

# Advanced Macro 2 - Assignment 1

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## Preliminary

We hereby declare that the answers to the given assignment are entirely our own, resulting from our own work effort only. Our team members contributed to the answers of the assignment in the following proportions:\

- Unterweger Lucas: 50
- Oberbrinkmann Sophia: 50

## Question 1: Business cycles stylized facts (4 Points)

We'll start by setting up our coding environment by importing necessary packages. (Code not shown.)

Now, we can import and clean our data set.

```
ireland <- read_excel("data/Ireland_GDPData.xlsx", sheet = 4)
```

```
## New names:
```

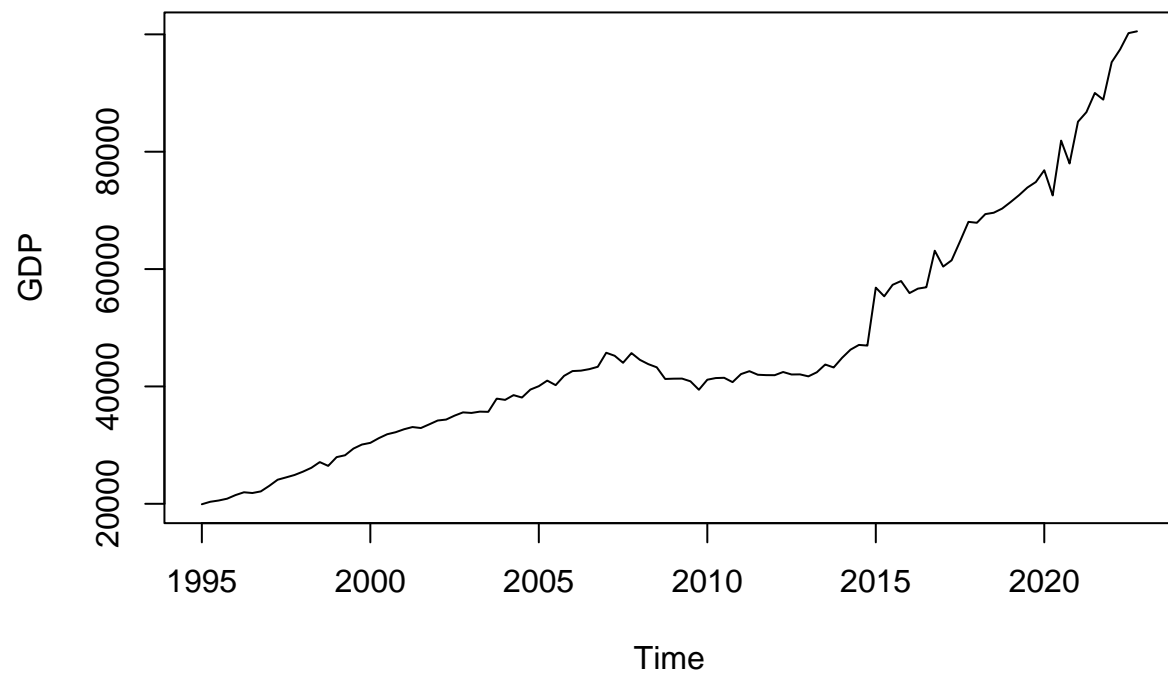
```
## * `` -> `...1`
```

```
colnames(ireland) <- c("t", "Y", "G", "C", "I")
head(ireland)
```

