

EDA of Armed Conflict

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```
library(here)
```

here() starts at /Users/apple/Documents/chl8010class2/armed_conflict

```
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    3.4.4      v tibble     3.2.1
v lubridate  1.9.3      v tidyr      1.3.1
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
```

```
finaldata <- read.csv(here("data", "finaldata.csv"), header = TRUE)
names(finaldata)
```

```
[1] "ISO"          "year"          "Country.Name" "MatMor"        "InfMor"
[6] "NeoMor"       "U5Mor"         "total_best"   "conflict"      "drought"
[11] "earthquake"   "region"        "gdp1000"      "OECD"          "OECD2023"
[16] "popdens"      "urban"         "agedep"       "male_edu"      "temp"
[21] "rainfall1000"
```

```
finaldata$earthquake[is.na(finaldata$earthquake)] <- 0
finaldata$drought[is.na(finaldata$drought)] <- 0
```

```
finaldata %>%
  dplyr::filter(Country.Name == "Canada")
```

	ISO	year	Country.Name	MatMor	InfMor	NeoMor	U5Mor	total_best	conflict	drought
1	CAN	2000	Canada	9	5.3	3.8	6.2	11	0	0
2	CAN	2001	Canada	10	5.3	3.8	6.2	23	0	0
3	CAN	2002	Canada	10	5.3	3.9	6.2	1	0	0
4	CAN	2003	Canada	10	5.3	3.9	6.2	0	0	0
5	CAN	2004	Canada	10	5.3	3.9	6.1	0	0	0
6	CAN	2005	Canada	11	5.2	3.9	6.1	0	0	0
7	CAN	2006	Canada	11	5.2	3.9	6.0	0	0	0
8	CAN	2007	Canada	11	5.1	3.8	6.0	0	0	0
9	CAN	2008	Canada	12	5.1	3.8	5.9	0	0	0
10	CAN	2009	Canada	12	5.0	3.8	5.8	0	0	0
11	CAN	2010	Canada	11	5.0	3.8	5.7	0	0	0
12	CAN	2011	Canada	11	4.9	3.7	5.7	0	0	0
13	CAN	2012	Canada	11	4.9	3.7	5.6	0	0	0
14	CAN	2013	Canada	11	4.8	3.6	5.5	0	0	0
15	CAN	2014	Canada	11	4.7	3.6	5.4	0	0	0
16	CAN	2015	Canada	11	4.7	3.6	5.4	0	0	0
17	CAN	2016	Canada	10	4.6	3.5	5.3	0	0	0
18	CAN	2017	Canada	10	4.6	3.4	5.2	0	0	0
19	CAN	2018	Canada	NA	4.5	3.3	5.1	0	0	0
20	CAN	2019	Canada	NA	4.4	3.3	5.1	0	0	0

	earthquake	region	gdp1000	OECD	OECD2023	popdens	urban
1	0	Northern America	24.27100	1	1	66.19704	56.14335
2	0	Northern America	23.82206	1	1	66.45361	56.40270
3	0	Northern America	24.25534	1	1	66.71112	56.67093
4	0	Northern America	28.30046	1	1	66.96384	56.94365
5	0	Northern America	32.14368	1	1	67.21715	57.20020
6	0	Northern America	36.38251	1	1	67.47283	57.41671
7	0	Northern America	40.50406	1	1	67.73674	57.59143
8	0	Northern America	44.65990	1	1	67.99444	57.75691
9	0	Northern America	46.71051	1	1	68.25765	57.97905
10	0	Northern America	40.87631	1	1	68.53354	58.24228
11	0	Northern America	47.56208	1	1	68.80739	58.52809
12	0	Northern America	52.22370	1	1	69.04842	58.81437
13	0	Northern America	52.66909	1	1	69.27604	59.05573

