

Test Exercise 6

```
In [152]: data <- read.csv("./data/TestExer6-CARS-round2.csv")
names(data)[1] <- "YEARMONTH"A

n <- nrow(data)

data$T <-c(seq(1, n))
data$YEAR <- substr(data$YEARMONTH, 1, 4)

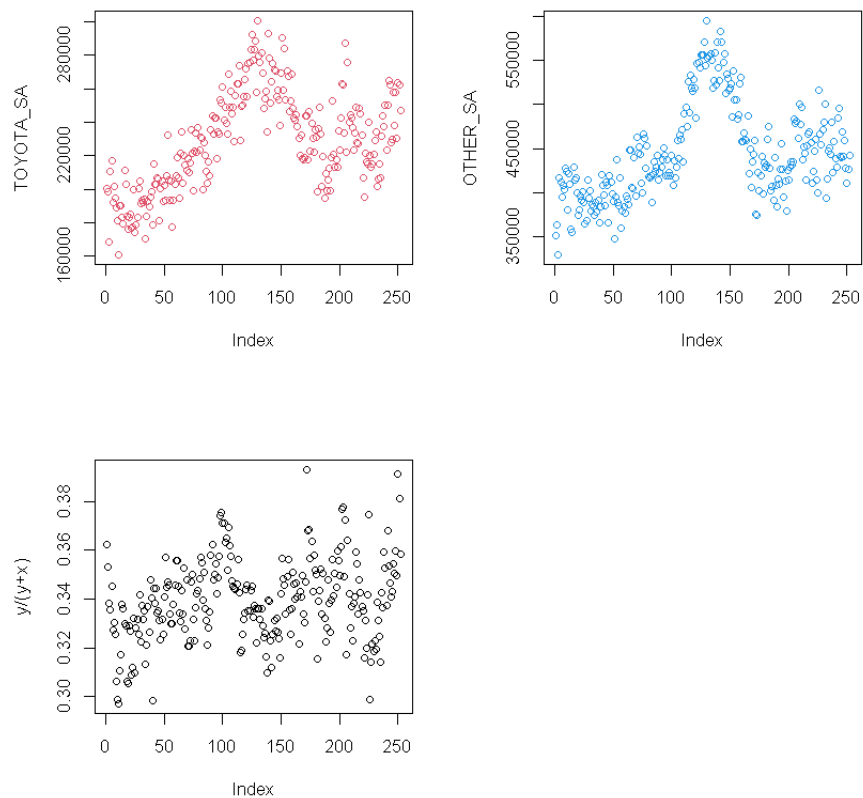
data$DTOYOTA_SA <- c(NA, diff(data$TOYOTA_SA, lag=1))
data$DOTHER_SA <- c(NA, diff(data$OTHER_SA, lag=1))0
for (i in c(1:12)){
  data[,paste("DTOYOTA_SA_1", i, sep="")] <- c(c(rep(NA, i)), data$DTOYOTA_SA[-((n-i+1):n)])
}
for (i in c(1:3)){
  data[,paste("DOTHER_SA_1", i, sep="")] <- c(c(rep(NA, i)), data$DOTHER_SA[-((n-i+1):n)])
}

data$TOYOTA_SA_l1 <- c(c(rep(NA, 1)), data$TOYOTA_SA[-((n-0):n)])
data$OTHER_SA_l1 <- c(c(rep(NA, 1)), data$OTHER_SA[-((n-0):n)])
train <- data[data$YEAR <= 1999,]
test <- data[data$YEAR > 1999,]"
```

Part (a)

y_t and x_t seems to be correlated. $\frac{y_t}{(y_t+x_t)}$ is rather stationary.

```
In [153]: # attach(mtcars)
par(mfrow=c(2,2))
plot(ata$TOYOTA_SA, ylab="TOYOTA_SA", col=2)
plot(data$OTHER_SA, ylab="OTHER_SA", col=4)
plot(data$TOYOTA_SA / (data$TOYOTA_SA + data$OTHER_SA), ylab="y/(y+x)
)")
```



