

FA2_Patayon

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2026-02-02

1. Introduction

In this formative assessment, you will review Unit 1, focusing on data wrangling, manipulation, and visualization.

```
library(tidyverse)
```

```
## — Attaching core tidyverse packages ————— tidyverse 2.0.0 —
## ✓ dplyr     1.1.4    ✓ readr     2.1.5
## ✓ forcats   1.0.0    ✓ stringr   1.5.1
## ✓ ggplot2   3.5.1    ✓ tibble    3.2.1
## ✓ lubridate 1.9.4    ✓ tidyr    1.3.1
## ✓ purrr    1.0.2
## — Conflicts ————— tidyverse_conflicts() —
## ✘ dplyr::filter() masks stats::filter()
## ✘ dplyr::lag()   masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

You will perform a data analysis to study trends in tuberculosis (TB) cases worldwide over time. The two relevant datasets are:

- **who.tsv** : Information about TB cases in various countries from 1980 to 2013 (Source: 2014 WHO Global Tuberculosis Report).
- **population.csv** : Population data of each country across time (Source: The World Bank).

2. Import

2.1. Preview the contents of **who.tsv** and **population.csv** by inspecting the files.

2.2. Import the data into tibbles named **who** and **population**.

```
df_who = who

df_population = read_csv("C:\\\\Users\\\\spike\\\\Downloads\\\\Population.csv")
```

```
## Rows: 266 Columns: 68
## — Column specification ——————
## Delimiter: ","
## chr (4): Country Name, Country Code, Indicator Name, Indicator Code
## dbl (64): 1960, 1961, 1962, 1963, 1964, 1965, 1966, 1967, 1968, 1969, 1970, ...
##
## 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.
```

```
df_who
```

```

## # A tibble: 7,240 × 60
##   country iso2 iso3 year new_sp_m014 new_sp_m1524 new_sp_m2534 new_sp_m3544
##   <chr>    <chr> <chr> <dbl>      <dbl>      <dbl>      <dbl>
## 1 Afghani... AF    AFG    1980       NA        NA        NA        NA
## 2 Afghani... AF    AFG    1981       NA        NA        NA        NA
## 3 Afghani... AF    AFG    1982       NA        NA        NA        NA
## 4 Afghani... AF    AFG    1983       NA        NA        NA        NA
## 5 Afghani... AF    AFG    1984       NA        NA        NA        NA
## 6 Afghani... AF    AFG    1985       NA        NA        NA        NA
## 7 Afghani... AF    AFG    1986       NA        NA        NA        NA
## 8 Afghani... AF    AFG    1987       NA        NA        NA        NA
## 9 Afghani... AF    AFG    1988       NA        NA        NA        NA
## 10 Afghani... AF   AFG    1989      NA        NA        NA        NA
## # i 7,230 more rows
## # i 52 more variables: new_sp_m4554 <dbl>, new_sp_m5564 <dbl>,
## #   new_sp_m65 <dbl>, new_sp_f014 <dbl>, new_sp_f1524 <dbl>,
## #   new_sp_f2534 <dbl>, new_sp_f3544 <dbl>, new_sp_f4554 <dbl>,
## #   new_sp_f5564 <dbl>, new_sp_f65 <dbl>, new_sn_m014 <dbl>,
## #   new_sn_m1524 <dbl>, new_sn_m2534 <dbl>, new_sn_m3544 <dbl>,
## #   new_sn_m4554 <dbl>, new_sn_m5564 <dbl>, new_sn_m65 <dbl>, ...

```

df_population

```

## # A tibble: 266 × 68
##   `Country Name` `Country Code` `Indicator Name` `Indicator Code` `1960` `1961`
##   <chr>          <chr>        <chr>           <chr>           <dbl>  <dbl>
## 1 Aruba          ABW          Population, tot... SP.POP.TOTL  5.49e4 5.56e4
## 2 Africa Easter... AFE          Population, tot... SP.POP.TOTL  1.30e8 1.34e8
## 3 Afghanistan    AFG          Population, tot... SP.POP.TOTL  9.04e6 9.21e6
## 4 Africa Wester... AFW          Population, tot... SP.POP.TOTL  9.76e7 9.97e7
## 5 Angola          AGO          Population, tot... SP.POP.TOTL  5.23e6 5.30e6
## 6 Albania         ALB          Population, tot... SP.POP.TOTL  1.61e6 1.66e6
## 7 Andorra         AND          Population, tot... SP.POP.TOTL  9.51e3 1.03e4
## 8 Arab World      ARB          Population, tot... SP.POP.TOTL  9.15e7 9.39e7
## 9 United Arab E... ARE          Population, tot... SP.POP.TOTL  1.31e5 1.38e5
## 10 Argentina      ARG          Population, tot... SP.POP.TOTL  2.04e7 2.07e7
## # i 256 more rows
## # i 62 more variables: `1962` <dbl>, `1963` <dbl>, `1964` <dbl>, `1965` <dbl>,
## #   `1966` <dbl>, `1967` <dbl>, `1968` <dbl>, `1969` <dbl>, `1970` <dbl>,
## #   `1971` <dbl>, `1972` <dbl>, `1973` <dbl>, `1974` <dbl>, `1975` <dbl>,
## #   `1976` <dbl>, `1977` <dbl>, `1978` <dbl>, `1979` <dbl>, `1980` <dbl>,
## #   `1981` <dbl>, `1982` <dbl>, `1983` <dbl>, `1984` <dbl>, `1985` <dbl>,
## #   `1986` <dbl>, `1987` <dbl>, `1988` <dbl>, `1989` <dbl>, `1990` <dbl>, ...

```

2.3. Determine the number of rows and columns in each tibble.

dim(df_who)

```
## [1] 7240 60
```

dim(df_population)

```
## [1] 266 68
```

2.4. Check the summary of variable types for `population.csv`. Fix any anomalies and store the corrected data in `population2`.

