# Homework 02: Exploring Data

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## Question 1

```
library(dplyr) # install.packages("dplyr")
## Attaching package: 'dplyr'
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
library(ggplot2) # install.packages("ggplot2")
library(scales) # install.packages("scales")
library(tidyr) # install.packages("tidyr")
library(haven) # install.packages("haven")
library(fredr) # Loading the database from https://cran.r-project.org/web/packages/fredr/vignettes/fred
#install.packages("fredr")
fredr_set_key("bdd1cf84197911e0b8c4346fdd943c50")
# usethis::edit_r_environ()
# FRED_API_KEY=bdd1cf84197911e0b8c4346fdd943c50
# https://fred.stlouisfed.org/series/RKNANPINA666NRUG
capital_stock <- fredr(</pre>
  series_id = "RKNANPINA666NRUG",
  observation_start = as.Date("1970-01-01"),
  observation_end = as.Date("2019-12-31")
head(capital_stock)
## # A tibble: 6 x 5
##
     date
               series_id
                                     value realtime_start realtime_end
                <chr>>
                                     <dbl> <date>
                                                          <date>
     <date>
## 1 1970-01-01 RKNANPINA666NRUG 2007558. 2025-03-06
                                                          2025-03-06
## 2 1971-01-01 RKNANPINA666NRUG 2125584. 2025-03-06
                                                          2025-03-06
## 3 1972-01-01 RKNANPINA666NRUG 2246936. 2025-03-06
                                                          2025-03-06
## 4 1973-01-01 RKNANPINA666NRUG 2356271. 2025-03-06
                                                          2025-03-06
## 5 1974-01-01 RKNANPINA666NRUG 2469048. 2025-03-06
                                                          2025-03-06
## 6 1975-01-01 RKNANPINA666NRUG 2572254. 2025-03-06
                                                          2025-03-06
# https://fred.stlouisfed.org/series/EMPENGINA148NRUG
persons_engaged <- fredr(</pre>
```

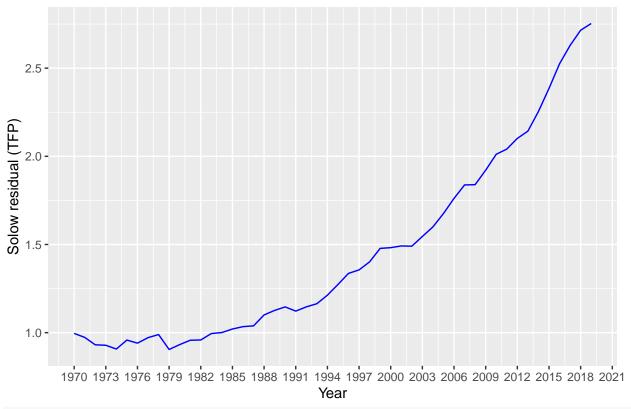
```
series_id = "EMPENGINA148NRUG",
  observation_start = as.Date("1970-01-01"),
  observation end = as.Date("2019-12-31")
)
#https://fred.stlouisfed.org/series/AVHWPEINA065NRUG
avg_hours_worked <- fredr(</pre>
  series_id = "AVHWPEINAO65NRUG",
  observation start = as.Date("1970-01-01"),
  observation_end = as.Date("2019-12-31")
head(persons_engaged)
## # A tibble: 6 x 5
##
                                 value realtime_start realtime_end
     date
           series_id
     <date>
                <chr>
                                 <dbl> <date>
                                                      <date>
## 1 1970-01-01 EMPENGINA148NRUG 195. 2025-03-06
                                                      2025-03-06
## 2 1971-01-01 EMPENGINA148NRUG 201. 2025-03-06
                                                      2025-03-06
## 3 1972-01-01 EMPENGINA148NRUG 208. 2025-03-06
                                                      2025-03-06
## 4 1973-01-01 EMPENGINA148NRUG 214. 2025-03-06
                                                      2025-03-06
## 5 1974-01-01 EMPENGINA148NRUG 221. 2025-03-06
                                                      2025-03-06
## 6 1975-01-01 EMPENGINA148NRUG 227. 2025-03-06
                                                      2025-03-06
head(avg_hours_worked)
## # A tibble: 6 x 5
##
    date
               series_id
                                 value realtime_start realtime_end
##
     <date>
                <chr>
                                 <dbl> <date>
                                                      <date>
## 1 1970-01-01 AVHWPEINA065NRUG 2077. 2025-03-06
                                                      2025-03-06
## 2 1971-01-01 AVHWPEINA065NRUG 2077. 2025-03-06
                                                      2025-03-06
## 3 1972-01-01 AVHWPEINA065NRUG 2077. 2025-03-06
                                                      2025-03-06
## 4 1973-01-01 AVHWPEINA065NRUG 2075. 2025-03-06
                                                      2025-03-06
## 5 1974-01-01 AVHWPEINA065NRUG 2072. 2025-03-06
                                                      2025-03-06
## 6 1975-01-01 AVHWPEINA065NRUG 2070. 2025-03-06
                                                      2025-03-06
# https://fred.stlouisfed.org/series/RGDPNAINA666NRUG
real_gdp <- fredr(</pre>
  series_id = "RGDPNAINA666NRUG",
  observation_start = as.Date("1970-01-01"),
  observation_end = as.Date("2019-12-31")
head(real_gdp)
## # A tibble: 6 x 5
##
     date
                series id
                                   value realtime_start realtime_end
                                   <dbl> <date>
## 1 1970-01-01 RGDPNAINA666NRUG 684889. 2025-03-06
                                                        2025-03-06
## 2 1971-01-01 RGDPNAINA666NRUG 696141. 2025-03-06
                                                        2025-03-06
## 3 1972-01-01 RGDPNAINA666NRUG 692289. 2025-03-06
                                                        2025-03-06
## 4 1973-01-01 RGDPNAINA666NRUG 715104. 2025-03-06
                                                        2025-03-06
## 5 1974-01-01 RGDPNAINA666NRUG 723580. 2025-03-06
                                                        2025-03-06
## 6 1975-01-01 RGDPNAINA666NRUG 789787. 2025-03-06
                                                        2025-03-06
```

```
# Rename 'value' columns for clarity
real_gdp <- real_gdp %>% rename(gdp = value)
capital_stock <- capital_stock %>% rename(capital = value)
persons_engaged <- persons_engaged %>% rename(persons = value)
avg_hours_worked <- avg_hours_worked %>% rename(hours = value)
# Merge datasets
merged_data_india <- real_gdp %>%
  inner_join(capital_stock, by = "date") %>%
  inner_join(persons_engaged, by = "date") %>%
  inner_join(avg_hours_worked, by = "date")
merged_data_india <- merged_data_india %>%
  mutate(labor = persons * hours)
alpha <- 0.33
merged_data_india <- merged_data_india %>%
  mutate(
    tfp = gdp / (capital^alpha * labor^(1 - alpha))
```