Part (b)

```
In [154]: TOYOTA_SA_ADF <- lm(DTOYOTA_SA~ TTOYOTA_SA_l1 + DTOYOTA_SA_l1 + DTOYOTA_SA_l2 + DTOYOTA_SA_l3, data=trin)

sprintf("Formula: DTOYOTA_SA = %.3f - %.3f*TOYOTA_SA_l1* - %.3f*DTOYOTA_SA_l1* - %.3f*DTOYOTA_SA_l2* - %.3f*DTOYOTA_SA_l3*", coef(TOYOTA_SA_ADF)[1], -1*coef(TOYOTA_SA_ADF)[2], -1*coef(TOYOTA_SA_ADF)[3], -1*coef(TOYOTA_SA_ADF)[4], -1*coef(TOYOTA_SA_ADF)[5])
sprintf("")sprintf("LOTOYOTA_SA_l1:")
sprintf("Coefficient = %.5f", sumy(LOGTOYOTA_SA_ADF)$coefficient[2, "Estimate"])
sprintf("td. Error = %.5f", sumy(LOGITTOYOTA_SA_ADF)$coefficient[2, "Std. Error"])
sprintf("t-valu = %.5f > -32.9 Do not rejec LOGIPtTOYOTA_SAis non stationary.", y(LOGIPmTOYOTA_SAADF))coefficient[32 "t value"]])
```

'Formula: DTOYOTA_SA = 19281.893 - 0.083*TOYOTA_SA_l1* - 0.563*DTOYOTA_SA_l1* - 0.324*DTOYOTA_SA_l2* - 0.064*DTOYOTA_SA_l3*'

''

'TOYOTA_SA_l1:'

'Coefficient = -0.08322'

'Std. Error = 0.03679'

't-value = -2.26228 > -2.9. Do not reject H0. TOYOTA_SA is non stationary.'

```
In [155]: OTHER_SA_ADF <- lm(DOTHER_SA ~ OTHER_SA_l1 + DOTHER_SA_l1 + DOTHER_SA_l2 + DOTHER_SA_l3, data=train)
sprintf("Formula: DOTHER_SA = %.3f - %.3f*OTHER_SA_l1* - %.3f*DOTHER_SA_l1* - %.3f*DOTHER_SA_l2* - %.3f*DOTHER_SA_l3*", coef(OTHER_SA_ADF)[1], -1*coef(OTHER_SA_ADF)[2], -1*coef(OTHER_SA_ADF)[3], -1*coef(OTHER_SA_ADF)[4], -1*coef(OTHER_SA_ADF)[5])
sprintf("")
sprintf("OTHER_SA_l1:")
sprintf("Coefficient = %.5f", summary(OTHER_SA_ADF)$coefficient[2, "Estimate"])
sprintf("Std. Error = %.5f", summary(OTHER_SA_ADF)$coefficient[2, "Std. Error"])
sprintf("t-value = %.5f > -2.9. Do not reject H0. OTHER_SA is non stationary.",summary(OTHER_SA_ADF)$coefficient[2, "t value"]])
```

Error in lm.fit(x, y, offset = offset, singular.ok = singular.ok, ...): 0 (non-NA) cases

Traceback:

```
1. lm(DOTHER_SA ~ OTHER_SA_l1 + DOTHER_SA_l1 + DOTHER_SA_l2 + DOTHER_SA_l3,
. data = train)
2. lm.fit(x, y, offset = offset, singular.ok = singular.ok, ...)
3. stop("0 (non-NA) cases")
```

Part (c)

```
In [156]: eg <- lm(TOYOTA_SA ~ OTHER_SA, data=train)
          sprintf("Formula: *TOYOTA_SA* = %.2f + %.2f*OTHER_SA*", coef(eg)[1]
            , coef(eg)[2])
```