```
glimpse(df_population)
```

```
## Rows: 266
## Columns: 68
## $ `Country Name` <chr> "Aruba", "Africa Eastern and Southern", "Afghanistan"...
## $ `Country Code` <chr> "ABW", "AFE", "AFG", "AFW", "AGO", "ALB", "AND", "ARB...
## $ `Indicator Name` <chr> "Population, total", "Population, total", "Population...
## $ `Indicator Code` <chr> "SP.POP.TOTL", "SP.POP.TOTL", "SP.POP.TOTL", "SP.POP...
## $ `1960` <dbl> 54922, 130072080, 9035043, 97630925, 5231654, 1608800...
## $ `1961` <dbl> 55578, 133534923, 9214083, 99706674, 5301583, 1659800...
## $ `1962` <dbl> 56320, 137171659, 9404406, 101854756, 5354310, 171131...
## $ `1963` <dbl> 57002, 140945536, 9604487, 104089175, 5408320, 176262...
## $ `1964` <dbl> 57619, 144904094, 9814318, 106388440, 5464187, 181413...
## $ `1965` <dbl> 58190, 149033472, 10036008, 108772632, 5521981, 18647...
## $ `1966` <dbl> 58694, 153281203, 10266395, 111246953, 5581386, 19145...
## $ `1967` <dbl> 58990, 157704381, 10505959, 113795019, 5641807, 19655...
## $ `1968` <dbl> 59069, 162329396, 10756922, 116444636, 5702699, 20222...
## $ `1969` <dbl> 59052, 167088245, 11017409, 119203521, 5763685, 20816...
## $ `1970` <dbl> 58950, 171984985, 11290128, 122086536, 5852788, 21354...
## $ `1971` <dbl> 58781, 177022314, 11567667, 125072948, 5991102, 21878...
## $ `1972` <dbl> 58047, 182126556, 11853696, 128176494, 6174262, 22431...
## $ `1973` <dbl> 58299, 187524135, 12157999, 131449942, 6388528, 22967...
## $ `1974` <dbl> 58349, 193186642, 12469127, 134911581, 6613367, 23501...
## $ `1975` <dbl> 58295, 198914573, 12773954, 138569918, 6842947, 24048...
## $ `1976` <dbl> 58368, 204802976, 13059851, 142337272, 7074664, 24585...
## $ `1977` <dbl> 58580, 210680842, 13340756, 146258576, 7317829, 25135...
## $ `1978` <dbl> 58776, 217074286, 13611441, 150402616, 7576734, 25662...
## $ `1979` <dbl> 59191, 223974122, 13655567, 154721711, 7847207, 26178...
## $ `1980` <dbl> 59909, 230792729, 13169311, 159166518, 8133872, 26719...
## $ `1981` <dbl> 60563, 238043099, 11937581, 163762473, 8435607, 27260...
## $ `1982` <dbl> 61276, 245822010, 10991378, 168585118, 8751648, 27842...
## $ `1983` <dbl> 62228, 253644643, 10917982, 173255157, 9082983, 28439...
## $ `1984` <dbl> 62901, 261458202, 11190221, 177880746, 9425917, 29044...
## $ `1985` <dbl> 61728, 269450407, 11426852, 182811038, 9779120, 29647...
## $ `1986` <dbl> 59931, 277621771, 11420074, 187889141, 10139450, 3022...
## $ `1987` <dbl> 59159, 286067346, 11387818, 193104347, 10497858, 3083...
## $ `1988` <dbl> 59331, 294498625, 11523298, 198485027, 10861291, 3142...
## $ `1989` <dbl> 60443, 302939121, 11874088, 204062274, 11238562, 3227...
## $ `1990` <dbl> 62753, 311748681, 12045660, 209566031, 11626360, 3286...
## $ `1991` <dbl> 65896, 320442961, 12238879, 215178709, 12023529, 3266...
## $ `1992` <dbl> 69005, 329082707, 13278974, 221191375, 12423712, 3247...
## $ `1993` <dbl> 73685, 338324002, 14943172, 227246778, 12827135, 3227...
## $ `1994` <dbl> 77595, 347441809, 16250794, 233360104, 13249764, 3207...
## $ `1995` <dbl> 79805, 356580375, 17065836, 239801875, 13699778, 3187...
## $ `1996` <dbl> 83021, 366138524, 17763266, 246415446, 14170973, 3168...
## $ `1997` <dbl> 86301, 375646235, 18452091, 253207584, 14660413, 3148...
## $ `1998` <dbl> 88451, 385505757, 19159996, 260297834, 15159370, 3128...
## $ `1999` <dbl> 89659, 395750933, 19887785, 267506298, 15667235, 3108...
## $ `2000` <dbl> 90588, 406156661, 20130327, 274968446, 16194869, 3089...
## $ `2001` <dbl> 91439, 416807868, 20284307, 282780717, 16747208, 3060...
## $ `2002` <dbl> 92074, 427820358, 21378117, 290841795, 17327699, 3051...
## $ `2003` <dbl> 93128, 439173286, 22733049, 299142845, 17943712, 3039...
## $ `2004` <dbl> 95138, 450928044, 23560654, 307725100, 18600423, 3026...
## $ `2005` <dbl> 97635, 463076637, 24404567, 316588476, 19291161, 3011...
## $ `2006` <dbl> 99405, 475606210, 25424094, 325663158, 20015279, 2992...
## $ `2007` <dbl> 100150, 488580707, 25909852, 334984176, 20778561, 297...
## $ `2008` <dbl> 100917, 502070763, 26482622, 344586109, 21578655, 294...
## $ `2009` <dbl> 101604, 516003448, 27466101, 354343844, 22414773, 292...
```

```
## $ `2010` <dbl> 101838, 530308387, 28284089, 364358270, 23294825, 291...
## $ `2011` <dbl> 102591, 544737983, 29347708, 374790143, 24218352, 290...
## $ `2012` <dbl> 104110, 559609961, 30560034, 385360349, 25177394, 290...
## $ `2013` <dbl> 105675, 575202699, 31622704, 396030207, 26165620, 289...
## $ `2014` <dbl> 106807, 590968990, 32792523, 406992047, 27160769, 288...
## $ `2015` <dbl> 107906, 607123269, 33831764, 418127845, 28157798, 288...
## $ `2016` <dbl> 108727, 623369401, 34700612, 429454743, 29183070, 287...
## $ `2017` <dbl> 108735, 640058741, 35688935, 440882906, 30234839, 287...
## $ `2018` <dbl> 108908, 657801085, 36743039, 452195915, 31297155, 286...
## $ `2019` <dbl> 109203, 675950189, 37856121, 463365429, 32375632, 285...
## $ `2020` <dbl> 108587, 694446100, 39068979, 474569351, 33451132, 283...
## $ `2021` <dbl> 107700, 713090928, 40000412, 485920997, 34532429, 281...
## $ `2022` <dbl> 107310, 731821393, 40578842, 497387180, 35635029, 277...
## $ `2023` <dbl> 107359, 750503764, 41454761, 509398589, 36749906, 274...
```