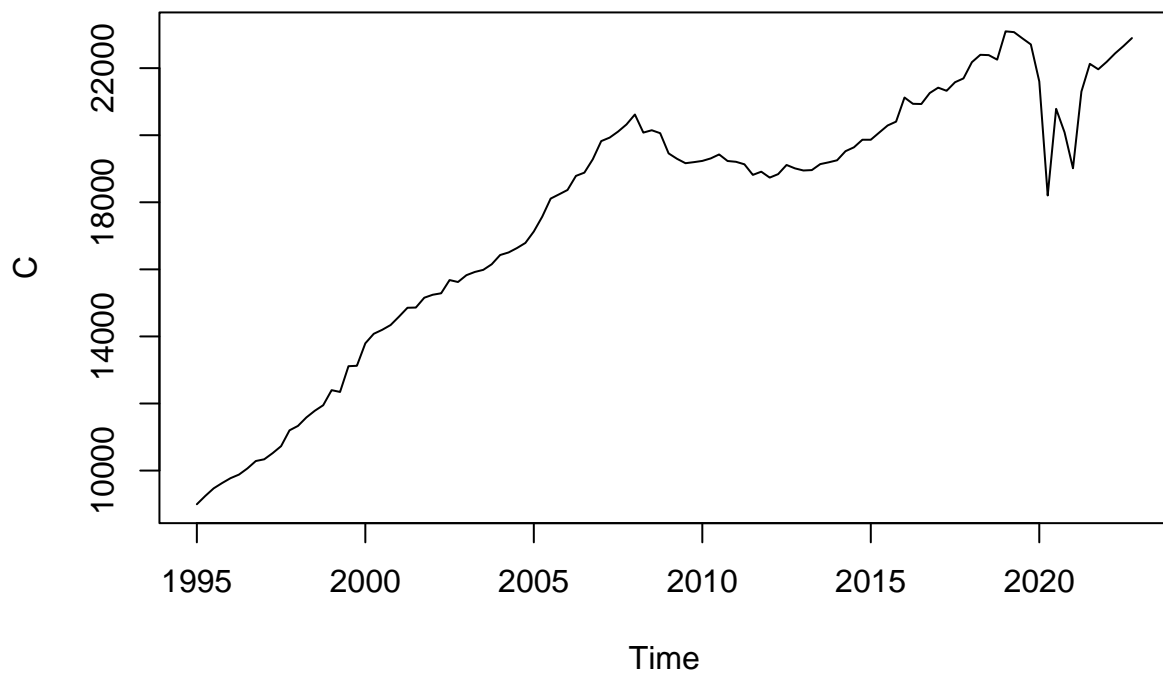
```
## # A tibble: 6 x 5
```

```
##   t           Y      G      C      I
##   <chr>      <dbl> <dbl> <dbl> <dbl>
## 1 1995-Q1 19924. 4407. 8996. 3573.
## 2 1995-Q2 20344. 4281. 9243. 3754.
## 3 1995-Q3 20559. 4288. 9470. 4123.
## 4 1995-Q4 20866. 4138. 9628. 4168.
## 5 1996-Q1 21485. 4561. 9773. 4158.
## 6 1996-Q2 21964. 4419. 9879. 4636.
```

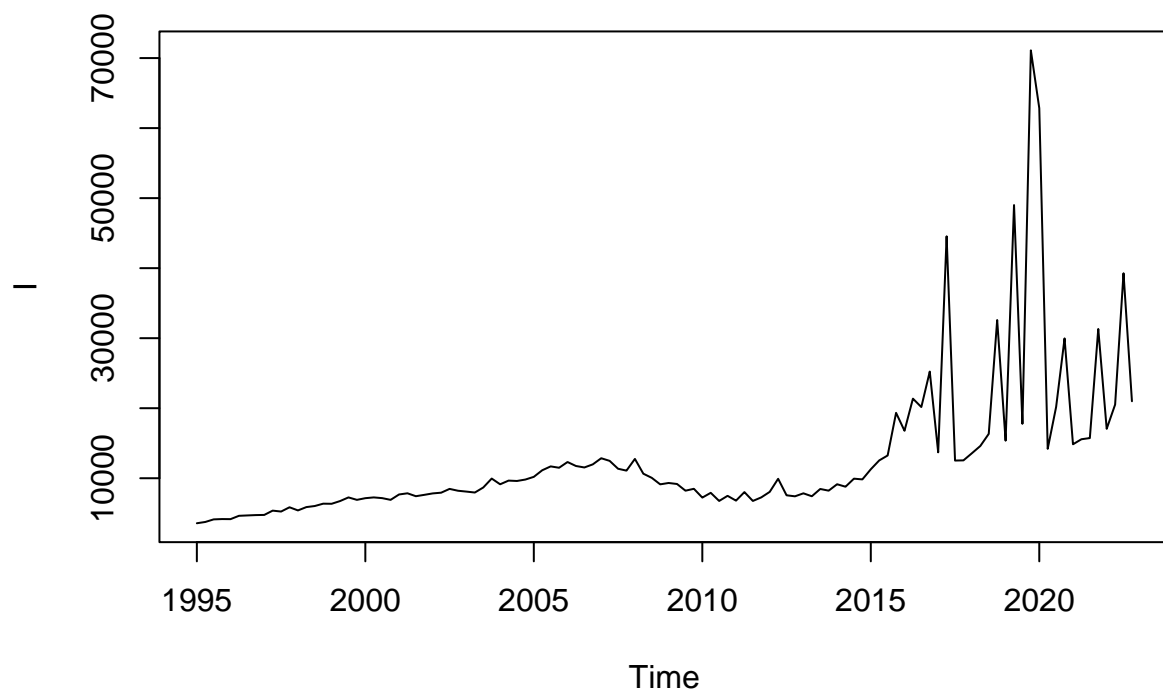
```
GDP <- ts(ireland$Y, start = 1995.0, frequency = 4)
C <- ts(ireland$C, start = 1995.0, frequency = 4)
I <- ts(ireland$I, start = 1995.0, frequency = 4)
G <- ts(ireland$G, start = 1995.0, frequency = 4)
plot(GDP)
```