'Formula: *TOYOTA_SA* = 26786.41 + 0.45*OTHER_SA'

```
In [157]: res_data <- train
          res_data$e <- summary(eg)$residuals
          res_data$De <- c(NA, diff(res_data$e, lag=1))

          n <- nrow(res_data)
          res_data$e_l1 <- c(c(rep(NA, 1)), res_data$e[-((n-0):n)])
          res_data$De_l1 <- c(c(rep(NA, 1)), res_data$De[-((n-0):n)])
          res_data$De_l2 <- c(c(rep(NA, 2)), res_data$De[-((n-1):n)])
          res_data$De_l3 <- c(c(rep(NA, 3)), res_data$De[-((n-2):n)])
```

```
In [158]: res_fit <- lm(De ~ e_l1 + De_l1 + De_l2 + De_l3, data=res_data)
          sprintf("Formula: De = %.3f - %.3f*e_l1* - %.3f*De_l1* - %.3f*De_l2
            * - %.3f*De_l3*", coef(res_fit)[1], -1*coef(res_fit)[2], -1*coef(re
              s_fit)[3], -1*coef(res_fit)[4], -1*coef(res_fit)[5])
          sprintf("t-value = %.5f < -3.4. Reject H0. TOYOTA_SA and OTHER_SA a
            re cointegrated.", summary(res_fit)$coefficient[3, "t value"])
```

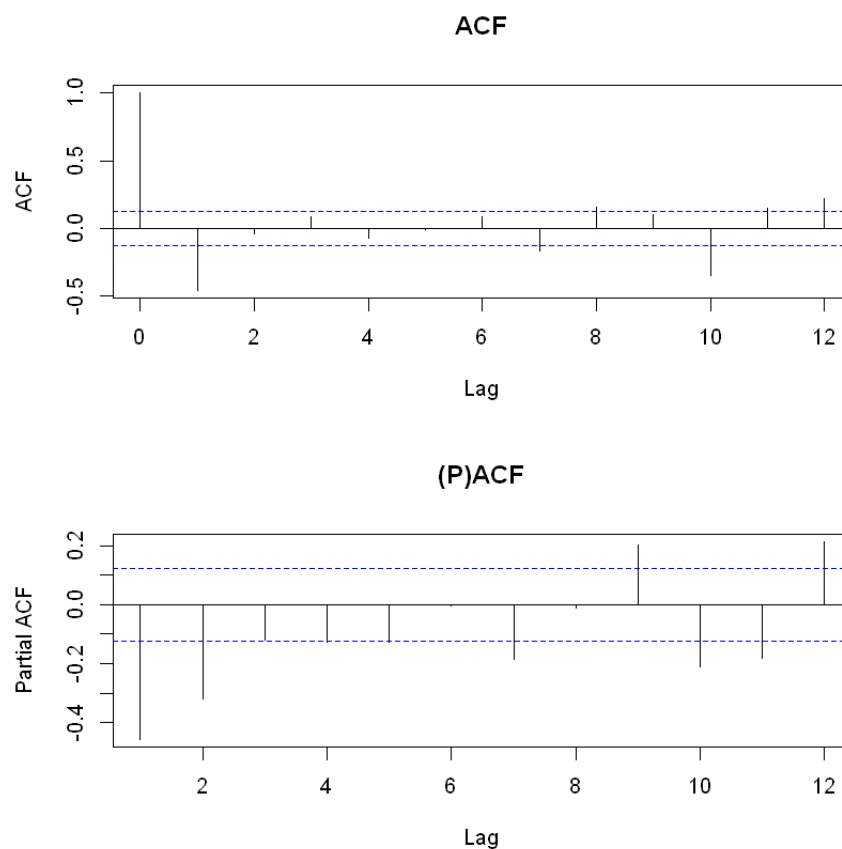
'Formula: De = 24.992 - 0.293*e_l1* - 0.286*De_l1* - 0.142*De_l2* - 0.096*De_l3'

't-value = -3.63959 < -3.4. Reject H0. TOYOTA_SA and OTHER_SA are cointegrated.'

