EDA

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Read in Data

```
library(tidyverse)
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr
       1.1.4
                   v readr
                                2.1.5
v forcats 1.0.0
                    v stringr
                                1.5.1
v ggplot2 3.5.1
                  v tibble 3.2.1
v lubridate 1.9.3
                    v tidyr
                                1.3.1
           1.0.2
v purrr
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()
                masks stats::lag()
i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become
library(here)
here() starts at C:/Users/10415/Documents/MSc Biostats/CHL8010/version control/CHL8010_Yanya
data <- read_csv(here("data", "merged.csv"))</pre>
Rows: 3720 Columns: 21
-- Column specification -----
Delimiter: ","
chr (3): country_name, ISO, region
dbl (18): year, gdp1000, OECD, OECD2023, popdens, urban, agedep, male_edu, t...
i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

Quick Look at Data

First look at the top lines.

data %>% head()

- # A tibble: 6 x 21 year gdp1000 OECD OECD2023 popdens urban agedep country_name ISO region <dbl> <dbl> <chr> <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> 14.1 16.3 1 Afghanistan AFG Southern~ 2000 NA0 108. 2 Afghanistan AFG Southern~ 2001 NA0 0 14.2 16.3 109. 3 Afghanistan 14.3 16.4 AFG Southern~ 2002 0.184 0 0 109. 4 Afghanistan AFG Southern~ 2003 0.200 0 0 14.4 16.6 109. AFG 5 Afghanistan 2004 0.222 0 0 15.2 16.7 109. Southern~ 0 6 Afghanistan AFG Southern~ 2005 0.255 0 15.3 16.9 108. # i 11 more variables: male_edu <dbl>, temp <dbl>, rainfall1000 <dbl>,
- death <dbl>, conflict <dbl>, maternalMor <dbl>, infantMor <dbl>,
- neonatalMor <dbl>, under5Mor <dbl>, drought <dbl>, earthquake <dbl>

Then the bottom lines

data %>% tail()

- # A tibble: 6 x 21 country_name ISO region year gdp1000 OECD OECD2023 popdens urban agedep <chr> <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> 1 Zimbabwe 1.41 26.5 24.4 85.9 ZWE Sub-Saha~ 2014 0 0 24.8 2 Zimbabwe ZWE Sub-Saha~ 2015 1.41 0 0 26.5 85.1 3 Zimbabwe ZWE Sub-Saha~ 2016 1.42 0 0 26.5 25.0 84.1 4 Zimbabwe ZWE Sub-Saha~ 2017 1.19 0 0 26.5 25.3 83.1 5 Zimbabwe ZWE Sub-Saha~ 2018 2.27 0 0 26.5 25.5 82.1 6 Zimbabwe ZWE Sub-Saha~ 2019 1.42 0 0 26.5 25.7 81.2
- # i 11 more variables: male_edu <dbl>, temp <dbl>, rainfall1000 <dbl>,
- death <dbl>, conflict <dbl>, maternalMor <dbl>, infantMor <dbl>,
- neonatalMor <dbl>, under5Mor <dbl>, drought <dbl>, earthquake <dbl>

Now randomly select a few lines