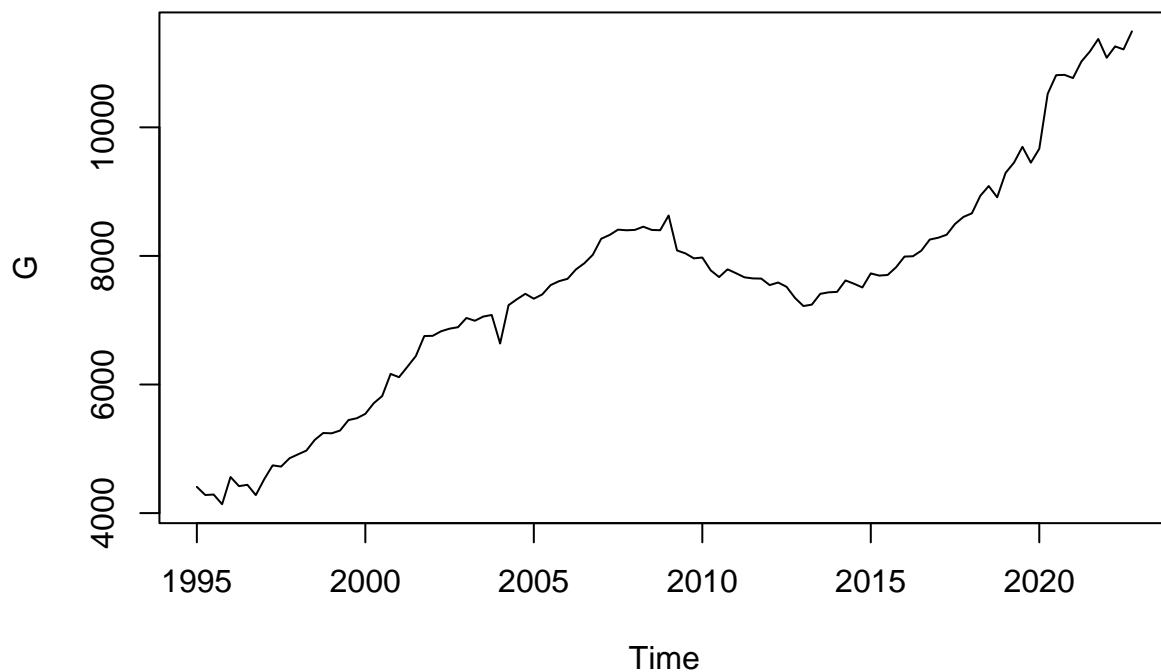
```
plot(C)
```



```
plot(I)
```



```
plot(G)
```



## Sample Statistics Let's compute the sample means:

```
CdivGDP <- C/GDP
IdivGDP <- I/GDP
GdivGDP <- G/GDP
```

```
mean(CdivGDP)
```

```
## [1] 0.4063485
```

```
mean(IdivGDP)
```

```
## [1] 0.2530171
```