Part (d)

From the graph and OLS, only lags 1 to 5, 10 and 12 are significant.

```
In [159]: par(mfrow=c(2,1))  
          acf(na.omit(data$DTOYOTA_SA), lag.max=12, main="ACF")  
          pacf(na.omit(data$DTOYOTA_SA), lag.max=12, main="(P)ACF")
```



```
In [160]: evaluate <- lm(DTOYOTA_SA ~ DTOYOTA_SA_l1 + DTOYOTA_SA_l2 + DTOYOTA  
_SA_l3 + DTOYOTA_SA_l4 + DTOYOTA_SA_l5 + DTOYOTA_SA_l6 + DTOYOTA_SA  
_l7 + DTOYOTA_SA_l8 + DTOYOTA_SA_l9 + DTOYOTA_SA_l10 + DTOYOTA_SA_l  
11 + DTOYOTA_SA_l12, data=train)
```

```
In [161]: summary(evaluate)
```

Call:

```
lm(formula = DTOYOTA_SA ~ DTOYOTA_SA_l1 + DTOYOTA_SA_l2 + DTOYOTA_SA_l3 +
    DTOYOTA_SA_l4 + DTOYOTA_SA_l5 + DTOYOTA_SA_l6 + DTOYOTA_SA_l7 +
    DTOYOTA_SA_l8 + DTOYOTA_SA_l9 + DTOYOTA_SA_l10 + DTOYOTA_SA_l11 +
    DTOYOTA_SA_l12, data = train)
```

Residuals:

Min	1Q	Median	3Q	Max
-40387	-9196	-402	8322	33993

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	619.04050	915.32772	0.676	0.499577
DTOYOTA_SA_l1	-0.61619	0.06668	-9.241	< 2e-16 ***
DTOYOTA_SA_l2	-0.30230	0.07886	-3.833	0.000166 ***
DTOYOTA_SA_l3	-0.25794	0.07931	-3.252	0.001331 **
DTOYOTA_SA_l4	-0.26978	0.08119	-3.323	0.001049 **
DTOYOTA_SA_l5	-0.23153	0.08357	-2.770	0.006091 **
DTOYOTA_SA_l6	-0.12142	0.08446	-1.438	0.152006
DTOYOTA_SA_l7	-0.13116	0.08437	-1.555	0.121516
DTOYOTA_SA_l8	0.04496	0.08331	0.540	0.589974
DTOYOTA_SA_l9	0.03573	0.08186	0.437	0.662905
DTOYOTA_SA_l10	-0.26465	0.07981	-3.316	0.001073 **
DTOYOTA_SA_l11	-0.04327	0.07877	-0.549	0.583349
DTOYOTA_SA_l12	0.21971	0.06624	3.317	0.001070 **

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 13740 on 214 degrees of freedom

(13 observations deleted due to missingness)

Multiple R-squared: 0.4642, Adjusted R-squared: 0.4342

F-statistic: 15.45 on 12 and 214 DF, p-value: < 2.2e-16

```
In [162]: fit <- lm(DTOYOTA_SA ~ DTOYOTA_SA_l1 + DTOYOTA_SA_l2 + DTOYOTA_SA_l3 +
    DTOYOTA_SA_l4 + DTOYOTA_SA_l5 + DTOYOTA_SA_l10 + DTOYOTA_SA_l12,
    data=train)
sprintf("Formula:")
sprintf("Dyt = %.3f - %.3f*Dyt_l1* - %.3f*Dyt_l2* - %.3f*Dyt_l3* -
    %.3f*Dyt_l4* - %.3f*Dyt_l5* - %.3f*Dyt_l10* + %.3f*Dyt_l12*", coef(
    fit)[1], -1*coef(fit)[2], -1*coef(fit)[3], -1*coef(fit)[4], -1*coef(
    fit)[5], -1*coef(fit)[6], -1*coef(fit)[7], coef(fit)[8])
```

'Formula:'

'Dyt = 561.613 - 0.598*Dyt_l1* - 0.263*Dyt_l2* - 0.227*Dyt_l3* - 0.230*Dyt_l4* -
0.152*Dyt_l5* - 0.268*Dyt_l10* + 0.246*Dyt_l12*'

