

Gini Coefficients and GDP Growth

Setup

If you have installed the package “tidyverse”, you do not need to run the next code chunk.

```
install.packages("tidyverse")
```

Although “readxl” and “broom” are part of the package “tidyverse”, you need to load separately as they are not core and not loaded with ‘tidyverse’.

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.3      v readr      2.1.4
## v forcats    1.0.0      v stringr   1.5.0
## v ggplot2    3.4.4      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.0
## v purrr      1.0.2
```

```
## -- Conflicts ----- tidyverse_conflicts() --
```

```
## x dplyr::filter() masks stats::filter()
```

```
## x dplyr::lag() masks stats::lag()
```

```
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(readxl) # a package for Excel files, a part of tidyverse but not core
```

```
library(broom) # a package for modeling, a part of tidyverse but not core
```

```
library(WDI)
```

Importing

For Excel, you need to specify which sheet you want to import by name as below or the number. You also need to skip several rows when blank rows are included.

```
df_gini_gdpgrowth_orig <- read_excel("data/P_Data_Extract_From_World_Development_Indicators 20 years db",
  col_types = c("text", "text", "text",
    "text", "numeric", "numeric", "numeric",
    "numeric", "numeric", "numeric",
    "numeric", "numeric", "numeric",
    "numeric", "numeric", "numeric",
    "numeric", "numeric", "numeric",
    "numeric", "numeric"), na = "\t ..", n_max = 532)
```

Check the data frames.

```
df_gini_gdpgrowth_orig |> head()
```

```
## # A tibble: 6 x 24
```

```
##   `Country Name` `Country Code` `Series Name`      `Series Code` `2003 [YR2003]`
```

```
##   <chr>          <chr>         <chr>          <chr>          <dbl>
```

```
## 1 Afghanistan  AFG           Gini index     SI.POV.GINI     NA
```

```
## 2 Afghanistan AFG GDP growth (annua~ NY.GDP.MKTP.~ 8.83
## 3 Albania ALB Gini index SI.POV.GINI NA
## 4 Albania ALB GDP growth (annua~ NY.GDP.MKTP.~ 5.53
## 5 Algeria DZA Gini index SI.POV.GINI NA
## 6 Algeria DZA GDP growth (annua~ NY.GDP.MKTP.~ 7.20
## # i 19 more variables: `2004 [YR2004]` <dbl>, `2005 [YR2005]` <dbl>,
## # `2006 [YR2006]` <dbl>, `2007 [YR2007]` <dbl>, `2008 [YR2008]` <dbl>,
## # `2009 [YR2009]` <dbl>, `2010 [YR2010]` <dbl>, `2011 [YR2011]` <dbl>,
## # `2012 [YR2012]` <dbl>, `2013 [YR2013]` <dbl>, `2014 [YR2014]` <dbl>,
## # `2015 [YR2015]` <dbl>, `2016 [YR2016]` <dbl>, `2017 [YR2017]` <dbl>,
## # `2018 [YR2018]` <dbl>, `2019 [YR2019]` <dbl>, `2020 [YR2020]` <dbl>,
## # `2021 [YR2021]` <dbl>, `2022 [YR2022]` <dbl>

df_gini_gdpgrowth2 <- WDI(indicator = c(gini = "SI.POV.GINI", gdpgrowth = "NY.GDP.MKTP.KD.ZG"))
df_gini_gdpgrowth2 |> head()