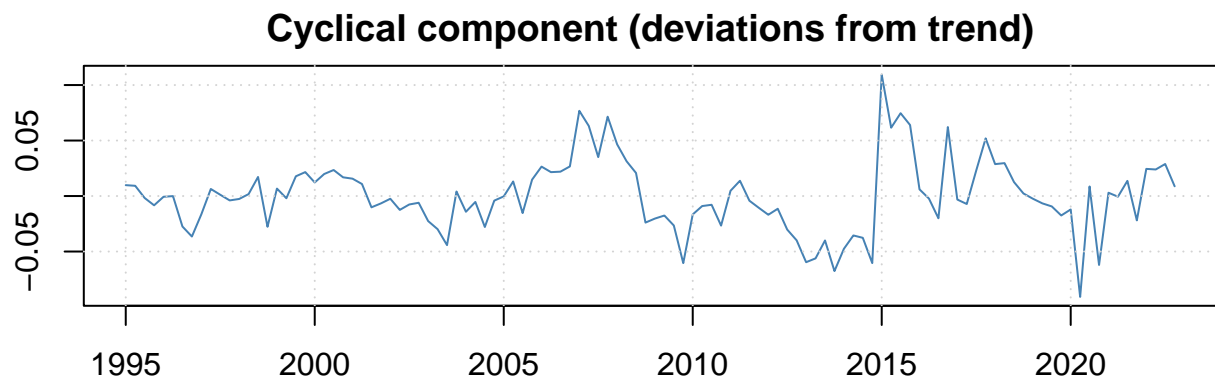
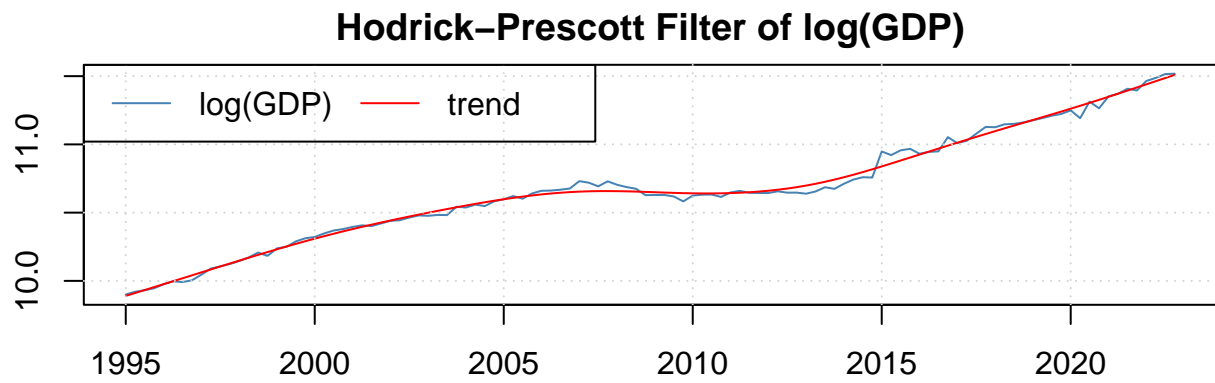
```
mean(GdivGDP)
```

```
## [1] 0.1719054
```

## Detrending

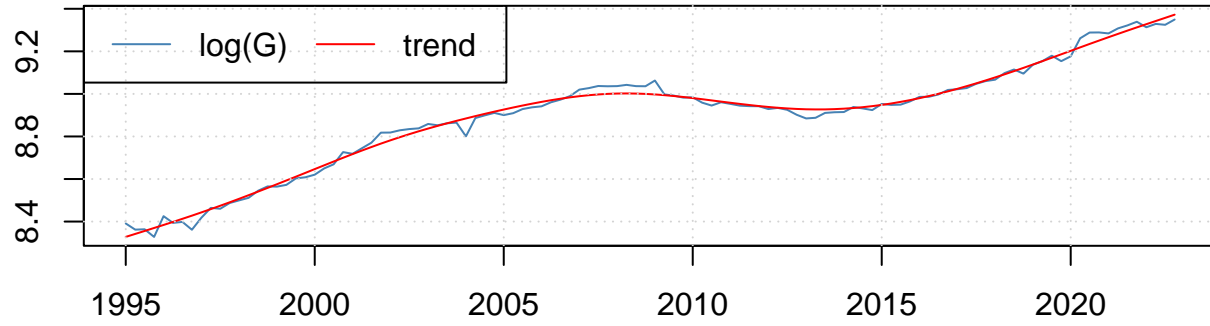
Let's apply the HP filter:

```
plot(hpfilter(log(GDP)))
```



```
plot(hpfilter(log(G)))
```