```
summary(df_population)
```

```

## Country Name      Country Code      Indicator Name      Indicator Code
## Length:266        Length:266        Length:266        Length:266
## Class :character  Class :character  Class :character  Class :character
## Mode  :character  Mode  :character  Mode  :character  Mode  :character
##
## 
## 
## 
##          1960           1961           1962
## Min.  :2.715e+03  Min.  :2.970e+03  Min.  :3.264e+03
## 1st Qu.:5.152e+05 1st Qu.:5.255e+05 1st Qu.:5.363e+05
## Median :3.660e+06  Median :3.747e+06  Median :3.832e+06
## Mean   :1.154e+08  Mean   :1.171e+08  Mean   :1.192e+08
## 3rd Qu.:2.686e+07  3rd Qu.:2.761e+07 3rd Qu.:2.837e+07
## Max.   :3.022e+09  Max.   :3.063e+09  Max.   :3.117e+09
## NA's   :2           NA's   :2           NA's   :2
##          1963           1964           1965
## Min.  :3.584e+03  Min.  :3.922e+03  Min.  :4.282e+03
## 1st Qu.:5.476e+05 1st Qu.:5.594e+05 1st Qu.:5.676e+05
## Median :3.920e+06  Median :4.010e+06  Median :4.103e+06
## Mean   :1.219e+08  Mean   :1.246e+08  Mean   :1.273e+08
## 3rd Qu.:2.915e+07  3rd Qu.:2.995e+07 3rd Qu.:3.076e+07
## Max.   :3.184e+09  Max.   :3.251e+09  Max.   :3.319e+09
## NA's   :2           NA's   :2           NA's   :2
##          1966           1967           1968
## Min.  :4.664e+03  Min.  :5.071e+03  Min.  :5.500e+03
## 1st Qu.:5.712e+05 1st Qu.:5.780e+05 1st Qu.:5.825e+05
## Median :4.199e+06  Median :4.298e+06  Median :4.396e+06
## Mean   :1.302e+08  Mean   :1.330e+08  Mean   :1.359e+08
## 3rd Qu.:3.148e+07  3rd Qu.:3.204e+07 3rd Qu.:3.247e+07
## Max.   :3.389e+09  Max.   :3.459e+09  Max.   :3.531e+09
## NA's   :2           NA's   :2           NA's   :2
##          1969           1970           1971
## Min.  :5.631e+03  Min.  :5.663e+03  Min.  :5.685e+03
## 1st Qu.:5.861e+05 1st Qu.:5.945e+05 1st Qu.:6.064e+05
## Median :4.503e+06  Median :4.564e+06  Median :4.607e+06
## Mean   :1.390e+08  Mean   :1.421e+08  Mean   :1.452e+08
## 3rd Qu.:3.277e+07  3rd Qu.:3.295e+07 3rd Qu.:3.322e+07
## Max.   :3.605e+09  Max.   :3.681e+09  Max.   :3.759e+09
## NA's   :2           NA's   :2           NA's   :2
##          1972           1973           1974
## Min.  :5.692e+03  Min.  :5.681e+03  Min.  :5.822e+03
## 1st Qu.:6.220e+05 1st Qu.:6.478e+05 1st Qu.:6.594e+05
## Median :4.697e+06  Median :4.834e+06  Median :4.958e+06
## Mean   :1.483e+08  Mean   :1.515e+08  Mean   :1.546e+08
## 3rd Qu.:3.378e+07  3rd Qu.:3.432e+07 3rd Qu.:3.486e+07
## Max.   :3.834e+09  Max.   :3.911e+09  Max.   :3.987e+09
## NA's   :2           NA's   :2           NA's   :2
##          1975           1976           1977
## Min.  :6.117e+03  Min.  :6.404e+03  Min.  :6.686e+03
## 1st Qu.:6.615e+05 1st Qu.:6.900e+05 1st Qu.:7.339e+05
## Median :5.053e+06  Median :5.162e+06  Median :5.301e+06
## Mean   :1.577e+08  Mean   :1.607e+08  Mean   :1.637e+08
## 3rd Qu.:3.540e+07  3rd Qu.:3.592e+07 3rd Qu.:3.644e+07
## Max.   :4.062e+09  Max.   :4.136e+09  Max.   :4.209e+09
## NA's   :2           NA's   :2           NA's   :2

```