```

14      0 Northern America 52.63517    1      1 69.50772 59.19713
15      0 Northern America 50.95600    1      1 69.76876 59.30361
16      0 Northern America 43.59614    1      1 69.98853 59.42627
17      0 Northern America 42.31560    1      1 70.21484 59.50521
18      0 Northern America 45.12943    1      1 70.40863 59.59325
19      0 Northern America 46.54864    1      1 70.63614 59.68433
20      0 Northern America 46.32867    1      1 70.83794 59.75984

      agedep male_edu      temp rainfall1000
1  46.34463 12.30281 5.486244    0.9971559
2  45.89632 12.35258 6.469105    0.8644873
3  45.46660 12.40182 5.979147    0.9460938
4  45.07468 12.45053 5.416964    1.0189234
5  44.67374 12.49870 5.556961    1.0008237
6  44.26641 12.54635 6.187472    1.0367199
7  43.96370 12.59349 6.895084    1.0917386
8  43.83612 12.64015 5.900051    1.0134091
9  43.85426 12.68634 5.650118    1.0693435
10 43.94937 12.73207 5.398867    0.9928497
11 44.13587 12.77735 6.781766    1.0379754
12 44.53578 12.82218 6.269133    1.1343442
13 45.18393 12.86660 7.249497    0.9747708
14 45.95404 12.91059 5.954381    1.0282075
15 46.75493 12.95414 5.584650    1.0377695
16 47.59164 12.99723 6.436884    0.9632446
17 48.41410 13.03988 7.184514    0.9677826
18 49.14806 13.08210 6.539669    1.0995322
19 49.80166 13.12388 6.539677    1.0991469
20 50.47739 13.16522 6.539633    1.0987523

```

```
summary(finaldata)
```

ISO	year	Country.Name	MatMor
Length:3720	Min. :2000	Length:3720	Min. : 2.0
Class :character	1st Qu.:2005	Class :character	1st Qu.: 17.0
Mode :character	Median :2010	Mode :character	Median : 66.0
	Mean :2010		Mean : 210.6
	3rd Qu.:2014		3rd Qu.: 299.8
	Max. :2019		Max. :2480.0
			NA's :426
InfMor	NeoMor	U5Mor	total_best
Min. : 1.60	Min. : 0.80	Min. : 2.00	Min. : 0.0
1st Qu.: 7.60	1st Qu.: 4.90	1st Qu.: 9.00	1st Qu.: 0.0

Median : 18.90	Median :12.10	Median : 22.20	Median : 0.0
Mean : 28.90	Mean :16.18	Mean : 40.50	Mean : 361.1
3rd Qu.: 44.52	3rd Qu.:25.32	3rd Qu.: 61.33	3rd Qu.: 2.0
Max. :138.10	Max. :60.90	Max. :224.90	Max. :78644.0
NA's :20	NA's :20	NA's :20	
conflict	drought	earthquake	region
Min. :0.0000	Min. :0.00000	Min. :0.00000	Length:3720
1st Qu.:0.0000	1st Qu.:0.00000	1st Qu.:0.00000	Class :character
Median :0.0000	Median :0.00000	Median :0.00000	Mode :character
Mean :0.1892	Mean :0.08737	Mean :0.08333	
3rd Qu.:0.0000	3rd Qu.:0.00000	3rd Qu.:0.00000	
Max. :1.0000	Max. :1.00000	Max. :1.00000	
gdp1000	OECD	OECD2023	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			
Min. :0.01993			
1st Qu.:0.59146			
Median :1.01288			
Mean :1.20216			
3rd Qu.:1.68706			
Max. :4.71081			
NA's :20			

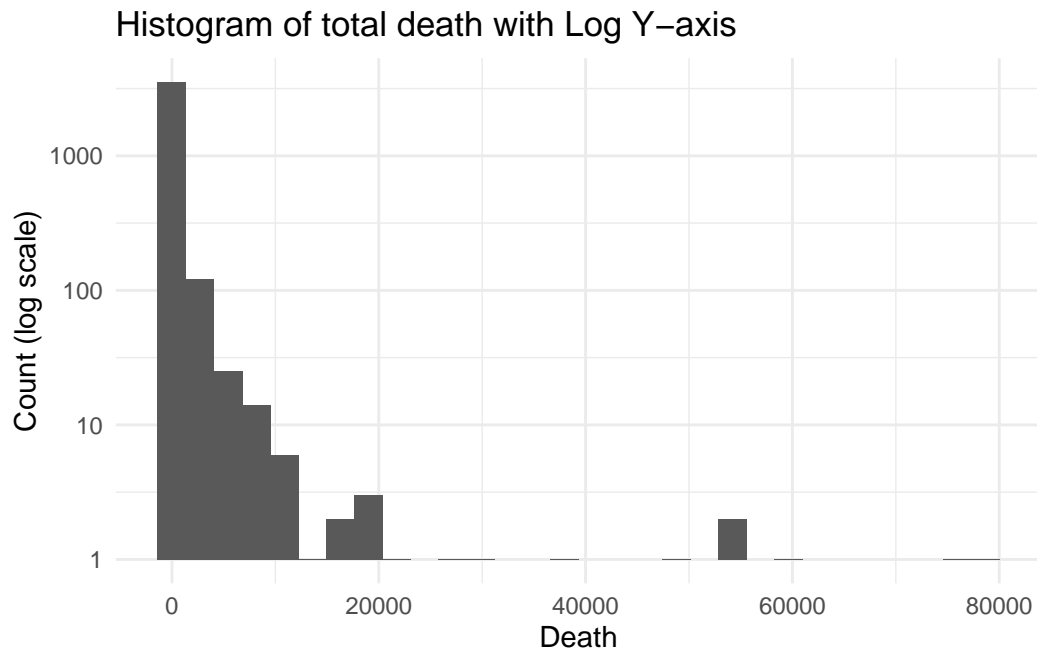
Maternal mortalities is the outcome with the most missingness of 426, followed by gdp with a missing count of 62. There seems to be a country that was not included in the covariates data, hence we see a missing count of 20 across several covariates.