```
data \%>% slice_sample(n = 6)
```

```
# A tibble: 6 x 21
  country_name
                  IS0
                         region year gdp1000
                                                OECD OECD2023 popdens urban agedep
                   <chr> <chr>
                                <dbl>
                                                         <dbl>
                                                                 <dbl> <dbl>
  <chr>>
                                         <dbl> <dbl>
                                                                               <dbl>
1 Niger
                                                                  9.46 11.1
                                                                               105.
                  NER
                         Sub-S~
                                 2008
                                         0.472
                                                   0
                                                             0
2 Guatemala
                  GTM
                         Latin~
                                 2008
                                         2.80
                                                   0
                                                             0
                                                                 20.7
                                                                       31.4
                                                                                79.6
                                                                 53.3
3 Lebanon
                  LBN
                         Weste~
                                 2003
                                         4.46
                                                   0
                                                             0
                                                                       48.7
                                                                                55.1
4 Equatorial Gui~ GNQ
                         Sub-S~
                                 2000
                                         1.53
                                                   0
                                                             0
                                                                  0
                                                                         6.87
                                                                                84.7
5 Kenya
                  KEN
                         Sub-S~
                                 2008
                                         0.916
                                                   0
                                                             0
                                                                 21.9
                                                                       37.7
                                                                                84.2
                                 2015 41.1
6 Germany
                  DEU
                         Weste~
                                                   1
                                                             1
                                                                 36.0
                                                                       43.0
                                                                                51.9
# i 11 more variables: male_edu <dbl>, temp <dbl>, rainfall1000 <dbl>,
    death <dbl>, conflict <dbl>, maternalMor <dbl>, infantMor <dbl>,
    neonatalMor <dbl>, under5Mor <dbl>, drought <dbl>, earthquake <dbl>
```

Check the class of all variables

data %>% glimpse()

