

Exploratory data analysis

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First of all, I would like to perform `summary()` to check some generic information about our data set.

```
library(here)
```

`here()` starts at `/Users/apple/Desktop/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    3.5.1      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
```

```
source("R/FinalMergedData.R")
```

Warning: Some values were not matched unambiguously: Africa Eastern and Southern, Africa West

`summarise()` has grouped output by 'year'. You can override using the
`.groups` argument.
`summarise()` has grouped output by 'Year'. You can override using the
`.groups` argument.

```
head(allfinal,n=10)
```

	country_name	ISO	region	year	gdp1000	OECD	OECD2023	popdens
1	Afghanistan	AFG	Southern Asia	2000	NA	0	0	14.13654
2	Afghanistan	AFG	Southern Asia	2001	NA	0	0	14.23156
3	Afghanistan	AFG	Southern Asia	2002	0.1835328	0	0	14.32270
4	Afghanistan	AFG	Southern Asia	2003	0.2004626	0	0	14.40691
5	Afghanistan	AFG	Southern Asia	2004	0.2216576	0	0	15.21947
6	Afghanistan	AFG	Southern Asia	2005	0.2550551	0	0	15.33619
7	Afghanistan	AFG	Southern Asia	2006	0.2740005	0	0	15.43982
8	Afghanistan	AFG	Southern Asia	2007	0.3750781	0	0	15.65217
9	Afghanistan	AFG	Southern Asia	2008	0.3878492	0	0	15.74447
10	Afghanistan	AFG	Southern Asia	2009	0.4438452	0	0	15.83043
	urban	agedep	male_edu	temp	rainfall1000	totaldeath	armconf1	
1	16.25324	108.3466	2.762086	12.69959	0.2763704	5065	1	
2	16.25661	108.9899	2.856936	12.85570	0.2793079	5394	1	
3	16.42654	109.3472	2.954241	12.71081	0.3805710	5553	1	
4	16.60701	109.4475	3.054121	12.16592	0.4288939	1157	1	
5	16.71367	109.2868	3.156706	13.04643	0.3754336	944	1	
6	16.85096	107.9646	3.262133	12.23141	0.4415680	817	1	
7	16.98105	106.3262	3.370551	12.96153	0.4437097	1711	1	
8	17.12259	108.3381	3.482112	12.47451	0.4092555	4982	1	
9	17.26919	109.2404	3.596977	12.63527	0.3901204	7020	1	
10	17.43508	106.8458	3.715306	12.61764	0.4808727	5660	1	
	MaternalMortalityRate	InfantMortalityRate	NeonatalMortalityRate					
1		1450	90.5			60.9		
2		1390	87.9			59.7		
3		1300	85.3			58.5		
4		1240	82.7			57.2		
5		1180	80.0			55.9		
6		1140	77.3			54.6		
7		1120	74.6			53.2		
8		1090	71.9			51.7		
9		1030	69.2			50.3		
10		993	66.7			48.9		
	Under5MortalityRate	drought	earthquake					
1	129.2	1	0					

2	125.2	0	1
3	121.1	0	1
4	116.9	0	1
5	112.6	0	1
6	108.4	0	1
7	104.1	1	1
8	99.9	0	0
9	95.7	1	0
10	91.7	0	1