```
library(ggplot2)
ggplot(data.frame(finaldata$total_best), aes(x = finaldata$total_best)) +
  geom_histogram() +
  scale_y_log10() +
  labs(title = "Histogram of total death with Log Y-axis",
       x = "Death",
       y = "Count (log scale)") +
  theme_minimal()
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

Warning: Transformation introduced infinite values in continuous y-axis

Warning: Removed 13 rows containing missing values (`geom_bar()`).



We see a lot outliers at around 55000 death, these could be caused by war happening at certain region.

```
library(ggplot2)
ggplot(data.frame(finaldata$MatMor), aes(x = finaldata$MatMor)) +
  geom_histogram() +
  scale_y_log10() +
```

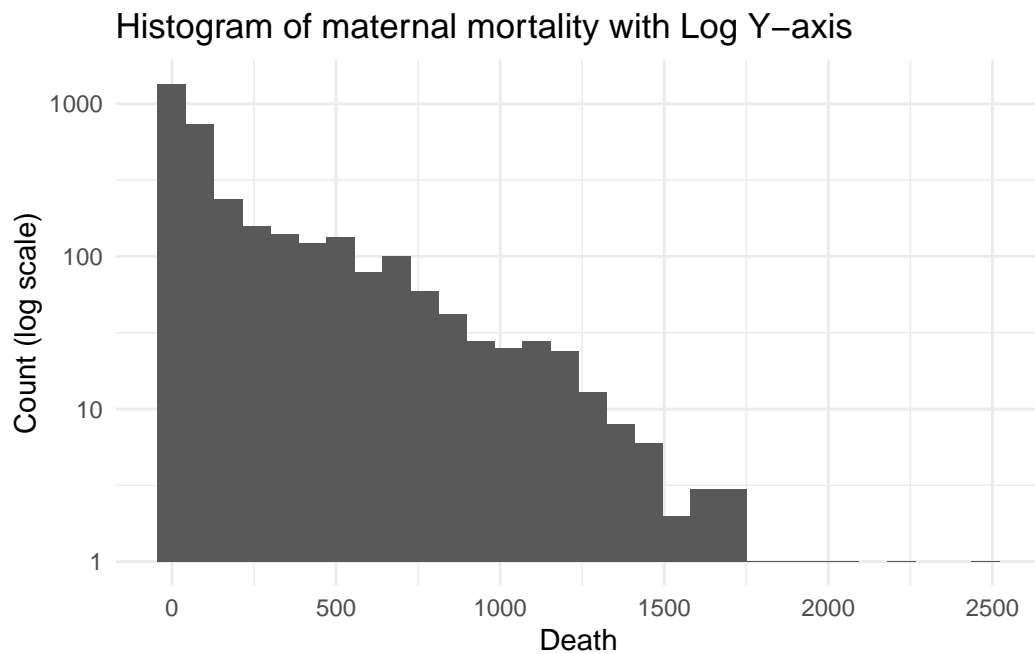
```
labs(title = "Histogram of maternal mortality with Log Y-axis",
     x = "Death",
     y = "Count (log scale)") +
theme_minimal()
```

`stat_bin()` using `bins = 30`. Pick better value with `binwidth`.

Warning: Removed 426 rows containing non-finite values (`stat_bin()`).

Warning: Transformation introduced infinite values in continuous y-axis

Warning: Removed 3 rows containing missing values (`geom_bar()`).