```

##          1978          1979          1980
## Min. :6.957e+03 Min. :7.184e+03 Min. :7.366e+03
## 1st Qu.:7.849e+05 1st Qu.:8.074e+05 1st Qu.:8.156e+05
## Median :5.392e+06 Median :5.549e+06 Median :5.744e+06
## Mean   :1.668e+08 Mean   :1.700e+08 Mean   :1.732e+08
## 3rd Qu.:3.701e+07 3rd Qu.:3.776e+07 3rd Qu.:3.855e+07
## Max.  :4.283e+09 Max.  :4.360e+09 Max.  :4.438e+09
## NA's   :2          NA's :2          NA's :2
##          1981          1982          1983
## Min. :7.526e+03 Min. :7.664e+03 Min. :7.789e+03
## 1st Qu.:8.259e+05 1st Qu.:8.392e+05 1st Qu.:8.549e+05
## Median :5.878e+06 Median :6.005e+06 Median :6.148e+06
## Mean   :1.765e+08 Mean   :1.800e+08 Mean   :1.835e+08
## 3rd Qu.:3.954e+07 3rd Qu.:4.056e+07 3rd Qu.:4.144e+07
## Max.  :4.517e+09 Max.  :4.600e+09 Max.  :4.683e+09
## NA's   :2          NA's :2          NA's :2
##          1984          1985          1986
## Min. :7.902e+03 Min. :8.046e+03 Min. :8.217e+03
## 1st Qu.:8.710e+05 1st Qu.:8.880e+05 1st Qu.:9.076e+05
## Median :6.285e+06 Median :6.423e+06 Median :6.542e+06
## Mean   :1.869e+08 Mean   :1.905e+08 Mean   :1.942e+08
## 3rd Qu.:4.225e+07 3rd Qu.:4.300e+07 3rd Qu.:4.370e+07
## Max.  :4.766e+09 Max.  :4.851e+09 Max.  :4.938e+09
## NA's   :2          NA's :2          NA's :2
##          1987          1988          1989
## Min. :8.371e+03 Min. :8.518e+03 Min. :8.662e+03
## 1st Qu.:9.300e+05 1st Qu.:9.631e+05 1st Qu.:9.980e+05
## Median :6.702e+06 Median :6.846e+06 Median :7.038e+06
## Mean   :1.979e+08 Mean   :2.017e+08 Mean   :2.055e+08
## 3rd Qu.:4.472e+07 3rd Qu.:4.604e+07 3rd Qu.:4.739e+07
## Max.  :5.027e+09 Max.  :5.117e+09 Max.  :5.208e+09
## NA's   :2          NA's :2          NA's :2
##          1990          1991          1992
## Min. :8.798e+03 Min. :8.928e+03 Min. :9.038e+03
## 1st Qu.:1.055e+06 1st Qu.:1.070e+06 1st Qu.:1.084e+06
## Median :7.130e+06 Median :7.271e+06 Median :7.382e+06
## Mean   :2.085e+08 Mean   :2.123e+08 Mean   :2.160e+08
## 3rd Qu.:4.761e+07 3rd Qu.:4.972e+07 3rd Qu.:5.185e+07
## Max.  :5.299e+09 Max.  :5.388e+09 Max.  :5.477e+09
## NA's   :1          NA's :1          NA's :1
##          1993          1994          1995
## Min. :9.126e+03 Min. :9.207e+03 Min. :9.280e+03
## 1st Qu.:1.097e+06 1st Qu.:1.113e+06 1st Qu.:1.123e+06
## Median :7.495e+06 Median :7.420e+06 Median :7.563e+06
## Mean   :2.197e+08 Mean   :2.233e+08 Mean   :2.269e+08
## 3rd Qu.:5.235e+07 3rd Qu.:5.208e+07 3rd Qu.:5.167e+07
## Max.  :5.565e+09 Max.  :5.651e+09 Max.  :5.737e+09
## NA's   :1          NA's :1          NA's :1
##          1996          1997          1998
## Min. :9.342e+03 Min. :9.400e+03 Min. :9.451e+03
## 1st Qu.:1.152e+06 1st Qu.:1.181e+06 1st Qu.:1.212e+06
## Median :7.708e+06 Median :7.838e+06 Median :8.002e+06
## Mean   :2.306e+08 Mean   :2.342e+08 Mean   :2.378e+08
## 3rd Qu.:5.123e+07 3rd Qu.:5.079e+07 3rd Qu.:5.038e+07
## Max.  :5.823e+09 Max.  :5.908e+09 Max.  :5.994e+09
## NA's   :1          NA's :1          NA's :1

```

| | 1999 | 2000 | 2001 |
|----|-------------------|-------------------|-------------------|
| ## | Min. :9.496e+03 | Min. :9.544e+03 | Min. :9.586e+03 |
| ## | 1st Qu.:1.243e+06 | 1st Qu.:1.276e+06 | 1st Qu.:1.310e+06 |
| ## | Median :8.151e+06 | Median :8.214e+06 | Median :8.287e+06 |
| ## | Mean :2.414e+08 | Mean :2.450e+08 | Mean :2.485e+08 |
| ## | 3rd Qu.:4.998e+07 | 3rd Qu.:5.051e+07 | 3rd Qu.:5.213e+07 |
| ## | Max. :6.078e+09 | Max. :6.162e+09 | Max. :6.245e+09 |
| ## | NA's :1 | NA's :1 | NA's :1 |
| ## | 2002 | 2003 | 2004 |
| ## | Min. :9.623e+03 | Min. :9.695e+03 | Min. :9.816e+03 |
| ## | 1st Qu.:1.332e+06 | 1st Qu.:1.339e+06 | 1st Qu.:1.350e+06 |
| ## | Median :8.400e+06 | Median :8.587e+06 | Median :8.817e+06 |
| ## | Mean :2.521e+08 | Mean :2.556e+08 | Mean :2.592e+08 |
| ## | 3rd Qu.:5.375e+07 | 3rd Qu.:5.534e+07 | 3rd Qu.:5.700e+07 |
| ## | Max. :6.328e+09 | Max. :6.410e+09 | Max. :6.493e+09 |
| ## | NA's :1 | NA's :1 | NA's :1 |
| ## | 2005 | 2006 | 2007 |
| ## | Min. :9.940e+03 | Min. :1.003e+04 | Min. :1.002e+04 |
| ## | 1st Qu.:1.355e+06 | 1st Qu.:1.347e+06 | 1st Qu.:1.341e+06 |
| ## | Median :9.030e+06 | Median :9.081e+06 | Median :9.148e+06 |
| ## | Mean :2.628e+08 | Mean :2.664e+08 | Mean :2.700e+08 |
| ## | 3rd Qu.:5.797e+07 | 3rd Qu.:5.814e+07 | 3rd Qu.:5.844e+07 |
| ## | Max. :6.576e+09 | Max. :6.660e+09 | Max. :6.744e+09 |
| ## | NA's :1 | NA's :1 | NA's :1 |
| ## | 2008 | 2009 | 2010 |
| ## | Min. :1.001e+04 | Min. :1.002e+04 | Min. :1.004e+04 |
| ## | 1st Qu.:1.427e+06 | 1st Qu.:1.526e+06 | 1st Qu.:1.566e+06 |
| ## | Median :9.228e+06 | Median :9.505e+06 | Median :9.746e+06 |
| ## | Mean :2.737e+08 | Mean :2.774e+08 | Mean :2.812e+08 |
| ## | 3rd Qu.:5.883e+07 | 3rd Qu.:5.910e+07 | 3rd Qu.:5.928e+07 |
| ## | Max. :6.830e+09 | Max. :6.916e+09 | Max. :7.001e+09 |
| ## | NA's :1 | NA's :1 | NA's :1 |
| ## | 2011 | 2012 | 2013 |
| ## | Min. :1.010e+04 | Min. :1.027e+04 | Min. :1.052e+04 |
| ## | 1st Qu.:1.608e+06 | 1st Qu.:1.651e+06 | 1st Qu.:1.696e+06 |
| ## | Median :9.915e+06 | Median :1.007e+07 | Median :1.019e+07 |
| ## | Mean :2.849e+08 | Mean :2.888e+08 | Mean :2.927e+08 |
| ## | 3rd Qu.:5.938e+07 | 3rd Qu.:5.954e+07 | 3rd Qu.:6.023e+07 |
| ## | Max. :7.086e+09 | Max. :7.176e+09 | Max. :7.265e+09 |
| ## | NA's :1 | NA's :1 | NA's :1 |
| ## | 2014 | 2015 | 2016 |
| ## | Min. :1.074e+04 | Min. :1.095e+04 | Min. :1.093e+04 |
| ## | 1st Qu.:1.741e+06 | 1st Qu.:1.786e+06 | 1st Qu.:1.778e+06 |
| ## | Median :1.032e+07 | Median :1.036e+07 | Median :1.033e+07 |
| ## | Mean :2.966e+08 | Mean :3.005e+08 | Mean :3.043e+08 |
| ## | 3rd Qu.:6.079e+07 | 3rd Qu.:6.073e+07 | 3rd Qu.:6.063e+07 |
| ## | Max. :7.354e+09 | Max. :7.441e+09 | Max. :7.529e+09 |
| ## | NA's :1 | NA's :1 | NA's :1 |
| ## | 2017 | 2018 | 2019 |
| ## | Min. :1.087e+04 | Min. :1.075e+04 | Min. :1.058e+04 |
| ## | 1st Qu.:1.791e+06 | 1st Qu.:1.797e+06 | 1st Qu.:1.789e+06 |
| ## | Median :1.026e+07 | Median :1.028e+07 | Median :1.042e+07 |
| ## | Mean :3.082e+08 | Mean :3.119e+08 | Mean :3.156e+08 |
| ## | 3rd Qu.:6.054e+07 | 3rd Qu.:6.042e+07 | 3rd Qu.:5.973e+07 |
| ## | Max. :7.614e+09 | Max. :7.696e+09 | Max. :7.777e+09 |
| ## | NA's :1 | NA's :1 | NA's :1 |

