Eurostat EDA

Sindre H. Øveraas, Alen Colakovic & Sebastian M. Fløysand

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Meta-data description

Nama 10r

Regional accounts are based on the same definitions and concepts as national accounts, this is because they are a specification of the national accounts.

Gross domestic product (GDP) is the standard measure of the value added created through the production of goods and services in a country during a certain period. As such, it also measures the income earned from that production, or the total amount spent on final goods and services (less imports). While GDP is the single most important indicator to capture economic activity, it falls short of providing a suitable measure of people's material well-being for which alternative indicators may be more appropriate.

@GDPSpendingGross

We have two ways of calculating this within regional accounts:

The first one is the output approach

Where the GDP is the sum of gross value added of the various institutional sectors or the various industries plus taxes and less subsidies on products. It is also the balancing item in the total economy production account."

2. Income approach

"GDP is the sum of uses in the total economy generation of income account: compensation of employees plus gross operating surplus and mixed income plus taxes on products less subsidies plus consumption of fixed capital.

Contrary to national accounts GDP is not compoled from the expenditure side in regional accounts due to data limitations on the inter-regional flows of goods and services.

The different measures for the regional GDP are absolute figures in ϵ and Purchasing Power Standards (PPS), figures per inhabitant and relative data compared to the EU Member States average." "Regional Economic Accounts (Reg_eco10)" (n.d.)

Sub-national GDP

```
Country codes: BE - Belgium, BG - Bulgaria, HR - Croatia, IT - Italy, AT - Austria, SE - Sweden, RS - Serbia
```

These are the RStudio packages we install and load to do our calculations.

```
library(tidyverse)
library(vtable)
library(dineq)
library(dplyr)
```

We downloaded from Eurostat our subset of countries sub-regional GDP (nama_10r_3gdp__custom_356493 and population (demo_r_pjanaggr3__custom_3579517_linear.csv) for the years 2000 - 2020. We then calculated the GDP per capita, and named the new data set "GDP Per Capita".

```
library(readr)
GDP <- read.csv('nama_10r_3gdp__custom_3564935_linear.csv')
Population <- read_csv("demo_r_pjanaggr3__custom_3579517_linear.csv")

gdpdata <- GDP %>%
    rename(Year = TIME_PERIOD, GDP = OBS_VALUE, Region = geo)

populationdata <- Population %>%
    rename(Year = TIME_PERIOD, Population = OBS_VALUE, Region = geo)

GDP_Per_Capita <- gdpdata %>%
    left_join(populationdata, by=c("Region", "Year")) %>%
    select(Region, Year, GDP, Population) %>%
    mutate(
    GDP_capita = (GDP * 1000000)/Population)
```

To report our descriptive statistics on GDP per capita, we used the summary command. This gave us the following descriptive statistics on the variables GDP, population and GDP capita.

```
GDP
                    Population
                                     GDP_capita
          74.55
                  Min. : 20320
                                   Min. : 1087
1st Qu.: 1738.28
                  1st Qu.: 164518
                                   1st Qu.:17180
Median: 5614.05 Median: 273920
                                   Median :25185
     : 10238.24
                        : 406217
                                         :24191
Mean
                  Mean
                                   Mean
3rd Qu.: 10640.23
                  3rd Qu.: 429030
                                    3rd Qu.:31351
Max. :181212.88
                  Max. :4355725
                                   Max. :72062
                  NA's
                         :771
                                    NA's
                                          :771
```

To calculate the population watertight GDP Ginie coefficient, we used the following command.

```
gini.wtd(GDP_Per_Capita$GDP_capita, weights = GDP_Per_Capita$Population)
```

[1] 0.2603924

To calculate the population watertight GDP Ginie coefficients for the European NUTS2 level, we used the following commands.

```
GDP_Per_Capita <- GDP_Per_Capita %>%
  mutate(NUTS2 = substr(GDP_Per_Capita$Region,1,4))

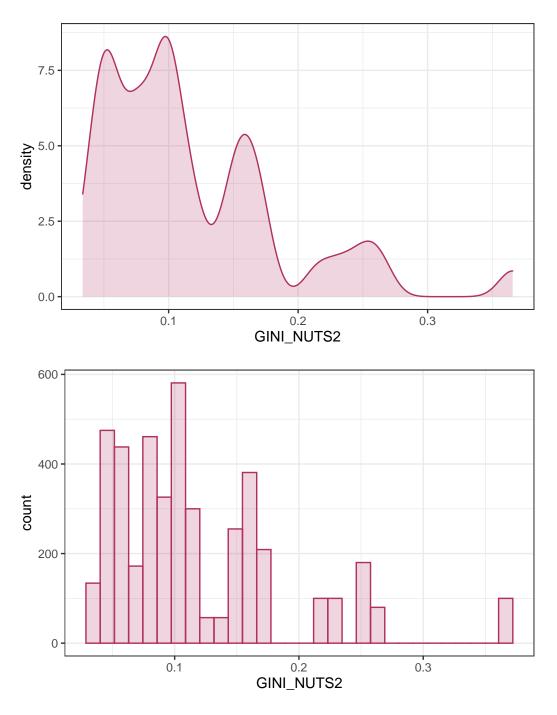
GDP_Per_Capita <- GDP_Per_Capita %>%
  mutate(NUTS = substr(GDP_Per_Capita$Region,1,2))

GDP_Per_Capita %<>%
  group_by(NUTS2) %>%
  na.exclude(GDP_Per_Capita) %>%
  mutate(GINI_NUTS2 = gini.wtd(GDP_capita, weights = Population)) %>%
  ungroup()
```

To summarize the data of the Ginie coefficients, we once more used the summarize com-

```
GINI_NUTS2
Min. :0.03367
1st Qu.:0.07065
Median :0.09839
Mean :0.11800
3rd Qu.:0.15525
Max. :0.36569
```

For visualization of our data (the Ginie coefficients for the European NUTS2 level of our selected countries), we produced a density plot and histogram, by the use of the ggplot2 package.



Looking at the plot above there is one outlier up against 0.4 with around 100 observations. The same result also seem to occur in the density plot.

 $[&]quot;Regional Economic Accounts (Reg_eco10)." \ n.d. \ https://ec.europa.eu/eurostat/cache/metadata/en/reg_eco10)."$