

John Hopkins Covid-19 Data Project

DM

2024-10-12

Load Data

The data comes from John Hopkins from the COVID-19 pandemic and was archived on Mar 10, 2023.

```
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.2      v readr      2.1.5
## v forcats    1.0.0      v stringr   1.5.0
## v ggplot2     3.5.0      v tibble    3.2.1
## v lubridate  1.9.3      v tidyr     1.3.0
## v purrr      1.0.1
## -- 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 errors
```

```
library(lubridate)
url_in <- "https://raw.githubusercontent.com/CSSEGISandData/COVID-19/master/csse_covid_19_data/csse_covid_19_data.csv"

file_names <-c("time_series_covid19_confirmed_global.csv",
               "time_series_covid19_deaths_global.csv",
               "time_series_covid19_confirmed_US.csv",
               "time_series_covid19_deaths_US.csv")

urls <- str_c(url_in, file_names)
urls
```

```
## [1] "https://raw.githubusercontent.com/CSSEGISandData/COVID-19/master/csse_covid_19_data/csse_covid_19_data.csv"
## [2] "https://raw.githubusercontent.com/CSSEGISandData/COVID-19/master/csse_covid_19_data/csse_covid_19_data.csv"
## [3] "https://raw.githubusercontent.com/CSSEGISandData/COVID-19/master/csse_covid_19_data/csse_covid_19_data.csv"
## [4] "https://raw.githubusercontent.com/CSSEGISandData/COVID-19/master/csse_covid_19_data/csse_covid_19_data.csv"
```

```
global_cases <- read_csv(urls[1])
```

```
## Rows: 289 Columns: 1147
## -- Column specification -----
## Delimiter: ","
## chr      (2): Province/State, Country/Region
```

```
## dbl (1145): Lat, Long, 1/22/20, 1/23/20, 1/24/20, 1/25/20, 1/26/20, 1/27/20,...
##
## 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.
```

```
global_deaths <- read_csv(urls[2])
```

```
## Rows: 289 Columns: 1147
## -- Column specification -----
## Delimiter: ","
## chr (2): Province/State, Country/Region
## dbl (1145): Lat, Long, 1/22/20, 1/23/20, 1/24/20, 1/25/20, 1/26/20, 1/27/20,...
##
## 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.
```

```
US_cases <- read_csv(urls[3])
```

```
## Rows: 3342 Columns: 1154
## -- Column specification -----
## Delimiter: ","
## chr (6): iso2, iso3, Admin2, Province_State, Country_Region, Combined_Key
## dbl (1148): UID, code3, FIPS, Lat, Long_, 1/22/20, 1/23/20, 1/24/20, 1/25/20...
##
## 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.
```

```
US_deaths <- read_csv(urls[4])
```

```
## Rows: 3342 Columns: 1155
## -- Column specification -----
## Delimiter: ","
## chr (6): iso2, iso3, Admin2, Province_State, Country_Region, Combined_Key
## dbl (1149): UID, code3, FIPS, Lat, Long_, Population, 1/22/20, 1/23/20, 1/24...
##
## 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.
```

