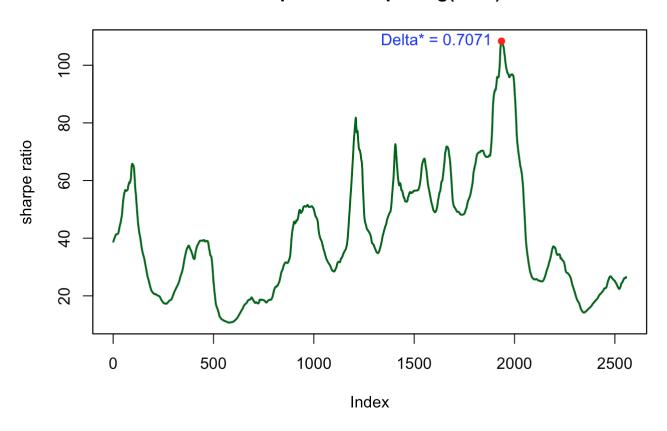


#fBasics::pacfPlot(data.resume\$h, lag.max=10)
forecast::ggPacf(data.resume\$h, lag.max=10, main="partial autocorrelation for h")

partial autocorrelation for h



Sharpe ratio for p1=log(BHP)



Sharpe ratio for p2=log(VALE)