```
tail(allfinal, n=10)
```

	country_name	ISO		region	year	gdp1000	OECD	OECD2023	popdens
3711	Zimbabwe	ZWE	Sub-Saharan	Africa	2010	0.9378403	0	0	25.51039
3712	Zimbabwe	ZWE	Sub-Saharan	Africa	2011	1.0826158	0	0	25.53206
3713	Zimbabwe	ZWE	Sub-Saharan	Africa	2012	1.2901940	0	0	25.55349
3714	Zimbabwe	ZWE	Sub-Saharan	Africa	2013	1.4083678	0	0	25.53286
3715	Zimbabwe	ZWE	Sub-Saharan	Africa	2014	1.4070343	0	0	26.52884
3716	Zimbabwe	ZWE	Sub-Saharan	Africa	2015	1.4103292	0	0	26.54454
3717	Zimbabwe	ZWE	Sub-Saharan	Africa	2016	1.4217878	0	0	26.53811
3718	Zimbabwe	ZWE	Sub-Saharan	Africa	2017	1.1921070	0	0	26.49281
3719	Zimbabwe	ZWE	Sub-Saharan	Africa	2018	2.2691770	0	0	26.47943
3720	Zimbabwe	ZWE	Sub-Saharan	Africa	2019	1.4218686	0	0	26.46341
	urban	agedep	male_edu	temp	rainfall1000	totaldeath	armconf1		
3711	23.28851	85.56457	8.250225	21.53473	0.7290925		0		0
3712	23.43075	86.40049	8.358820	20.87452	0.8582386		0		0
3713	23.70160	86.71712	8.466529	20.98071	0.6259767		1		0
3714	24.04603	86.44543	8.573429	20.77221	0.6717220		1		0
3715	24.40427	85.87550	8.679591	20.87651	0.6777257		0		0
3716	24.75233	85.08337	8.785078	21.45470	0.4490721		0		0
3717	25.02842	84.11222	8.889947	21.39290	0.4939246		0		0
3718	25.29333	83.10129	8.994252	20.85962	0.9533149		0		0
3719	25.53759	82.12335	9.098048	20.86041	0.9535655		0		0
3720	25.70572	81.20786	9.201384	20.86120	0.9538138		4		0
	MaternalMortalityRate	InfantMortalityRate	NeonatalMortalityRate						
3711		598		52.1				30.8	
3712		557		50.8				30.1	
3713		528		46.5				29.4	
3714		509		44.8				28.7	
3715		494		42.9				28.2	
3716		480		42.1				27.8	
3717		468		40.8				27.4	

3718	458	39.9	27.0
3719	NA	38.8	26.6
3720	NA	38.1	26.2
Under5MortalityRate drought earthquake			
3711	86.4	1	0
3712	80.8	0	0
3713	72.2	0	0
3714	66.3	1	0
3715	62.7	0	0
3716	61.3	0	0
3717	58.7	0	0
3718	57.0	1	0
3719	54.8	0	0
3720	54.2	0	0

```
summary(allfinal)
```

country_name	ISO	region	year
Length:3720	Length:3720	Length:3720	Min. :2000
Class :character	Class :character	Class :character	1st Qu.:2005
Mode :character	Mode :character	Mode :character	Median :2010
			Mean :2010
			3rd Qu.:2014
			Max. :2019
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	totaldeath	armconf1	MaternalMortalityRate

Min. :0.01993	Min. : 0.0	Min. :0.0000	Min. : 2.0
1st Qu.:0.59146	1st Qu.: 0.0	1st Qu.:0.0000	1st Qu.: 17.0
Median :1.01288	Median : 0.0	Median :0.0000	Median : 66.0
Mean :1.20216	Mean : 361.1	Mean :0.1892	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	Max. :2480.0
NA's :20			NA's :426

InfantMortalityRate	NeonatalMortalityRate	Under5MortalityRate
Min. : 1.60	Min. : 0.80	Min. : 2.00
1st Qu.: 7.60	1st Qu.: 4.90	1st Qu.: 9.00
Median : 18.90	Median :12.10	Median : 22.20
Mean : 28.90	Mean :16.18	Mean : 40.50
3rd Qu.: 44.52	3rd Qu.:25.32	3rd Qu.: 61.33
Max. :138.10	Max. :60.90	Max. :224.90
NA's :20	NA's :20	NA's :20

drought	earthquake
Min. :0.00000	Min. :0.00000
1st Qu.:0.00000	1st Qu.:0.00000
Median :0.00000	Median :0.00000
Mean :0.08737	Mean :0.08333
3rd Qu.:0.00000	3rd Qu.:0.00000
Max. :1.00000	Max. :1.00000

From the `summery()`, there are 426 missing values for Maternal Mortality Rate, 62 missing values for `gpd100`, 20 missing values for `popdens`, `urban`, `male_edu`, `temp`, `rainfall1000`, Infant Mortality Rate, Neonatal Mortality Rate and Under5 Mortality Rate.

Then I would like to look at Total Death in details.

The Minimum value for total death is 20, and Maximum is 78644. The range is pretty wide.

I would like to locate which country has the maximum total death and visualize it by year to see the patterns for this specific country.

```
# visualize Maternal Mortality Rate by year.
max_death_country <- allfinal$country_name[which.max(allfinal$totaldeath)]
print(max_death_country)
```

```
[1] "Syria"
```

```
library(ggplot2)
```

And the answer is Syria.