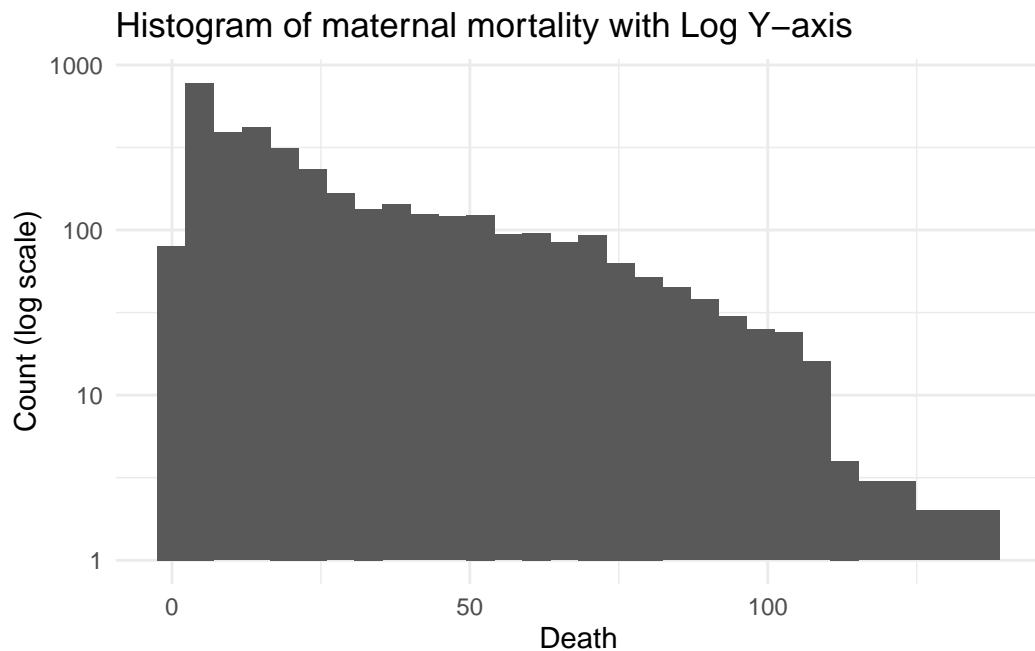


Maternal mortality looks fine, and not any outliers.

```
library(ggplot2)
ggplot(data.frame(finaldata$InfMor), aes(x = finaldata$InfMor)) +
  geom_histogram() +
  scale_y_log10() +
  labs(title = "Histogram of maternal mortality with Log Y-axis",
       x = "Death",
       y = "Count (log scale)") +
  theme_minimal()
```

``stat_bin()`` using ``bins = 30``. Pick better value with ``binwidth``.

Warning: Removed 20 rows containing non-finite values (``stat_bin()``).