```

##          2020           2021           2022
## Min. :1.040e+04  Min. :1.019e+04  Min. :9.992e+03
## 1st Qu.:1.790e+06 1st Qu.:1.786e+06 1st Qu.:1.804e+06
## Median :1.070e+07 Median :1.051e+07 Median :1.049e+07
## Mean   :3.192e+08 Mean   :3.224e+08 Mean   :3.255e+08
## 3rd Qu.:6.097e+07 3rd Qu.:6.283e+07 3rd Qu.:6.471e+07
## Max.   :7.856e+09 Max.   :7.921e+09 Max.   :7.990e+09
## NA's    :1          NA's    :1          NA's    :1
##          2023
## Min. :9.816e+03
## 1st Qu.:1.828e+06
## Median :1.064e+07
## Mean   :3.288e+08
## 3rd Qu.:6.662e+07
## Max.   :8.062e+09
## NA's   :1

```

- What the summary tells us
 - The dataset has 266 rows and 68 columns
- Key identifier variables:
 - Country Name – character
 - Country Code – character
 - Indicator Name – character
 - Indicator Code – character
- Year columns (1960 to 2023) are correctly read as numeric ()
- Some year columns contain NA values (usually for aggregates like regions)

Rename columns for consistency

```

population2 <- df_population %>%
  rename(
    country = `Country Name`,
    country_code = `Country Code`
  ) %>%
  select(-`Indicator Name`, -`Indicator Code`)

```

Remove unnecessary indicator columns

```
population2
```

```

## # A tibble: 266 × 66
##   country country_code `1960` `1961` `1962` `1963` `1964` `1965` `1966` `1967`
##   <chr>    <chr>      <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>   <dbl>
## 1 Aruba     ABW      5.49e4  5.56e4  5.63e4  5.70e4  5.76e4  5.82e4  5.87e4  5.90e4
## 2 Africa ... AFE      1.30e8  1.34e8  1.37e8  1.41e8  1.45e8  1.49e8  1.53e8  1.58e8
## 3 Afghani... AFG      9.04e6  9.21e6  9.40e6  9.60e6  9.81e6  1.00e7  1.03e7  1.05e7
## 4 Africa ... AFW      9.76e7  9.97e7  1.02e8  1.04e8  1.06e8  1.09e8  1.11e8  1.14e8
## 5 Angola    AGO      5.23e6  5.30e6  5.35e6  5.41e6  5.46e6  5.52e6  5.58e6  5.64e6
## 6 Albania   ALB      1.61e6  1.66e6  1.71e6  1.76e6  1.81e6  1.86e6  1.91e6  1.97e6
## 7 Andorra   AND      9.51e3  1.03e4  1.11e4  1.19e4  1.28e4  1.36e4  1.46e4  1.58e4
## 8 Arab Wo... ARB      9.15e7  9.39e7  9.64e7  9.90e7  1.02e8  1.04e8  1.07e8  1.10e8
## 9 United ... ARE      1.31e5  1.38e5  1.45e5  1.52e5  1.60e5  1.67e5  1.74e5  1.80e5
## 10 Argenti... ARG     2.04e7  2.07e7  2.11e7  2.14e7  2.18e7  2.21e7  2.25e7  2.28e7
## # i 256 more rows
## # i 56 more variables: `1968` <dbl>, `1969` <dbl>, `1970` <dbl>, `1971` <dbl>,
## #   `1972` <dbl>, `1973` <dbl>, `1974` <dbl>, `1975` <dbl>, `1976` <dbl>,
## #   `1977` <dbl>, `1978` <dbl>, `1979` <dbl>, `1980` <dbl>, `1981` <dbl>,
## #   `1982` <dbl>, `1983` <dbl>, `1984` <dbl>, `1985` <dbl>, `1986` <dbl>,
## #   `1987` <dbl>, `1988` <dbl>, `1989` <dbl>, `1990` <dbl>, `1991` <dbl>,
## #   `1992` <dbl>, `1993` <dbl>, `1994` <dbl>, `1995` <dbl>, `1996` <dbl>, ...

```

The summary shows that the population dataset contains 266 rows and 68 columns. The country identifiers are stored as character variables, while yearly population values from 1960 to 2023 are stored as numeric variables. However, the dataset is in wide format, with each year represented as a separate column, and it includes regional aggregates rather than only individual countries. Additionally, some population values are missing. These anomalies were addressed by renaming variables for consistency and removing unnecessary indicator columns, storing the corrected dataset as `population2`.

3. Tidy Data

3.1 who Dataset

Description of Columns:

- **country** : Country name.
- **iso2** : Two-digit country code.
- **iso3** : Three-digit country code.
- **year** : Year.
- Variables like `new_ep_f014` :
 - `ep` : TB type (e.g., `rel` = relapse, `ep` = extrapulmonary).
 - `f` : Sex (e.g., `f` = female).
 - `014` : Age group (e.g., 0-14 years).

Steps:

3.1.1 Identify the variables in the dataset.

```
colnames(df_who)
```

```

## [1] "country"      "iso2"        "iso3"        "year"        "new_sp_m014"
## [6] "new_sp_m1524" "new_sp_m2534" "new_sp_m3544" "new_sp_m4554" "new_sp_m5564"
## [11] "new_sp_m65"   "new_sp_f014"   "new_sp_f1524" "new_sp_f2534" "new_sp_f3544"
## [16] "new_sp_f4554" "new_sp_f5564" "new_sp_f65"   "new_sn_m014"  "new_sn_m1524"
## [21] "new_sn_m2534" "new_sn_m3544" "new_sn_m4554" "new_sn_m5564" "new_sn_m65"
## [26] "new_sn_f014"  "new_sn_f1524" "new_sn_f2534" "new_sn_f3544" "new_sn_f4554"
## [31] "new_sn_f5564" "new_sn_f65"   "new_ep_m014"  "new_ep_m1524" "new_ep_m2534"
## [36] "new_ep_m3544" "new_ep_m4554" "new_ep_m5564" "new_ep_m65"   "new_ep_f014"
## [41] "new_ep_f1524" "new_ep_f2534" "new_ep_f3544" "new_ep_f4554" "new_ep_f5564"
## [46] "new_ep_f65"   "newrel_m014"  "newrel_m1524" "newrel_m2534" "newrel_m3544"
## [51] "newrel_m4554" "newrel_m5564" "newrel_m65"   "newrel_f014"  "newrel_f1524"
## [56] "newrel_f2534" "newrel_f3544" "newrel_f4554" "newrel_f5564" "newrel_f65"

```

3.1.2 Perform a pivot operation to make the data tidy, storing the result in `who2`.