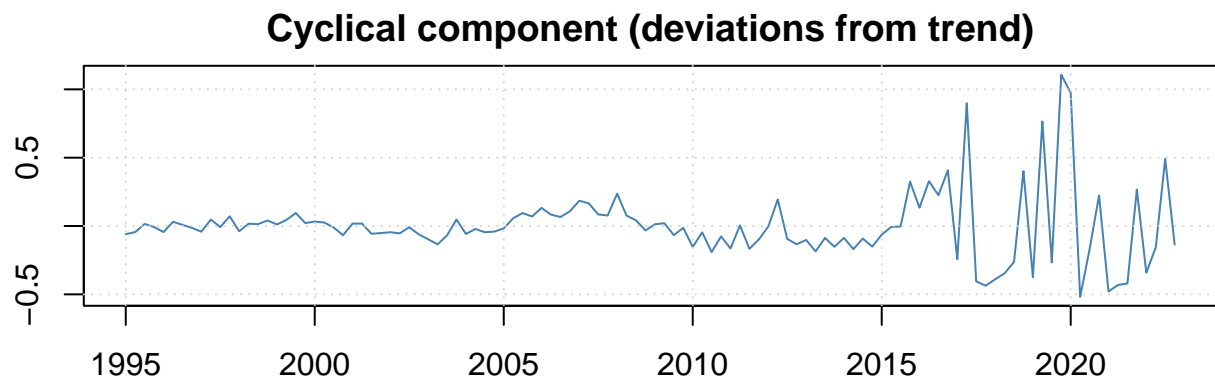
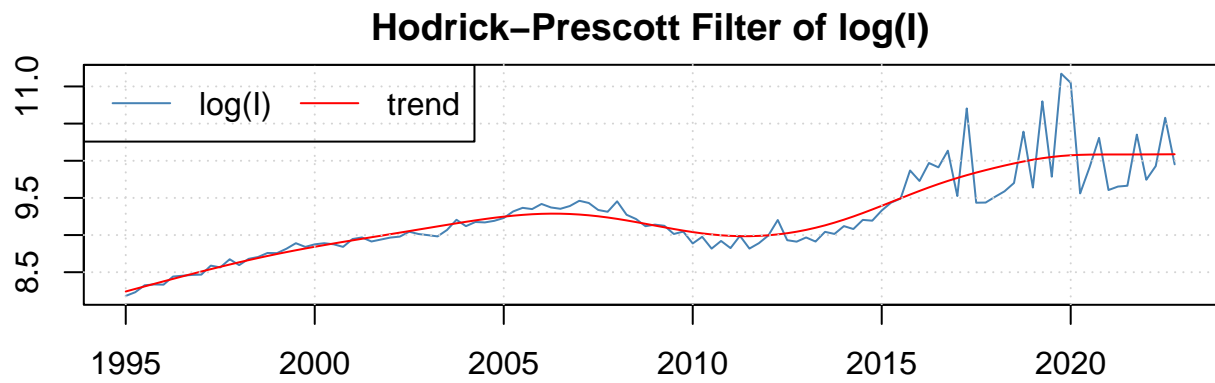
### Hodrick–Prescott Filter of $\log(G)$