## country iso2c iso3c year gini gdpgrowth
## 1 Afghanistan AF AFG 1960 NA NA
## 2 Afghanistan AF AFG 1961 NA NA
## 3 Afghanistan AF AFG 1962 NA NA
## 4 Afghanistan AF AFG 1963 NA NA
## 5 Afghanistan AF AFG 1964 NA NA
## 6 Afghanistan AF AFG 1965 NA NA
```

Tidying

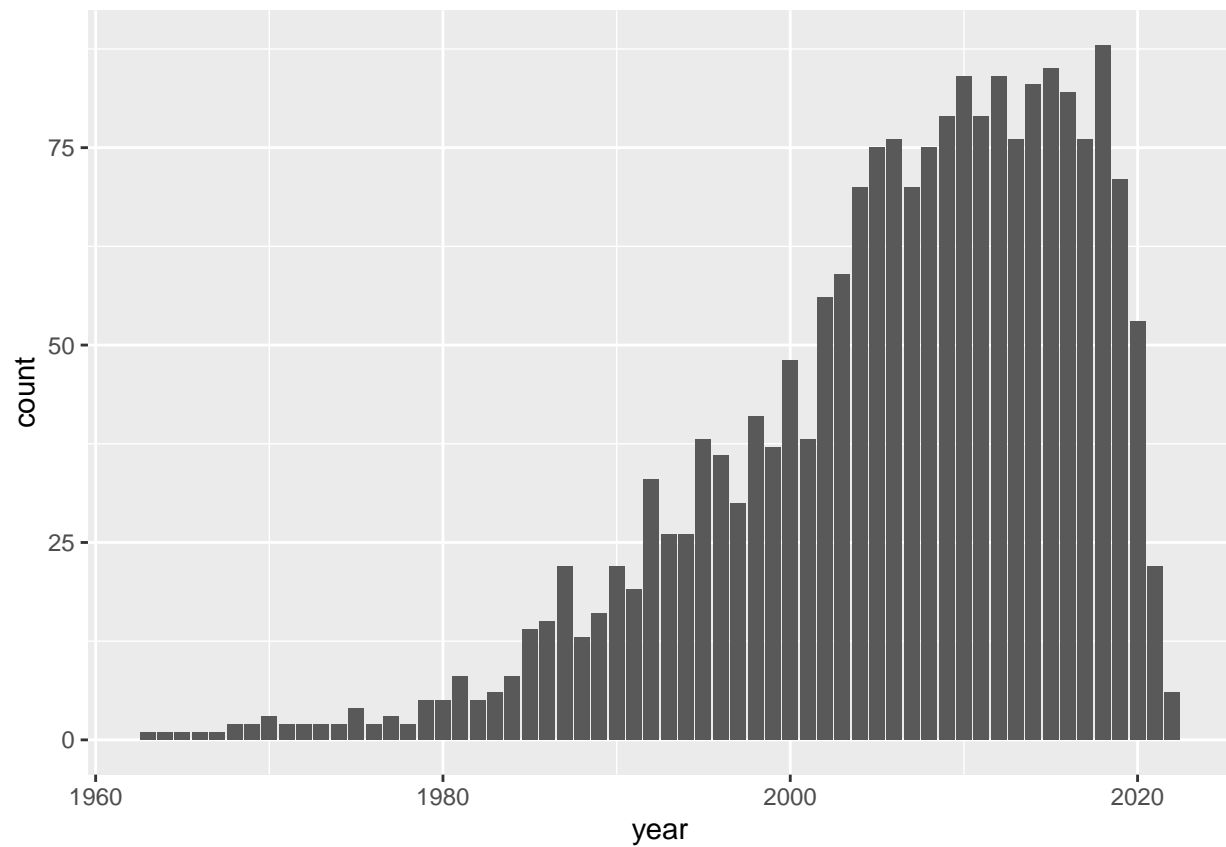
After tidying the original data, we obtained about the same as the one obtained above using WDI package. So I recommend you to get the data using WDI.

```
df_gini_gdpgrowth <- df_gini_gdpgrowth_orig |> pivot_longer(5:24, names_to = "year", values_to = "value")
mutate(year = as.numeric(str_sub(year, 1, 4))) |> select(-4) |>
pivot_wider(names_from = `Series Name`, values_from = value)
df_gini_gdpgrowth |> head()
```