```
library(ggplot2)
library(gridExtra)
```

Attaching package: 'gridExtra'

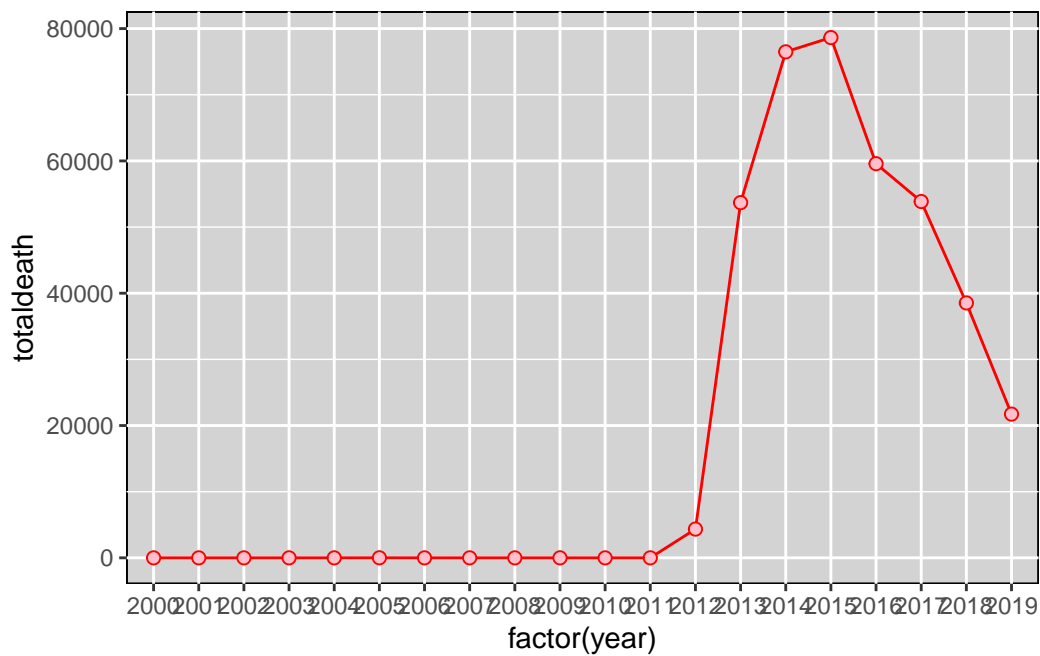
The following object is masked from 'package:dplyr':

combine

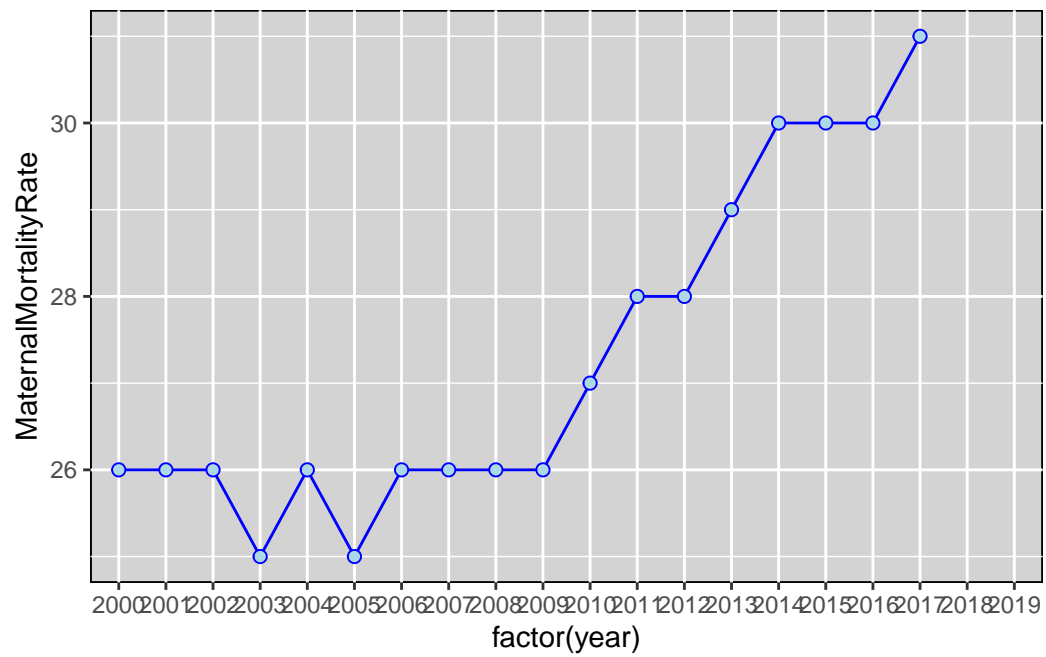
```
Syria <- allfinal[allfinal$country_name == "Syria",]
plot1 <- Syria %>%
  ggplot(mapping = aes(x = factor(year), y = totaldeath)) +
  geom_line(na.rm = TRUE, group = 1, color="red") +
  geom_point(na.rm = TRUE, shape=21, color="red", fill="pink", size=2) +
  scale_x_discrete(breaks = unique(Syria$year)) +
  theme(panel.background = element_rect(fill = "lightgrey", color = "black"))
plot2 <- Syria %>%
  ggplot(mapping = aes(x = factor(year), y = MaternalMortalityRate)) +
  geom_line(na.rm = TRUE, group = 1, color = "blue") +
  geom_point(na.rm = TRUE, shape = 21, color = "blue", fill = "lightblue", size = 2) +
  scale_x_discrete(breaks = unique(Syria$year)) +
  theme(panel.background = element_rect(fill = "lightgrey", color = "black"))
plot3 <- Syria %>%
  ggplot(mapping = aes(x = factor(year), y = InfantMortalityRate)) +
  geom_line(na.rm = TRUE, group = 1, color = "orange") +
  geom_point(na.rm = TRUE, shape = 21, color = "orange", fill = "orange", size = 2) +
  scale_x_discrete(breaks = unique(Syria$year)) +
  theme(panel.background = element_rect(fill = "lightgrey", color = "black"))
plot4 <- Syria %>%
  ggplot(mapping = aes(x = factor(year), y = NeonatalMortalityRate)) +
  geom_line(na.rm = TRUE, group = 1, color = "purple") +
  geom_point(na.rm = TRUE, shape = 21, color = "purple", fill = "purple", size = 2) +
  scale_x_discrete(breaks = unique(Syria$year)) +
  theme(panel.background = element_rect(fill = "lightgrey", color = "black"))
plot5 <- Syria %>%
```