```

who2 <- who %>%
  pivot_longer(
    cols = starts_with("new"),
    names_to = "key",
    values_to = "cases"
  )

```

Remove NA values

```

who2 <- who2 %>%
  filter(!is.na(cases))

```

3.1.3 Separate values like `new_ep_f014` into components (e.g., `new`, `ep`, `f014`). Remove the column containing `new`, and store the result in `who3`.

```

who3 <- who2 %>%
  separate(key, into = c("new", "type", "sex_age"), sep = "_") %>%
  select(-new)

```

```

## Warning: Expected 3 pieces. Missing pieces filled with `NA` in 2580 rows [243, 244, 679,
## 680, 681, 682, 683, 684, 685, 686, 687, 688, 689, 690, 691, 692, 903, 904, 905,
## 906, ...].

```

3.1.4 Further separate values like `f014` into `f` and `014`, storing the result in `who_tidy`.

```

who_tidy <- who3 %>%
  separate(sex_age, into = c("sex", "age"), sep = 1)

```

3.2 Population Dataset

3.2.1 Identify the variables in this dataset.

```
colnames(population2)
```

```
## [1] "country"      "country_code"   "1960"        "1961"        "1962"
## [6] "1963"          "1964"          "1965"        "1966"        "1967"
## [11] "1968"          "1969"          "1970"        "1971"        "1972"
## [16] "1973"          "1974"          "1975"        "1976"        "1977"
## [21] "1978"          "1979"          "1980"        "1981"        "1982"
## [26] "1983"          "1984"          "1985"        "1986"        "1987"
## [31] "1988"          "1989"          "1990"        "1991"        "1992"
## [36] "1993"          "1994"          "1995"        "1996"        "1997"
## [41] "1998"          "1999"          "2000"        "2001"        "2002"
## [46] "2003"          "2004"          "2005"        "2006"        "2007"
## [51] "2008"          "2009"          "2010"        "2011"        "2012"
## [56] "2013"          "2014"          "2015"        "2016"        "2017"
## [61] "2018"          "2019"          "2020"        "2021"        "2022"
## [66] "2023"
```

3.2.2 Perform a pivot operation to tidy the data, storing the result in `population3`.

```
population3 <- population2 %>%
  pivot_longer(
    cols = matches("^[0-9]{4}$"),
    names_to = "year",
    values_to = "population"
  )
```

3.2.3 Cast the population variable to an appropriate data type, storing the result in `population_tidy`.

```
population_tidy <- population3 %>%
  mutate(
    year = as.integer(year),
    population = as.numeric(population)
  )
```

3.3 Join Datasets

3.3.1 Identify the variable(s) required to join `who_tidy` and `population_tidy`.

```
colnames(who_tidy)
```

```
## [1] "country" "iso2"     "iso3"     "year"     "type"     "sex"      "age"
## [8] "cases"
```

```
colnames(population_tidy)
```

```
## [1] "country"      "country_code"   "year"        "population"
```

3.3.2 Rename columns as needed to align variable names between datasets.

```
population_tidy <- population_tidy %>%
  rename(country = country)
```

3.3.3 Join the datasets into a tibble called `tuberculosis`.

```
tuberculosis <- who_tidy %>%  
  left_join(population_tidy, by = c("country", "year"))
```

3.4 Clean Up Data

3.4.1 Remove unnecessary variables from `tuberculosis`.

```
tuberculosis <- tuberculosis %>%  
  select(country, year, type, sex, age, cases, population)
```

3.4.2 Filter out rows with `NA` values.

3.4.3 Save the cleaned data back into `tuberculosis`.

```
tuberculosis <- tuberculosis %>%  
  select(country, year, type, sex, age, cases, population) %>%  
  drop_na()
```

4. Data Manipulation

4.1 Determine the total TB cases among men and women in the 21st century in the United States. Identify which sex had more cases.

```
us_cases <- tuberculosis %>%  
  filter(  
    country == "United States of America",  
    year >= 2000  
  ) %>%  
  group_by(sex) %>%  
  summarise(total_cases = sum(cases))
```

```
us_cases
```

```
## # A tibble: 0 × 2  
## # i 2 variables: sex <chr>, total_cases <dbl>
```

4.2 Create a new variable, `cases_per_100k`, representing TB cases per 100,000 people by year, sex, age group, and TB type.

```
tuberculosis <- tuberculosis %>%  
  mutate(  
    cases_per_100k = (cases / population) * 100000  
  )  
tuberculosis
```

```

## # A tibble: 61,282 × 8
##   country      year type sex   age   cases population cases_per_100k
##   <chr>        <dbl> <chr> <chr> <dbl>     <dbl>          <dbl>
## 1 Afghanistan 1997 sp    m    014      0  18452091      0
## 2 Afghanistan 1997 sp    m    1524     10  18452091    0.0542
## 3 Afghanistan 1997 sp    m    2534      6  18452091    0.0325
## 4 Afghanistan 1997 sp    m    3544      3  18452091    0.0163
## 5 Afghanistan 1997 sp    m    4554      5  18452091    0.0271
## 6 Afghanistan 1997 sp    m    5564      2  18452091    0.0108
## 7 Afghanistan 1997 sp    m    65       0  18452091      0
## 8 Afghanistan 1997 sp    f    014      5  18452091    0.0271
## 9 Afghanistan 1997 sp    f    1524     38  18452091    0.206
## 10 Afghanistan 1997 sp   f    2534     36  18452091    0.195
## # i 61,272 more rows

```

4.3 Identify:

- The country and year with the highest cases per 100k.
 - The country and year with the lowest cases per 100k.