### Cyclical component (deviations from trend)

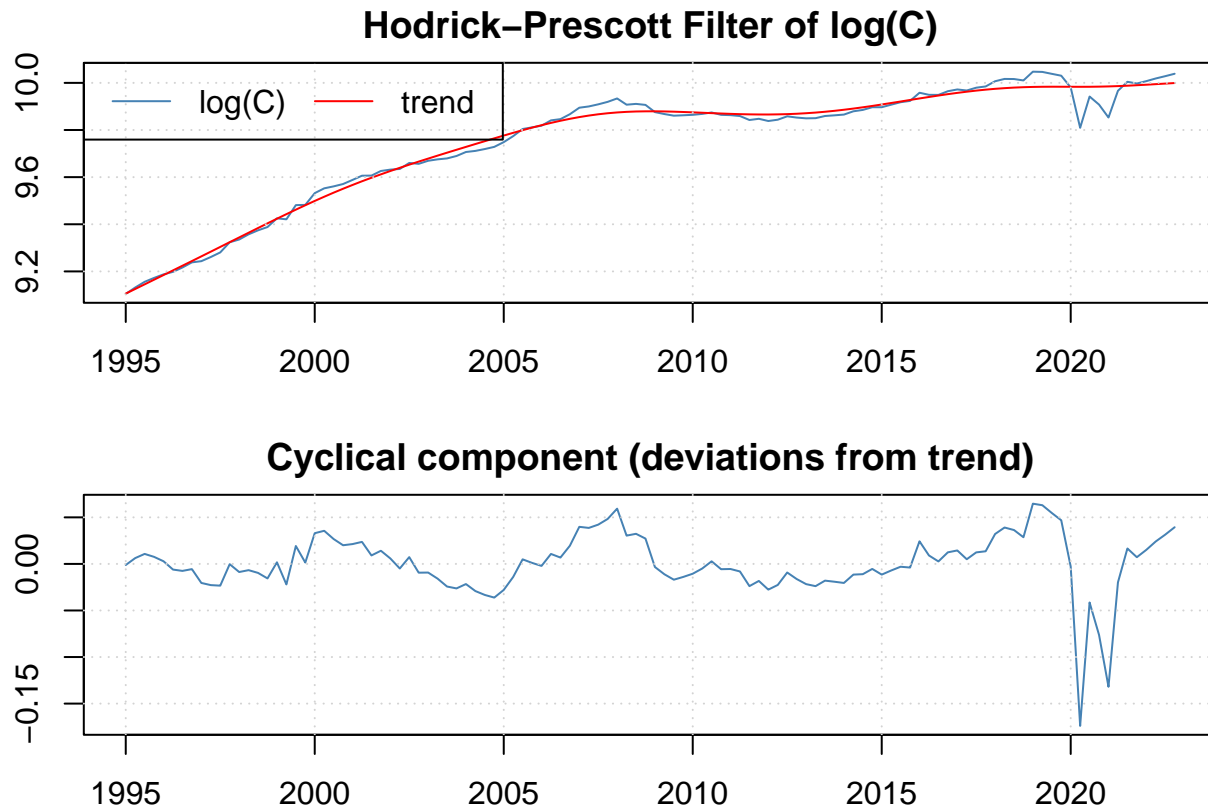


```
plot(hpfilter(log(I)))
```



```
plot(hpfilter(log(C)))
```





## Business Cycle Stylized Facts

```
trend_G <- hpfilter(log(G))[1]$cycle
trend_C <- hpfilter(log(C))[1]$cycle
trend_I <- hpfilter(log(I))[1]$cycle
trend_GDP <- hpfilter(log(GDP))[1]$cycle

output_table <- data.frame(
  names = c("GDP", "I", "C", "G"),
  standard_deviation = c(sd(trend_GDP), sd(trend_I), sd(trend_C), sd(trend_G)),
  relative_sd = c(sd(trend_GDP)/sd(trend_GDP), sd(trend_I)/sd(trend_GDP), sd(trend_C)/sd(trend_GDP), sd(trend_G)/sd(trend_GDP)),
  cont_output_corr = c(cor(trend_GDP, trend_GDP), cor(trend_I, trend_GDP), cor(trend_C, trend_GDP), cor(trend_G, trend_GDP))
)
output_table
```

##	names	standard_deviation	relative_sd	cont_output_corr
## 1	GDP	0.03230920	1.0000000	1.0000000
## 2	I	0.25468238	7.8826591	0.1112197
## 3	C	0.03193810	0.9885142	0.4768935
## 4	G	0.02470573	0.7646655	0.1549818

## Window 1

```
GDP_W1 <- window(GDP, end=2007.75)
I_W1 <- window(I, end=2007.75)
```

```

C_W1 <- window(C, end=2007.75)
G_W1 <- window(G, end=2007.75)

trend_G <- hpfiler(log(G_W1))[1]$cycle
trend_C <- hpfiler(log(C_W1))[1]$cycle
trend_I <- hpfiler(log(I_W1))[1]$cycle
trend_GDP <- hpfiler(log(GDP_W1))[1]$cycle

output_table_W1 <- data.frame(
  names = c("GDP", "I", "C", "G"),
  standard_deviation = c(sd(trend_GDP), sd(trend_I), sd(trend_C), sd(trend_G)),
  relative_sd = c(sd(trend_GDP)/sd(trend_GDP), sd(trend_I)/sd(trend_GDP), sd(trend_C)/sd(trend_GDP), sd(trend_G)/sd(trend_GDP)),
  cont_output_corr = c(cor(trend_GDP, trend_GDP), cor(trend_I, trend_GDP), cor(trend_C, trend_GDP), cor(trend_G, trend_GDP))
)