Assuming a Cobb-Douglas production function with  $\alpha = 0.33$ , find and plot the Solow residual of the Indian economy. Are there any patterns that catch your attention in that plot?

```
ggplot(merged_data_india, aes(x = date, y = tfp)) +
  geom_line(color = "blue") +
  labs(
    title = "Solow Residual (Total Factor Productivity) for India (1970-2019)",
    x = "Year",
    y = "Solow residual (TFP)"
  ) +
  scale_x_date(date_breaks = "3 years", date_labels = "%Y")
```

## Solow Residual (Total Factor Productivity) for India (1970–2019)



#### # theme\_minimal()

- 1. Gradual increase before 1990: the slope rate is low and the growth is slow.
- 2. Fast sustained increase after 1991: might be correlated to the economic reform leading to an increase technology adoption, innovation, which increase the TFP (Singer, 1978).
- 3. Some fluctuations around 2007-2010: the stagnation of growth of TFP is likely due to the global financial crisis.

# Question 02

```
Maddison_data <- read_dta("maddison2023_web.dta")

# MPD version 2023: Bolt, Jutta and Jan Luiten van Zanden (2024). Maddison style estimates of the evolu
```

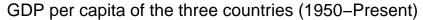
### Geary-Khamis dollar definition

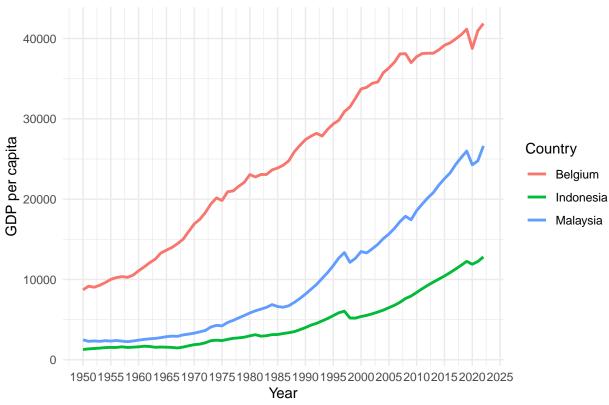
Geary-Khamis dollar is a type of dollar that has the same amount of PPP as US Dollars in the US at a given point of time (Bolt, 2024). For example, people use the 1990 and 2016 USD as the standard of reference. The PPP as the basis of comparison is key here because it standardizes the currency by examining the same bundle of goods in different countries.