```

highest_cases <- tuberculosis %>%
  arrange(desc(cases_per_100k)) %>%
  slice(1)

```

highest_cases

```

## # A tibble: 1 × 8
##   country      year type sex   age   cases population cases_per_100k
##   <chr>        <dbl> <chr> <chr> <dbl>     <dbl>          <dbl>
## 1 Samoa       2009 sp    f    65     1111    191513      580.

```

Lowest cases per 100k (non-zero to avoid misleading zeros)

```

lowest_cases <- tuberculosis %>%
  filter(cases_per_100k > 0) %>%
  arrange(cases_per_100k) %>%
  slice(1)

```

lowest_cases

```

## # A tibble: 1 × 8
##   country              year type sex   age   cases population cases_per_100k
##   <chr>                <dbl> <chr> <chr> <dbl>     <dbl>          <dbl>
## 1 Russian Federation 2000 sp    m    014      1  146596869    0.000682

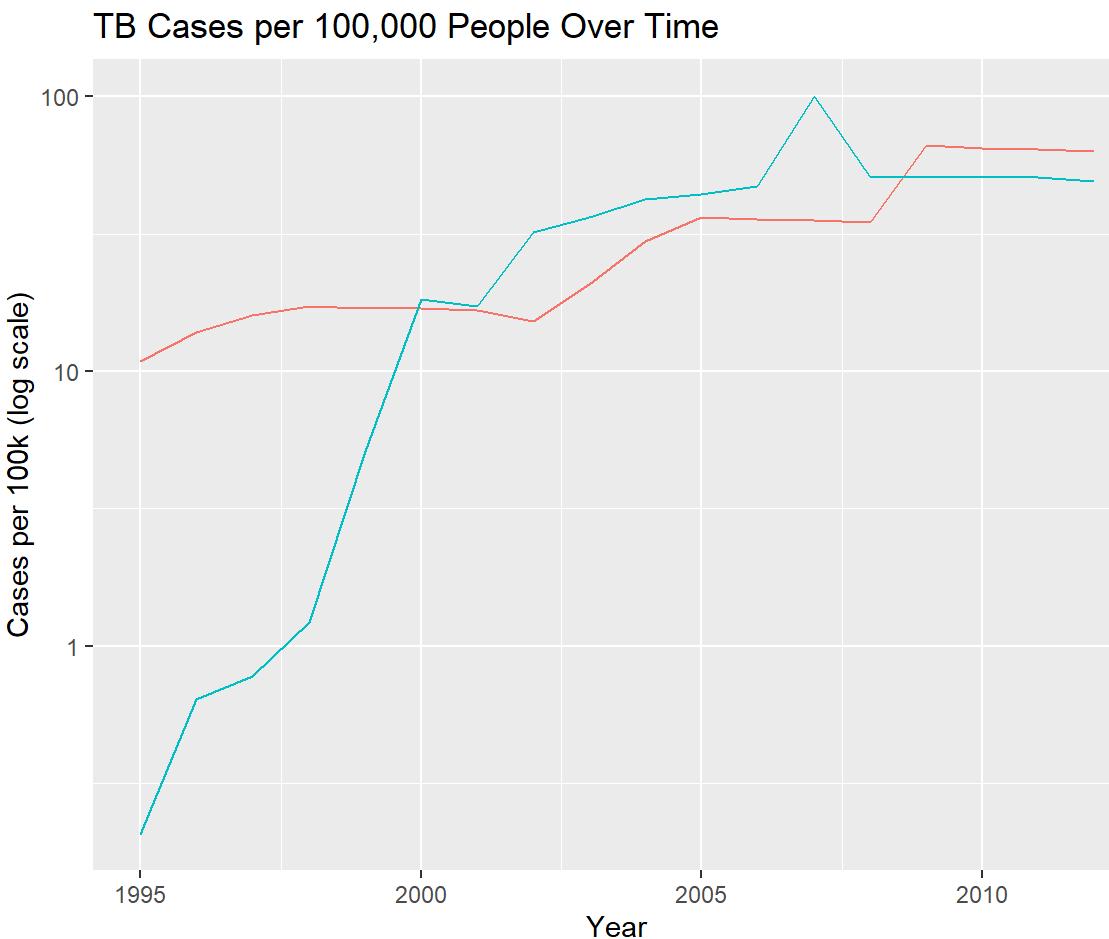
```

5. Data Visualization

5.1 Plot the total cases per 100k as a function of year for China, India, and the United States:

- Use a log scale on the y-axis (`scale_y_log10()`).
 - Describe emerging patterns.

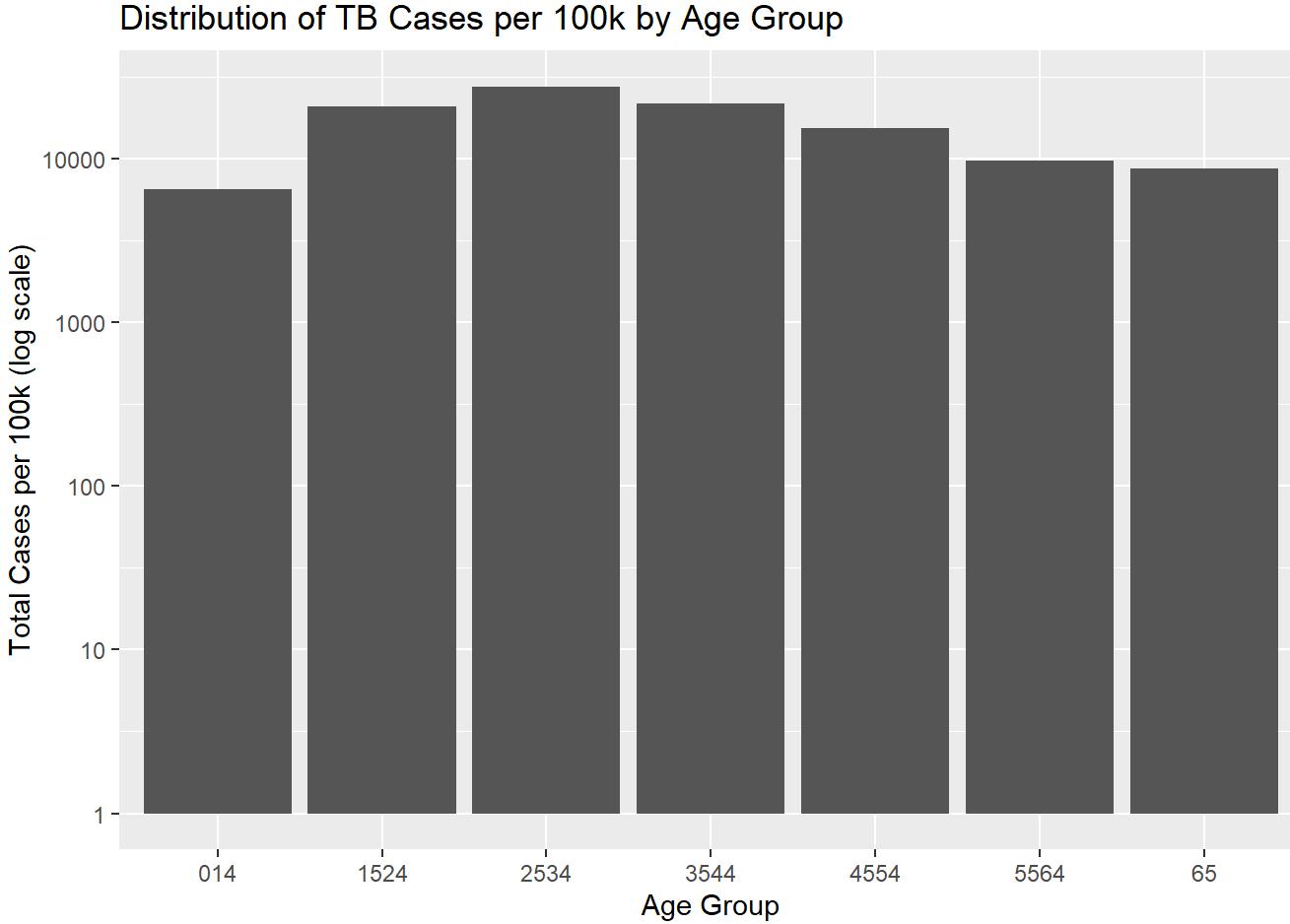
```
tuberculosis %>%
  filter(country %in% c("China", "India", "United States of America")) %>%
  group_by(country, year) %>%
  summarise(total_cases_100k = sum(cases_per_100k), .groups = "drop") %>%
  ggplot(aes(x = year, y = total_cases_100k, color = country)) +
  geom_line() +
  scale_y_log10() +
  labs(
    title = "TB Cases per 100,000 People Over Time",
    y = "Cases per 100k (log scale)",
    x = "Year"
  )
)
```



5.2 Compare distributions of total cases per 100k (summed over years, sexes, and TB types) across age groups:

- - Use a log scale on the y-axis.
 - Highlight observed patterns.