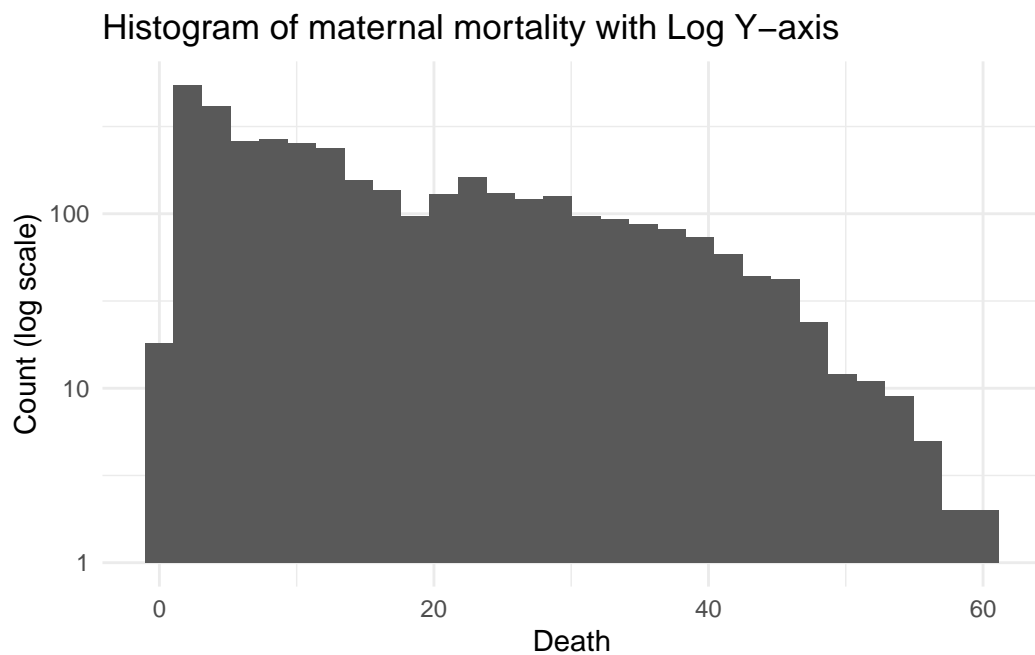


Infant mortalities looks fine as well.

```
library(ggplot2)
ggplot(data.frame(finaldata$NeoMor), aes(x = finaldata$NeoMor)) +
  geom_histogram() +
  scale_y_log10() +
  labs(title = "Histogram of maternal mortality with Log Y-axis",
       x = "Death",
       y = "Count (log scale)") +
  theme_minimal()
```

``stat_bin()`` using ``bins = 30``. Pick better value with ``binwidth``.

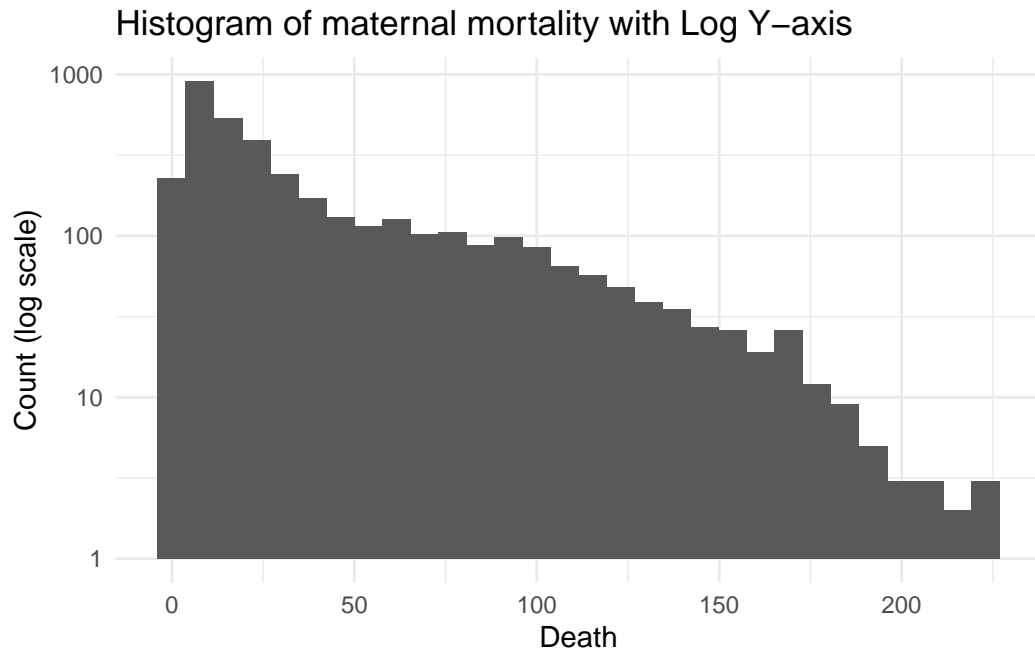
Warning: Removed 20 rows containing non-finite values (``stat_bin()``).



```
library(ggplot2)
ggplot(data.frame(finaldata$U5Mor), aes(x = finaldata$U5Mor)) +
  geom_histogram() +
  scale_y_log10() +
  labs(title = "Histogram of maternal mortality with Log Y-axis",
       x = "Death",
       y = "Count (log scale)") +
  theme_minimal()
```

``stat_bin()`` using ``bins = 30``. Pick better value with ``binwidth``.

Warning: Removed 20 rows containing non-finite values (``stat_bin()``).



```
cor(finaldata[,c(4:7)], use = "complete.obs")
```

	MatMor	InfMor	NeoMor	U5Mor
MatMor	1.0000000	0.8785612	0.8354908	0.8994877
InfMor	0.8785612	1.0000000	0.9590878	0.9861117
NeoMor	0.8354908	0.9590878	1.0000000	0.9278720
U5Mor	0.8994877	0.9861117	0.9278720	1.0000000

We can see the main outcomes are highly correlated, so we need to be cautious when running them one by one.