```
ggplot(mapping = aes(x = factor(year), y = Under5MortalityRate)) +
  geom_line(na.rm = TRUE, group = 1, color = "green") +
  geom_point(na.rm = TRUE, shape = 21, color = "green", fill = "green", size = 2) +
  scale_x_discrete(breaks = unique(Syria$year)) +
  theme(panel.background = element_rect(fill = "lightgrey", color = "black"))
```

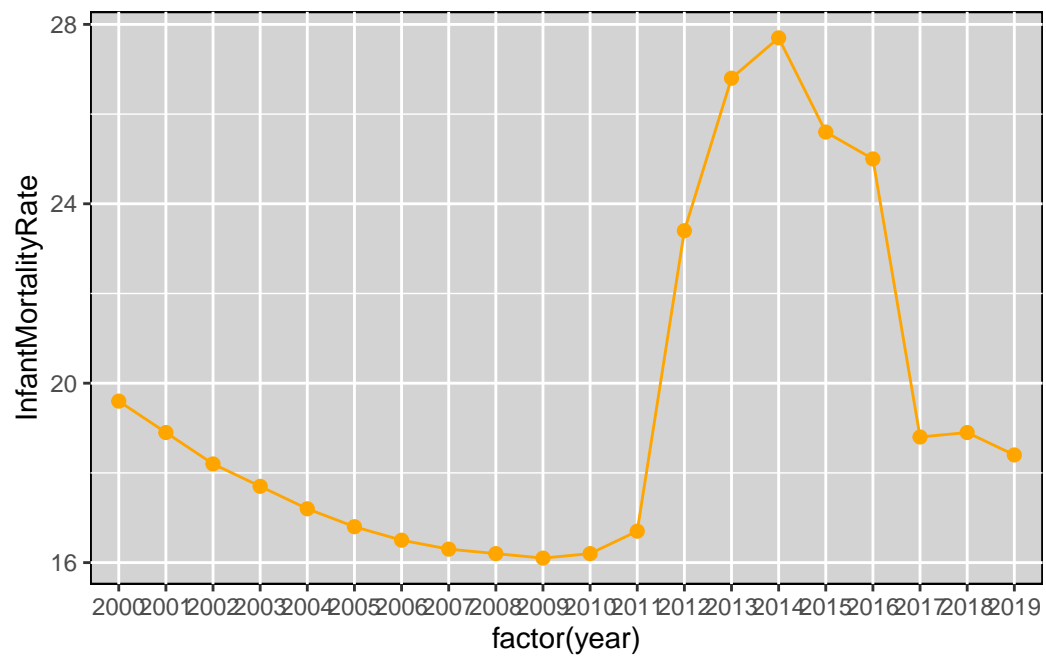
plot1



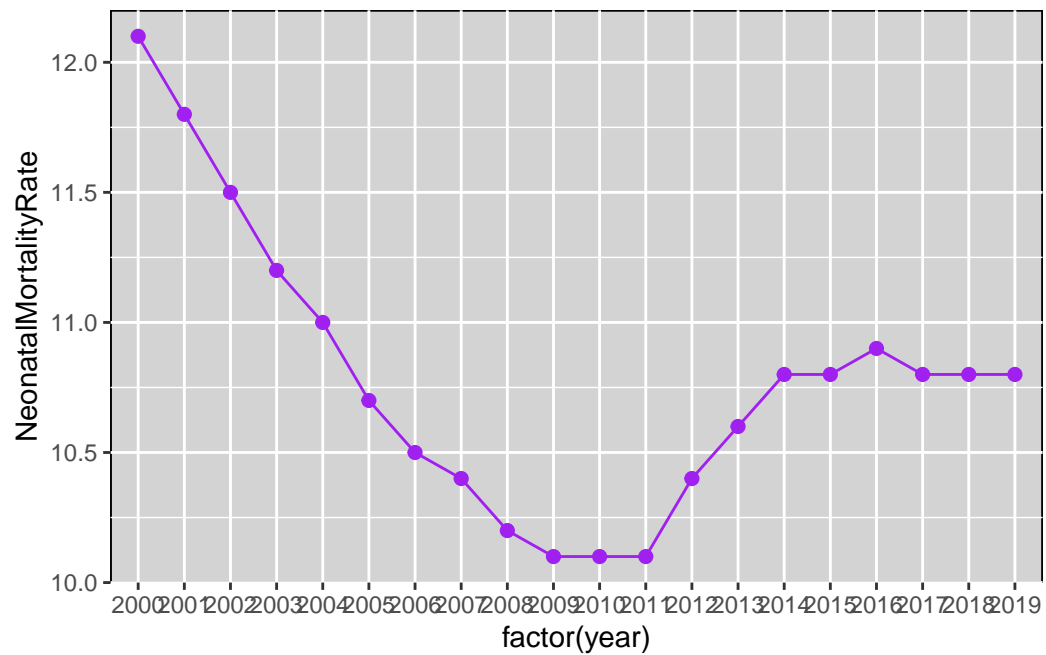
plot2



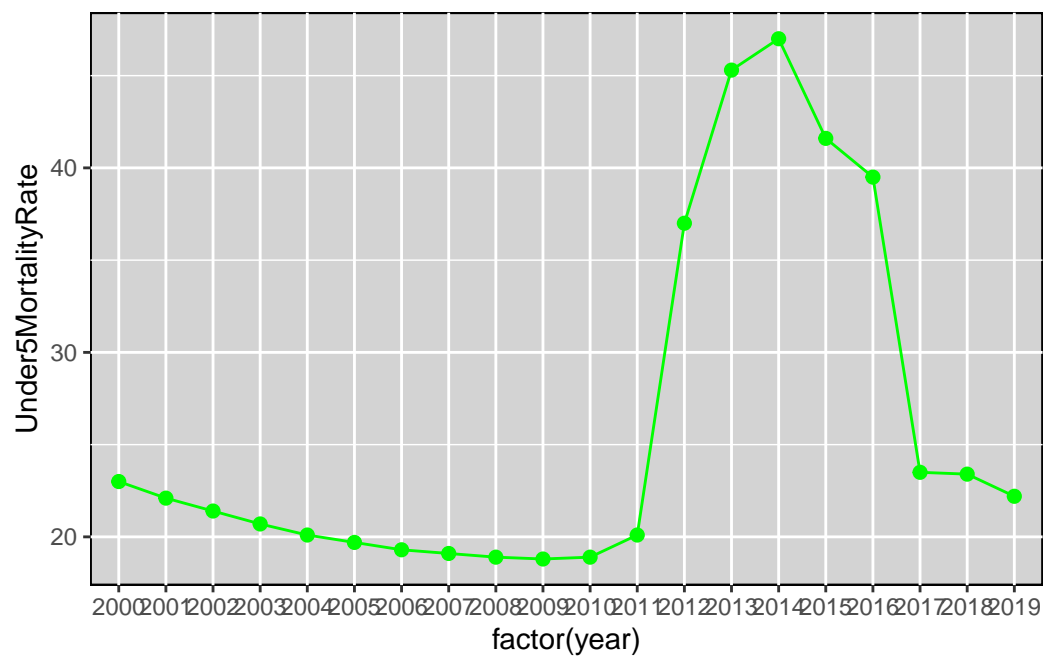
plot3



plot4



plot5



```
#plotall4 <- grid.arrange(plot2, plot3, plot4, plot5, ncol = 2)
#plotall4
```