```

## Window 2

```

GDP_W2 <- window(GDP, start=2008.0)
I_W2 <- window(I, start=2008.0)
C_W2 <- window(C, start=2008.0)
G_W2 <- window(G, start=2008.0)

trend_G <- hpfiler(log(G_W2))[1]$cycle
trend_C <- hpfiler(log(C_W2))[1]$cycle
trend_I <- hpfiler(log(I_W2))[1]$cycle
trend_GDP <- hpfiler(log(GDP_W2))[1]$cycle

output_table_W2 <- data.frame(
  names = c("GDP", "I", "C", "G"),
  standard_deviation = c(sd(trend_GDP), sd(trend_I), sd(trend_C), sd(trend_G)),
  relative_sd = c(sd(trend_GDP)/sd(trend_GDP), sd(trend_I)/sd(trend_GDP), sd(trend_C)/sd(trend_GDP), sd(trend_G)/sd(trend_GDP)),
  cont_output_corr = c(cor(trend_GDP, trend_GDP), cor(trend_I, trend_GDP), cor(trend_C, trend_GDP), cor(trend_G, trend_GDP))
)

```

## Output

```
output_table
```

```

##  names standard_deviation relative_sd cont_output_corr
## 1  GDP          0.03230920    1.0000000         1.0000000
## 2   I           0.25468238    7.8826591         0.1112197
## 3   C           0.03193810    0.9885142         0.4768935
## 4   G           0.02470573    0.7646655         0.1549818

```

```
output_table_W1
```

```

##  names standard_deviation relative_sd cont_output_corr
## 1  GDP          0.01582877    1.0000000         1.0000000
## 2   I           0.05050510    3.190716         0.4005525
## 3   C           0.01591679    1.005561         0.4630622
## 4   G           0.02558119    1.616120         0.1368501

```

```
output_table_W2
```

```

##  names standard_deviation relative_sd cont_output_corr

```

## 1	GDP	0.03803500	1.0000000	1.000000000
## 2	I	0.34357331	9.0330832	0.054996823
## 3	C	0.03783447	0.9947279	0.392198387
## 4	G	0.01911173	0.5024774	0.006569063

## Question 2: A real business cycle model with energy price shocks (5 points)

Expected discounted utility of the representative household:

$$U = E_0 \left[ \sum_{t=0}^{\infty} \beta^t \frac{C_t^{1-\sigma}}{1-\sigma} + \theta \log(1 - N_t) \right] \quad (1)$$

subject to

$$C_t + K_{t+1} + P_t E N_t \leq A_t K_t^\alpha N_t^\gamma E N_t^{1-\alpha-\gamma} + (1-\delta)K_t \quad (2)$$

### (a) Social Planner's Intertemporal Optimization Problem

We start by setting up our Lagrangian system where we maximize (1) which the constraint (2). The Lagrangian is then given by

$$L_t = E_0 \left[ \sum_{t=0}^{\infty} \beta^t \left\{ \frac{C_t^{1-\sigma} - 1}{1-\sigma} + \theta \log(1 - N_t) + \lambda_t \left( A_t K_t^\alpha N_t^\gamma E N_t^{1-\alpha-\gamma} + (1-\delta)K_t - C_t - K_{t+1} - P_t E N_t \right) \right\} \right]$$

We can now use *first order optimality conditions* and optimize w.r.t the *current consumption*  $C_t$ , *tomorrows capital stock*  $K_{t+1}$ , *hours worked*  $N_t$ , *energy consumption*  $E N_t$  and the *Lagrange multiplier*  $\lambda_t$ . This results in the following five equations:

$$\frac{\partial L}{\partial C_t} = E_0 \beta^t \left\{ (1-\sigma) \frac{C_t^{-\sigma}}{1-\sigma} - \lambda_t \right\} \stackrel{!}{=} 0 \quad (3)$$

$$\iff C_t^{-\sigma} = \lambda_t \quad (4)$$

**Question 3: Understanding impulse responses and model simulation  
(total of 8 points)**

**(a) Steady State**

The following parameters are given:  $\beta = 0.99$ ,  $\sigma = 1$ ,  $\gamma = 0.65$ ,  $\theta = 3.48$ ,  $\alpha = 0.3$ ,  $\delta = 0.025$ ,  $\bar{A} = 1$ ,  $\bar{P} = 1$ ,  
 $\rho_A = 0.95$ ,  $\sigma_A = 0.007$ ,  $\rho_P = 0.5$ ,  $\sigma_P = 0.00001$