Macro Economic Time Series Analysis

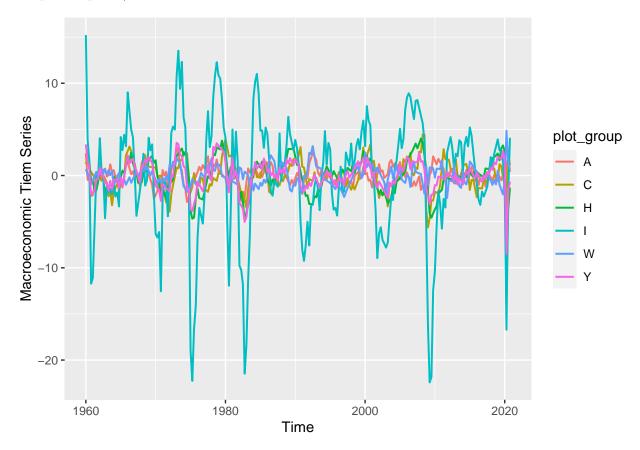
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```
library("mFilter")
library("ggplot2")
library("ggfortify")
library("zoo")
FALSE
FALSE Attaching package: 'zoo'
FALSE The following objects are masked from 'package:base':
FALSE
FALSE
          as.Date, as.Date.numeric
Ques 1: Business Cycle Facts
Transforming the data
## Uploading and Transforming Data
data = read.csv("C:/Users/arora/Desktop/Second Year/Quarter 3/Macroeconomics 3/HW1/Final Data.csv")
## data for Y
loggdp = round(log(data[,2]/data[,9]),2) ## log of GDP per capita
tloggdp = ts(loggdp, start = c(1960,1), frequency = 4) ## transforming into a time series
hpgdp = hpfilter(tloggdp) ## applying hp filter
Yc = hpgdp$cycle*100 ## cyclical part obtain from HP filter
## data for C
pce = round(log(data[,3]/(data[,8]*data[,9])),2) ## transforming personal expenditure data
tpce = ts(pce, start = c(1960,1), frequency = 4) ## transforming into a time series
hppce = hpfilter(tpce) ## applying hp filter
Cc = hppce$cycle*100 ## cyclical part obtain from HP filter
## data for I
inv = round(log(data[,4]/(data[,9])),2) ## transforming investment data
tinv = ts(inv, start = c(1960,1), frequency = 4) ## transforming into a time series
hpinv = hpfilter(tinv) ## applying hp filter
Ic = hpinv$cycle*100 ## cyclical part obtain from HP filter
## data for H
hours = round(log(data[,5]*10000/(data[,9])),2) ## transforming number of hours data
thours = ts(hours, start = c(1960,1), frequency = 4) ## transforming into a time series
hphours = hpfilter(thours) ## applying hp filter
```

```
Hc = hphours$cycle*100 ## cyclical part obtain from HP filter
## data for W
comp = round(log(data[,6]),2) ## transforming compensation per hour data
tcomp = ts(comp, start = c(1960,1), frequency = 4) ## transforming into a time series
hpcomp = hpfilter(tcomp) ## applying hp filter
Wc = hpcomp$cycle*100 ## cyclical part obtain from HP filter
## data for A
tech = round(log(data[,7]),2) ## transforming compensation per hour data
ttech = ts(tech, start = c(1960,1), frequency = 4) ## transforming into a time series
hptech = hpfilter(ttech) ## applying hp filter
Ac = hptech$cycle*100 ## cyclical part obtain from HP filter
## converting back time series into dataframe for the plots
Y = data.frame(Y=as.matrix(Yc), date=time(Yc))
C = data.frame(C=as.matrix(Cc), date=time(Cc))
I = data.frame(I=as.matrix(Ic), date=time(Ic))
H = data.frame(H=as.matrix(Hc), date=time(Hc))
W = data.frame(W=as.matrix(Wc), date=time(Wc))
A = data.frame(A=as.matrix(Ac), date=time(Ac))
```

The plot for part a)



Calculating the statistics for all the variables

```
## Autocorrelations
autoY = cor(Y[1:nrow(Y)-1,1],Y[2:nrow(Y),1])
autoC = cor(C[1:nrow(C)-1,1],Y[2:nrow(C),1])
autoI = cor(I[1:nrow(I)-1,1],Y[2:nrow(I),1])
autoH = cor(H[1:nrow(H)-1,1],Y[2:nrow(H),1])
autoW = cor(W[1:nrow(W)-1,1],Y[2:nrow(W),1])
autoA = cor(A[1:nrow(A)-1,1],Y[2:nrow(A),1])
## Correlations with Y
CY = cor(Y[,1],C[,1])
IY = cor(Y[,1],I[,1])
HY = cor(Y[,1],H[,1])
WY = cor(Y[,1],W[,1])
AY = cor(Y[,1],A[,1])
## Standard deviations
sdY = sd(Y[,1])
sdC = sd(C[,1])
sdI = sd(I[,1])
sdH = sd(H[,1])
sdW = sd(W[,1])
sdA = sd(A[,1])
## Relative standard deviation with respect to Output
sdCY = sdC/sdY
sdIY = sdI/sdY
sdHY = sdH/sdY
sdWY = sdW/sdY
sdAY = sdA/sdY
```

Table for part b) of question 1

```
variable Autocorr St_Dev Corr_with_Y Rel_St_Dev
##
## 1
            Y
                   0.71
                           1.58
                                        1.00
                                                    1.00
## 2
            С
                   0.35
                           1.61
                                        0.56
                                                    1.02
## 3
            Ι
                   0.71
                           6.51
                                        0.85
                                                    4.13
            Η
                                                    1.27
## 4
                   0.59
                           2.00
                                        0.86
## 5
             W
                   0.16
                           1.03
                                       -0.02
                                                    0.65
                                                    0.66
## 6
             Α
                   0.37
                           1.04
                                        0.29
```

Part c) The main stylized facts that RBC model should aim to explain are

- 1) output and consumption have the same volatility.
- 2) Investment is at least 4 times as volatile as the output
- 3) consumption does not show one to one correlation with output signaling that agents do consumption smoothing
- 4) Wages are not at all correlated with output may be because of contracts as they are predetermined
- 5) working hours and investments are significantly correlated with output.