```
Rows: 3,720
Columns: 21
$ country_name <chr> "Afghanistan", "Afghanistan", "Afghanistan", "Afghanistan"
             <chr> "AFG", "AFG", "AFG", "AFG", "AFG", "AFG", "AFG", "AFG", "~
$ ISO
             <chr> "Southern Asia", "Southern Asia", "Southern Asia", "South~
$ region
$ year
             <dbl> 2000, 2001, 2002, 2003, 2004, 2005, 2006, 2007, 2008, 200~
             <dbl> NA, NA, 0.1835328, 0.2004626, 0.2216576, 0.2550551, 0.274~
$ gdp1000
$ OECD
             $ OECD2023
             $ popdens
             <dbl> 14.13654, 14.23156, 14.32270, 14.40691, 15.21947, 15.3361~
$ urban
             <dbl> 16.25324, 16.25661, 16.42654, 16.60701, 16.71367, 16.8509~
             <dbl> 108.34663, 108.98989, 109.34716, 109.44753, 109.28682, 10~
$ agedep
$ male_edu
             <dbl> 2.762086, 2.856936, 2.954241, 3.054121, 3.156706, 3.26213~
             <dbl> 12.69959, 12.85570, 12.71081, 12.16592, 13.04643, 12.2314~
$ temp
$ rainfall1000 <dbl> 0.2763704, 0.2793079, 0.3805710, 0.4288939, 0.3754336, 0.~
$ death
             <dbl> 5065, 5394, 5553, 1157, 944, 817, 1711, 4982, 7020, 5660,~
             $ conflict
$ maternalMor
             <dbl> 1450, 1390, 1300, 1240, 1180, 1140, 1120, 1090, 1030, 993~
             <dbl> 90.5, 87.9, 85.3, 82.7, 80.0, 77.3, 74.6, 71.9, 69.2, 66.~
$ infantMor
$ neonatalMor
             <dbl> 60.9, 59.7, 58.5, 57.2, 55.9, 54.6, 53.2, 51.7, 50.3, 48.~
$ under5Mor
             <dbl> 129.2, 125.2, 121.1, 116.9, 112.6, 108.4, 104.1, 99.9, 95~
             <dbl> 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 1, ~
$ drought
$ earthquake
             <dbl> 0, 1, 1, 1, 1, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 0, 0, ~
```

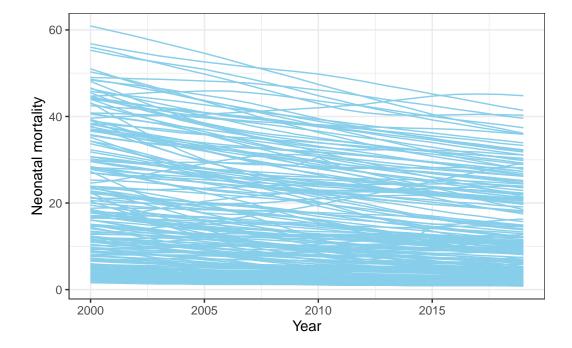
And let's see some basic statistics of the variables

country_name	ISO	region	year
Length: 3720	Length: 3720	Length: 3720	Min. :2000
Class : character	Class :characte	r Class:chara	acter 1st Qu.:2005
Mode :character	Mode :characte	r Mode :chara	acter Median :2010
			Mean :2010
			3rd Qu.:2014
			Max. :2019
gdp1000	OECD	0ECD2023	popdens
Min. : 0.1105	Min. :0.000	Min. :0.0000	Min. : 0.00
1st Qu.: 1.2383	1st Qu.:0.000	1st Qu.:0.0000	1st Qu.:14.79
Median: 4.0719	Median :0.000	Median :0.0000	Median :27.52
Mean : 11.4917	Mean :0.171	Mean :0.1882	Mean :30.57
3rd Qu.: 13.1531	3rd Qu.:0.000	3rd Qu.:0.0000	3rd Qu.:40.72
Max. :123.6787	Max. :1.000	Max. :1.0000	Max. :99.86
NA's :62			NA's :20
urban	agedep	male_edu	temp
Min. : 0.1025	Min. : 16.17	Min. : 1.067	Min. :-2.405
1st Qu.:17.2872	1st Qu.: 47.94	1st Qu.: 5.904	1st Qu.:12.928
Median :30.2535	Median : 55.51	Median : 8.368	Median :21.958
Mean :30.6948	Mean : 61.94	Mean : 8.258	Mean :19.625
3rd Qu.:41.6558	3rd Qu.: 77.11	3rd Qu.:10.849	3rd Qu.:25.869
Max. :93.4135	Max. :111.48	Max. :14.441	Max. :29.676
NA's :20		NA's :20	NA's :20
rainfall1000	death	conflict	maternalMor
Min. :0.01993	Min. : 0.0	Min. :0.0000	O Min. : 2.0
1st Qu.:0.59146	1st Qu.: 0.0	1st Qu.:0.0000	0 1st Qu.: 17.0
Median :1.01288	Median: 0.0	Median :0.0000	O Median : 66.0
Mean :1.20216	Mean : 361.1	Mean :0.1892	2 Mean : 210.6
3rd Qu.:1.68706	3rd Qu.: 2.0	3rd Qu.:0.0000	3rd Qu.: 299.8
Max. :4.71081	Max. :78644.0	Max. :1.0000	0 Max. :2480.0
NA's :20			NA's :426
infantMor	${\tt neonatalMor}$	under5Mor	drought
Min. : 1.60	Min. : 0.80 M	in. : 2.00	Min. :0.00000
1st Qu.: 7.60	1st Qu.: 4.90 1	st Qu.: 9.00	1st Qu.:0.00000
Median : 18.90	Median:12.10 M	edian : 22.20	Median :0.00000
Mean : 28.90	Mean :16.18 M	ean : 40.50	Mean :0.08737
3rd Qu.: 44.52	3rd Qu.:25.32 3	rd Qu.: 61.33	3rd Qu.:0.00000
Max. :138.10	Max. :60.90 M	ax. :224.90	Max. :1.00000
NA's :20	NA's :20 N	A's :20	

earthquake Min. :0.00000 1st Qu.:0.00000 Median :0.00333 3rd Qu.:0.00000 Max. :1.00000

Mortality Trend

```
data |>
  ggplot(aes(x = year, y = neonatalMor, group = ISO)) +
  geom_line(color = "skyblue") +
  xlim(c(2000,2019)) +
  labs(y = "Neonatal mortality", x = "Year") +
  theme_bw()
```



Mortality Trend by OECD

```
data |>
  ggplot(aes(x = year, y = maternalMor, group = ISO)) +
  geom_line(aes(color = as.factor(conflict)), alpha = 0.5) +
  xlim(c(2000,2019)) +
  scale_y_continuous(trans='log10') +
  labs(y = "Maternal mortality", x = "Year", color = "Armed conflict") +
  theme_bw()
```