```
tuberculosis %>%
  group_by(age) %>%
  summarise(total_cases_100k = sum(cases_per_100k), .groups = "drop") %>%
  ggplot(aes(x = age, y = total_cases_100k)) +
  geom_col() +
  scale_y_log10() +
  labs(
    title = "Distribution of TB Cases per 100k by Age Group",
    x = "Age Group",
    y = "Total Cases per 100k (log scale)"
  )
```



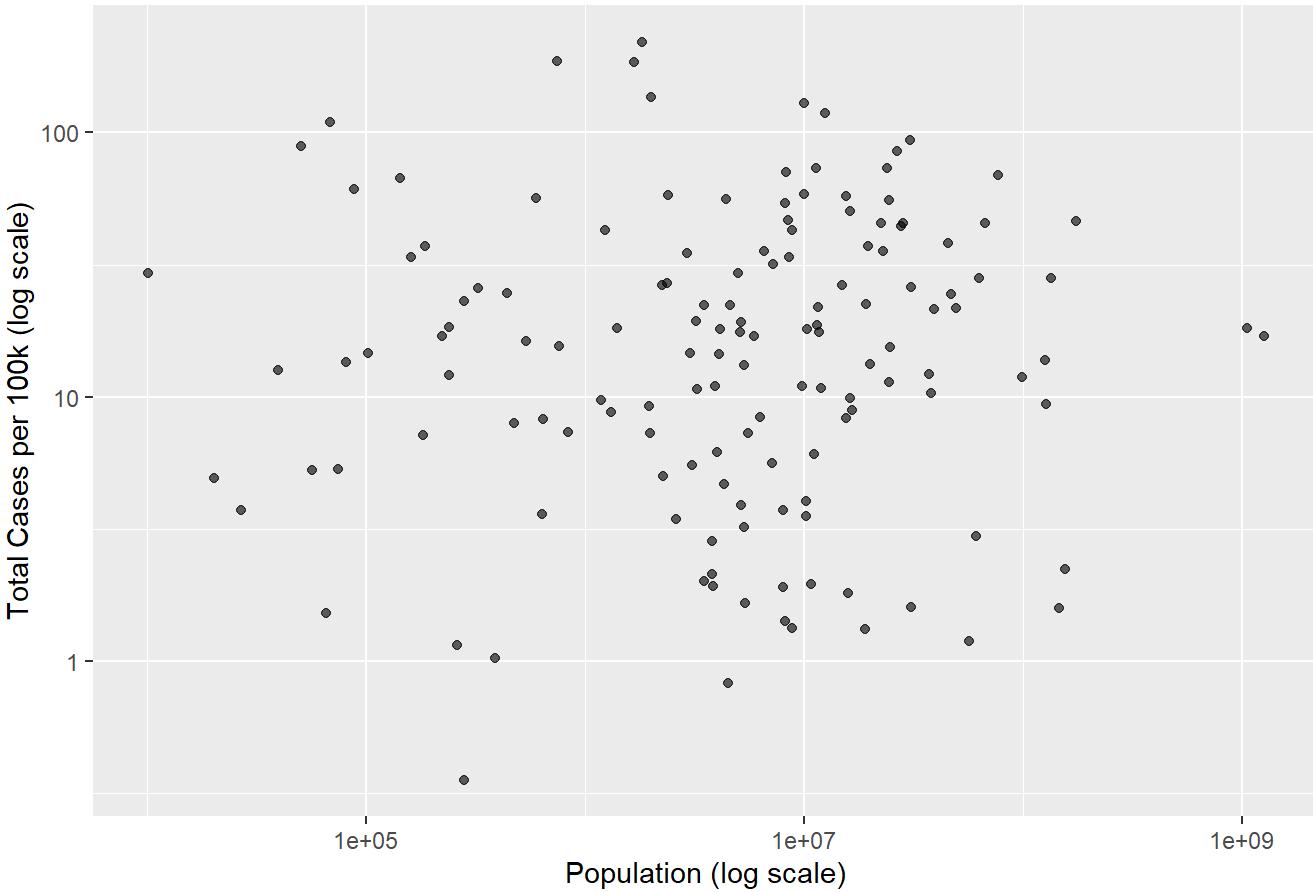
5.3 Create a plot to evaluate whether the number of cases per 100k in 2000 was related to a country's population:

- ○ Conclude based on the visualization.

```
tuberculosis %>%
  filter(year == 2000) %>%
  group_by(country) %>%
  summarise(
    total_cases_100k = sum(cases_per_100k),
    population = mean(population),
    .groups = "drop"
  ) %>%
  ggplot(aes(x = population, y = total_cases_100k)) +
  geom_point(alpha = 0.6) +
  scale_x_log10() +
  scale_y_log10() +
  labs(
    title = "TB Cases per 100k vs Population Size (Year 2000)",
    x = "Population (log scale)",
    y = "Total Cases per 100k (log scale)"
  )

```

TB Cases per 100k vs Population Size (Year 2000)



The visualization suggests no strong direct relationship between population size and TB cases per 100,000. Countries with large populations do not necessarily experience higher TB rates, indicating that TB burden is more closely linked to public health conditions than population size alone.