### Plot GDP per capita

Plot GDP per capita for Belgium, Malaysia and Indonesia since 1950 (use just one graph for the three countries to be able to compare them).

```
head(Maddison_data)
## # A tibble: 6 x 6
    countrycode country
                             region
                                                        year gdppc
                                                                     pop
     <chr>
                 <chr>
                             <chr>
                                                       <dbl> <dbl> <dbl>
                 Afghanistan South and South East Asia
## 1 AFG
                                                           1
                                                                NA
                                                                      NA
## 2 AFG
                 Afghanistan South and South East Asia
                                                         730
                                                                NA
                                                                      NA
## 3 AFG
                 Afghanistan South and South East Asia 1000
                                                                NA
                                                                      NA
## 4 AFG
                 Afghanistan South and South East Asia 1090
                                                                      NA
                                                                NA
## 5 AFG
                 Afghanistan South and South East Asia 1150
                                                                NA
                                                                      NA
## 6 AFG
                 Afghanistan South and South East Asia 1252
                                                                      NA
# Filter the data for Belgium, Malaysia, and Indonesia since 1950
filtered_Maddison_data <- Maddison_data %>%
 filter(country %in% c("Belgium", "Malaysia", "Indonesia"), year >= 1950)
# Plot GDP per capita for the three countries
ggplot(filtered_Maddison_data, aes(x = year, y = gdppc, color = country)) +
  geom_line(size = 1) +
 labs(title = "GDP per capita of the three countries (1950-Present)",
       x = "Year",
      y = "GDP per capita",
      color = "Country") +
  scale_x_continuous(breaks = seq(1950, 2025, by = 5)) +
 theme_minimal()
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```





### Catch-up growth

Assume that GDP per capita in all three countries will continue to grow at the average growth rate they have experienced during the last 10 years of available data. How long will it take for Malaysia and Indonesia to catch up with Belgium in terms of GDP per capita? Detail the steps you take and discuss your results.

```
# Filter the data for the last 10 years
last_10_years <- Maddison_data %>%
  filter(year >= 2022-9) %>%
  filter(country %in% c("Belgium", "Malaysia", "Indonesia"))
# Compute the average annual growth rate (avg_growth_rate) for each country
growth_rates <- last_10_years %>%
  group_by(country) %>%
  summarize(
   gdp_initial = first(gdppc),
   gdp_final = last(gdppc),
    avg_growth_rate = (gdp_final / gdp_initial)^(1/10) - 1
  )
# Get the most recent GDP per capita for each country
latest_gdp <- Maddison_data %>%
  filter(year == max(year), country %in% c("Belgium", "Malaysia", "Indonesia")) %>%
  select(country, gdppc)
# Merge growth rates with latest GDP data
growth_rates <- left_join(growth_rates, latest_gdp, by = "country")
```

```
# Belgium's GDP per capita (target)
gdp_belgium <- growth_rates %>% filter(country == "Belgium") %>% pull(gdppc)
# Function:
# Input: gdp_start, gdp_target, gdp_growth_rate
# Output: catchup time
catchup_time <- function(gdp_start, growth_rate, gdp_target) {</pre>
  log(gdp target / gdp start) / log(1 + growth rate)
}
# Compute the catch-up time required
func_catchup_years <- growth_rates %>%
  filter(country %in% c("Malaysia", "Indonesia")) %>%
  mutate(years_to_catch_up = catchup_time(gdppc, avg_growth_rate, gdp_belgium))
print(func_catchup_years)
## # A tibble: 2 x 6
##
               gdp_initial gdp_final avg_growth_rate gdppc years_to_catch_up
     country
     <chr>>
                     <dbl>
                               <dbl>
                                                <dbl> <dbl>
                                                                          <dbl>
## 1 Indonesia
                     9676.
                              12802.
                                               0.0284 12802.
                                                                          42.3
## 2 Malaysia
                    20819.
                              26629.
                                               0.0249 26629.
                                                                          18.4
```

Therefore, although the year-on-year growth rates for Indonesia and Malasia are similar, because Indonesia's initial GDP per capita was so low at \$9676 compared to \$20819 in Malaysia, the catch-up growth for Indonesia (42.34 years) will be much longer than that of Malaysia (18.39 years).

#### References

Bolt, J. (2024). Maddison-style estimates of the evolution of the world economy: A new 2023 update. Journal of Economic Surveys, 39(2), 631-671. https://doi.org/10.1111/joes.12618

Singer, H. W. (1978). Foreign trade regimes and economic development: India: Jagdish N. bhagwati and T.N. srinivasan, (national bureau of economic research, new york, 1975-distributed by columbia university press) pp. xxiv + 261, \$15.00. Elsevier B.V. https://doi.org/10.1016/0022-1996(78)90011-9