Part (e)

```
In [163]: new_data <- data
new_data$EQ <- new_data$TOYOTA_SA_l1 - coef(eg)[2]*new_data$OTHER_S
A_l1

new_train <- new_data[new_data$YEAR <= 1999,]
new_test <- new_data[new_data$YEAR > 1999,]
```

```
In [164]: ECM_fit <- lm(DTOYOTA_SA ~ EQ + DTOYOTA_SA_l1 + DTOYOTA_SA_l2 + DTO
YOTA_SA_l3 + DTOYOTA_SA_l4 + DTOYOTA_SA_l5 + DTOYOTA_SA_l10 + DTOYO
TA_SA_l12, data=new_train)
sprintf("Formula:")
sprintf("Dyt = %.3f - %.3f(*yt_l1* - 0.45*xt_l1*) - %.3f*Dyt_l1* -
%.3f*Dyt_l2* - %.3f*Dyt_l3* - %.3f*Dyt_l4* - %.3f*Dyt_l5* - %.3f*Dy
t_l10* + %.3f*Dyt_l12*", coef(ECM_fit)[1], -1*coef(ECM_fit)[2], -1*
coef(ECM_fit)[3], -1*coef(ECM_fit)[4], -1*coef(ECM_fit)[5], -1*coef
(ECM_fit)[6], -1*coef(ECM_fit)[7], coef(ECM_fit)[8], coef(ECM_fit)[
9])
```

'Formula:'

'Dyt = 4602.114 - 0.150(*yt_l1* - 0.45*xt_l1*) - 0.523*Dyt_l1* - 0.187*Dyt_l2* -
0.158*Dyt_l3* - 0.185*Dyt_l4* - 0.133*Dyt_l5* - -0.274*Dyt_l10* + 0.252*Dyt_l12*'

```
In [165]: rsq_0 <- as.numeric(summary(fit)$r.squared)
rsq_1 <- as.numeric(summary(ECM_fit)$r.squared)
n <- nrow(na.omit(train))
g <- 1 #no. of excluded variables
k <- 9 #total number of variables in unrestricted model
F5 <- (rsq_1 - rsq_0) * (n - k) / (1 - rsq_1) / g

sprintf("Critical value at 5%% level=%.3f", qf(.95, df1=g, df2=n-k)
)
sprintf("Critical value at 1%% level=%.3f", qf(.99, df1=g, df2=n-k)
)

sprintf("F = %.3f > 3.9, therefore reject H0 at 5%% level", F)
sprintf("F = %.3f < 6.8, therefore do not reject H0 at 1%% level",
F)
```

Warning message in qf(0.95, df1 = g, df2 = n - k):
"NaNs produced"

'Critical value at 5% level=NaN'

Warning message in qf(0.99, df1 = g, df2 = n - k):
"NaNs produced"

'Critical value at 1% level=NaN'

'F = 4.630 > 3.9, therefore reject H0 at 5% level'

'F = 4.630 < 6.8, therefore do not reject H0 at 1% level'

Part (f)

```
In [166]: result <- test[,c("YEAR", "DToyota_SA")]
```

```
In [167]: result$AR <- predict(fit, newdata=new_test)
result$ECM <- predict(ECM_fit, newdata=new_test)
```

```
In [168]: result_error <- result
result_error$AR <- result_error$DToyota_SA - result_error$AR
result_error$ECM <- result_error$DToyota_SA - result_error$ECM
```



```
In [169]: n <- nrow(result_error)

for (col in c("AR", "ECM")){
  values <- result_error[, col]
  print(col)
  print(eval(sprintf("RMSE = %.5f", sqrt(1/n*sum(values**2)))))
  print(eval(sprintf("MAE = %.5f", 1/n*sum(abs(values)))))
  print("")
}

[1] "AR"
[1] "RMSE = 16991.79876"
[1] "MAE = 14703.21887"
[1] ""
[1] "ECM"
[1] "RMSE = 18202.60915"
[1] "MAE = 15552.91652"
[1] ""
```

AR model (without the Error Correction) does better in the out-of-sample set with lower RMSE and MAE.

In []:

In []: