

# Tingji-AssignmentE

```
library(tidyverse)
library(knitr)
```

## Part 1

```
wdbd_raw <- read.csv("WDBD.csv")

#Remove metadata rows at the bottom
wdbd_clean <- wdbd_raw[1:2976,] %>%
  select(-Series.Code, -Country.Code) #Series name and code are one to one
```

After briefly read the csv file, I recognized the rows after 2976 are explanation for units, so only keep 2976 rows. And the series names and codes are one to one, so remove the series.codes col. And for the same reason, remove country code col.

```
#Reshape the data set
#pivot_longer
wdbd_longer <- wdbd_clean %>%
  pivot_longer(cols = 3:9, names_to = "Year", values_to = "Value") %>%
  mutate(Year = str_extract(Year, "\\d{4}")) %>% # 提取4位数字
  mutate(Year = as.integer(Year)) #Change Year to 4 digit year number

#pivot_wider
wdbd_wider <- wdbd_longer %>%
  mutate(Value = na_if(Value, "..")) %>%
  mutate(Value = as.numeric(Value)) %>%
  pivot_wider(names_from = Series.Name,
              values_from = Value)

# summary(wdbd_wider)
wdbd_tidy <- wdbd_wider %>% select(-(38:43), -47)
```

Use pivot\_longer() to gather year columns into a single Year column, Use pivot\_wider() if you want to restructure Series IDs and Series Names. Also, remove 7 total empty cols with all NAs(Unusual)

```
summary(wdbd_tidy$`Death rate, crude (per 1,000 people)`)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
0.841	6.170	7.385	7.719	9.233	17.134	62

Death rate with Min = 0.841 which is extremely small, but it make sense in Qatar with very few population.

```
summary(wdbd_tidy$`Adjusted net savings, including particulate emission damage (current US$)`)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.	NA's
------	---------	--------	------	---------	------	------

-1.325e+10 2.429e+09 2.379e+10 1.362e+11 9.641e+10 2.870e+12 218

Adjusted net savings with e+10 values(extremely high), which is also make sense for country's saving

```
head(wdbd_tidy)
```

```
# A tibble: 6 × 43
  Country.Name Year Access to clean fuels and technolo...1 Access to clean fuel...2
  <chr>         <int>                                <dbl>                <dbl>
1 Afghanistan  2018                                14.5                 31.4
2 Afghanistan  2019                                15.6                 32.6
3 Afghanistan  2020                                16.4                 33.8
4 Afghanistan  2021                                17.4                 34.9
5 Afghanistan  2022                                18.5                 36.1
6 Afghanistan  2023                                NA                    NA
# i abbreviated names:
#   1`Access to clean fuels and technologies for cooking, rural (% of rural population)`,
#   2`Access to clean fuels and technologies for cooking (% of population)`
# i 39 more variables:
#   `Access to clean fuels and technologies for cooking, urban (% of urban population)` <dbl>,
#   `Access to electricity (% of population)` <dbl>,
#   `Access to electricity, rural (% of rural population)` <dbl>, ...
```

## Part 2

```
movie_raw = read.csv("movies.csv")
movie_clean <- movie_raw %>%
  separate_wider_delim(
    cols = genres,
    delim = "|",
    names_sep = "_",
    too_few = "align_start"
  ) %>%
  select(-genres_5, -genres_6, -genres_7, -genres_4)

movies_tidy <- movie_clean %>%
  mutate(
    year = str_extract(title, "\\(\\d{4}\\)"),
    year = as.integer(str_remove_all(year, "[()]")),
    title = str_trim(str_remove(title, "\\(\\d{4}\\)"))
  ) %>%
  relocate(year, .after = title)

# movies_longer <- movie_clean %>%
#   pivot_longer(
#     cols = starts_with("genres_"),
#     names_to = "genre_num",
#     values_to = "genre",
#     values_drop_na = TRUE
#   ) %>%
#   select(-genre_num) %>%
#   distinct(movieId, title, genre) %>%
#   mutate(value = 1)
```

```

#
# movies_dummy <- movies_longer %>%
#   pivot_wider(
#     names_from = genre,
#     values_from = value,
#     values_fill = 0
#   )
#
# movies_dummy <- movie_clean %>%
#   select(movieId, title) %>%
#   left_join(movies_dummy, by = c("movieId", "title")) %>%
#   mutate(across(where(is.numeric), ~replace_na(., 0))) # 把空的 dummy 填 0
#
# #get date from ( ) from ChatGPT
# movies_tidy <- movies_dummy %>%
#   mutate(
#     year = str_extract(title, "\\(\\d{4}\\)"),
#     year = as.integer(str_remove_all(year, "[()]")),
#     title = str_trim(str_remove(title, "\\(\\d{4}\\)"))
#   ) %>%
#   relocate(year, .after = title)

```

```

links = read.csv("links.csv")
ratings = read.csv("ratings.csv")
tags = read.csv("tags.csv")

tags_summary <- tags %>%
  group_by(movieId) %>%
  summarise(
    all_tags = str_c(unique(tag), collapse = ", ")
  )
links <- links %>%
  filter(movieId %in% movies_tidy$movieId)

avg_ratings <- ratings %>%
  group_by(movieId) %>%
  summarise(
    avg_ratings = mean(rating, na.rm = TRUE),
    num_ratings = n()
  ) %>%
  filter(movieId %in% movies_tidy$movieId)

movies_tidy <- movies_tidy %>%
  left_join(tags_summary, by = "movieId") %>%
  left_join(links, by = "movieId") %>%
  left_join(avg_ratings, by = "movieId")
head(movies_tidy)

```

# A tibble: 6 × 11

	movieId	title	year	genres_1	genres_2	genres_3	all_tags	imdbId	tmdbId
	<int>	<chr>	<int>	<chr>	<chr>	<chr>	<chr>	<int>	<int>
1	182337	Cinétracts	1968	(no gen...	<NA>	<NA>	antholo...	2.08e5	287929
2	195495	Familia	2005	Drama	<NA>	<NA>	addicti...	4.26e5	42052

```
3 3078 Liberty Heigh... 1999 Drama <NA> <NA> Hebrew,... 1.66e5 27141
4 134704 Comedy Centra... 2011 Comedy <NA> <NA> The Com... 1.99e6 296192
5 219976 47 Hours to L... 2019 Horror Thriller <NA> <NA> 7.84e6 615017
6 205715 Reis 2017 (no gen... <NA> <NA> <NA> 5.99e6 421682
# i 2 more variables: avg_ratings <dbl>, num_ratings <int>
```