```
## # A tibble: 6 x 5
##   `Country Name` `Country Code` year `Gini index` `GDP growth (annual %)`
##   <chr>          <chr>      <dbl>      <dbl>          <dbl>
## 1 Afghanistan AFG          2003         NA           8.83
## 2 Afghanistan AFG          2004         NA           1.41
## 3 Afghanistan AFG          2005         NA          11.2
## 4 Afghanistan AFG          2006         NA           5.36
## 5 Afghanistan AFG          2007         NA          13.8
## 6 Afghanistan AFG          2008         NA           3.92
```

Number of available data

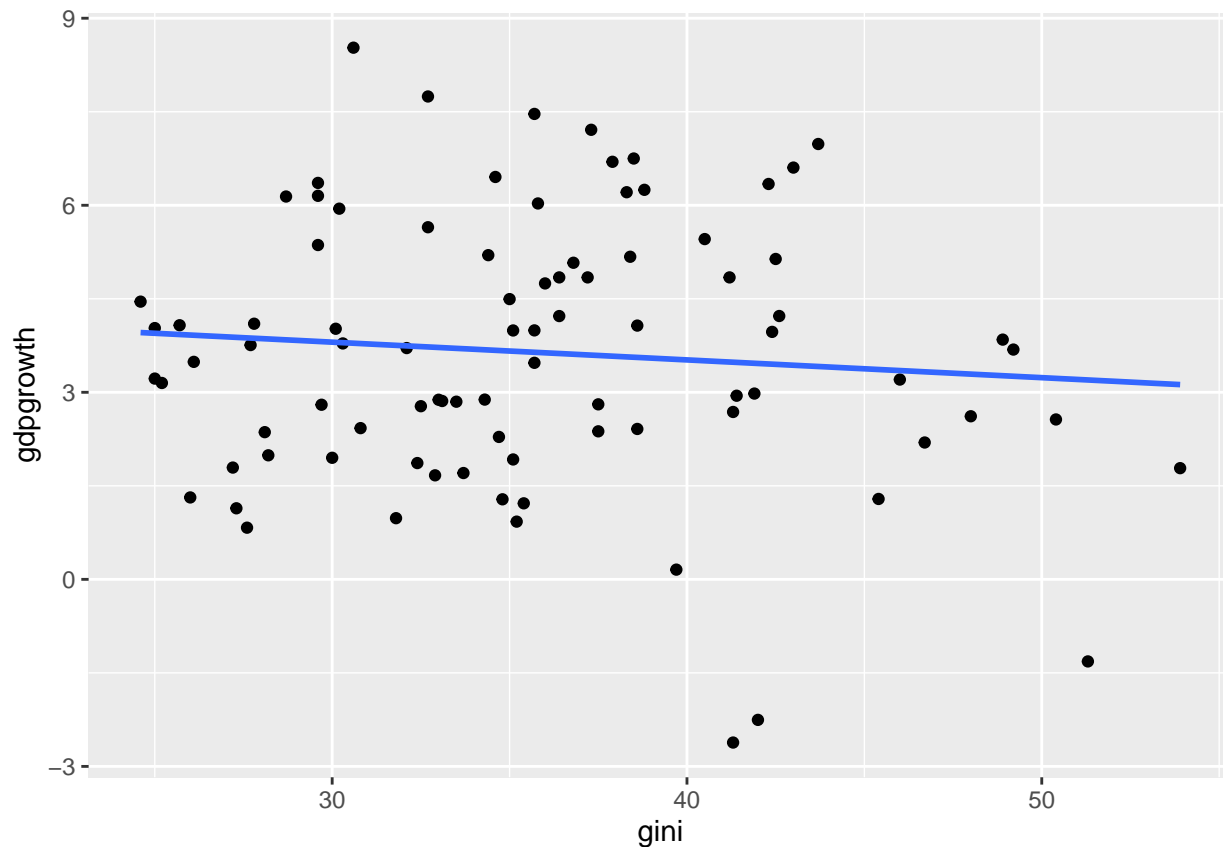
```
df_gini_gdpgrowth2 |> drop_na(gini, gdpgrowth) |>
ggplot(aes(year)) + geom_bar()
```



Scatter Plots

It is better to fix a year.

```
YEAR <- 2018
df_gini_gdpgrowth2 |> filter (year == YEAR) |> drop_na(gini, gdpgrowth) |>
  ggplot(aes(gini, gdpgrowth)) + geom_point() + geom_smooth(method = "lm", formula = y~x, se = FALSE)
```



Models

```
YEAR <- 2018
model0 <- df_gini_gdpgrowth2 |> filter (year == YEAR) |> drop_na(gini, gdpgrowth) |> lm(gdpgrowth ~ gini)
model0 |> summary()
```

```
##
## Call:
## lm(formula = gdpgrowth ~ gini, data = drop_na(filter(df_gini_gdpgrowth2,
##   year == YEAR), gini, gdpgrowth))
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-6.1000	-1.3442	-0.0737	1.5272	4.7402

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	4.65888	1.25244	3.720	0.000355 ***
gini	-0.02848	0.03448	-0.826	0.411083

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.154 on 86 degrees of freedom
## Multiple R-squared:  0.007871,    Adjusted R-squared:  -0.003665
## F-statistic: 0.6823 on 1 and 86 DF,  p-value: 0.4111
```

```
model0 |> summary() |> glance()
```

```
## # A tibble: 1 x 8
##   r.squared adj.r.squared sigma statistic p.value    df df.residual  nobs
##   <dbl>      <dbl> <dbl>      <dbl>   <dbl> <dbl>      <int> <dbl>
## 1  0.00787    -0.00367  2.15      0.682   0.411     1         86     88
```

```
model0 |> summary() |> tidy()
```

```
## # A tibble: 2 x 5
##   term          estimate std.error statistic  p.value
##   <chr>          <dbl>     <dbl>      <dbl>    <dbl>
## 1 (Intercept)    4.66      1.25      3.72  0.000355
## 2 gini          -0.0285    0.0345    -0.826  0.411
```

```
model0 |> augment() |> head()
```

```
## # A tibble: 6 x 8
##   gdpgrowth gini .fitted .resid  .hat .sigma .cooksd .std.resid
##   <dbl> <dbl>   <dbl> <dbl>  <dbl> <dbl>   <dbl>      <dbl>
## 1    4.02  30.1    3.80  0.218 0.0194  2.17  0.000103    0.102
## 2   -1.32  51.3    3.20 -4.51 0.0737  2.11  0.189      -2.18
## 3   -2.62  41.3    3.48 -6.10 0.0194  2.06  0.0808     -2.86
## 4    5.20  34.4    3.68  1.52 0.0118  2.16  0.00301     0.710
## 5    2.88  34.3    3.68 -0.799 0.0119  2.16  0.000837    -0.373
## 6    2.43  30.8    3.78 -1.36 0.0175  2.16  0.00360     -0.635
```