Cleaning and Transforming Data

```
#tidy data

global_cases <- global_cases %>%
  pivot_longer(cols = -c('Province/State', 'Country/Region', Lat, Long),
               names_to = "date",
               values_to = "cases") %>%
  select(-c(Lat, Long))

head(global_cases, n=10)
```

```
## # A tibble: 10 x 4
##   'Province/State' 'Country/Region' date    cases
##   <chr>            <chr>          <chr>  <dbl>
## 1 <NA>            Afghanistan  1/22/20    0
## 2 <NA>            Afghanistan  1/23/20    0
## 3 <NA>            Afghanistan  1/24/20    0
## 4 <NA>            Afghanistan  1/25/20    0
## 5 <NA>            Afghanistan  1/26/20    0
## 6 <NA>            Afghanistan  1/27/20    0
## 7 <NA>            Afghanistan  1/28/20    0
## 8 <NA>            Afghanistan  1/29/20    0
## 9 <NA>            Afghanistan  1/30/20    0
## 10 <NA>           Afghanistan  1/31/20    0
```

```
global_deaths <- global_deaths %>%
  pivot_longer(cols = -c('Province/State', 'Country/Region', Lat, Long),
    names_to = "date",
    values_to = "deaths") %>%
  select(-c(Lat, Long))

head(global_deaths, n=10)
```

```
## # A tibble: 10 x 4
##   'Province/State' 'Country/Region' date    deaths
##   <chr>            <chr>          <chr>  <dbl>
## 1 <NA>            Afghanistan  1/22/20    0
## 2 <NA>            Afghanistan  1/23/20    0
## 3 <NA>            Afghanistan  1/24/20    0
## 4 <NA>            Afghanistan  1/25/20    0
## 5 <NA>            Afghanistan  1/26/20    0
## 6 <NA>            Afghanistan  1/27/20    0
## 7 <NA>            Afghanistan  1/28/20    0
## 8 <NA>            Afghanistan  1/29/20    0
## 9 <NA>            Afghanistan  1/30/20    0
## 10 <NA>           Afghanistan  1/31/20    0
```

```
#transform data
#combine the cases into deaths per date into one variable called global
global <- global_cases %>%
  full_join(global_deaths) %>%
  rename(`Country_Region`='Country/Region',
    Province_State = 'Province/State') %>%
  mutate(date = mdy(date))
```

```
## Joining with 'by = join_by('Province/State', 'Country/Region', date)'
```

```
global <- global %>% filter(cases >0)
summary(global)
```

```
## Province_State    Country_Region      date      cases
## Length:306827     Length:306827      Min.   :2020-01-22  Min.   :      1
## Class :character  Class :character   1st Qu.:2020-12-12  1st Qu.:    1316
```

```
## Mode :character Mode :character Median :2021-09-16 Median : 20365
## Mean :2021-09-11 Mean : 1032863
## 3rd Qu.:2022-06-15 3rd Qu.: 271281
## Max. :2023-03-09 Max. :103802702
## deaths
## Min. : 0
## 1st Qu.: 7
## Median : 214
## Mean : 14405
## 3rd Qu.: 3665
## Max. :1123836
```

```
#double check amount of cases close to maximum
global %>% filter(cases > 103000000)
```

```
## # A tibble: 23 x 5
## Province_State Country_Region date cases deaths
## <chr> <chr> <date> <dbl> <dbl>
## 1 <NA> US 2023-02-15 103023231 1115741
## 2 <NA> US 2023-02-16 103083910 1116851
## 3 <NA> US 2023-02-17 103131898 1117572
## 4 <NA> US 2023-02-18 103134605 1117589
## 5 <NA> US 2023-02-19 103136077 1117590
## 6 <NA> US 2023-02-20 103138119 1117663
## 7 <NA> US 2023-02-21 103198669 1118025
## 8 <NA> US 2023-02-22 103308832 1118886
## 9 <NA> US 2023-02-23 103365511 1119521
## 10 <NA> US 2023-02-24 103378408 1119573
## # i 13 more rows
```

```
#now clean and wrangle US cases
US_cases %>%
  pivot_longer(cols = -(UID:Combined_Key),
               names_to = "date",
               values_to = "cases")
```

```
## # A tibble: 3,819,906 x 13
## UID iso2 iso3 code3 FIPS Admin2 Province_State Country_Region Lat
## <dbl> <chr> <chr> <dbl> <dbl> <chr> <chr> <chr> <dbl>
## 1 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 2 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 3 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 4 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 5 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 6 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 7 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 8 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 9 84001001 US USA 840 1001 Autauga Alabama US 32.5
## 10 84001001 US USA 840 1001 Autauga Alabama US 32.5
## # i 3,819,896 more rows
## # i 4 more variables: Long_ <dbl>, Combined_Key <chr>, date <chr>, cases <dbl>
```

```
US_cases <- US_cases %>%
  pivot_longer(cols = -(UID:Combined_Key),
               names_to = "date",
               values_to = "cases") %>%
  select(Admin2:cases) %>%
  mutate(date=mdy(date)) %>%
  select(-c(Lat,Long_))

head(US_cases, n=10)
```

```
## # A tibble: 10 x 6
##   Admin2 Province_State Country_Region Combined_Key      date      cases
##   <chr>    <chr>          <chr>          <chr>      <date>    <dbl>
## 1 Autauga Alabama        US      Autauga, Alabama, US 2020-01-22      0
## 2 Autauga Alabama        US      Autauga, Alabama, US 2020-01-23      0
## 3 Autauga Alabama        US      Autauga, Alabama, US 2020-01-24      0
## 4 Autauga Alabama        US      Autauga, Alabama, US 2020-01-25      0
## 5 Autauga Alabama        US      Autauga, Alabama, US 2020-01-26      0
## 6 Autauga Alabama        US      Autauga, Alabama, US 2020-01-27      0
## 7 Autauga Alabama        US      Autauga, Alabama, US 2020-01-28      0
## 8 Autauga Alabama        US      Autauga, Alabama, US 2020-01-29      0
## 9 Autauga Alabama        US      Autauga, Alabama, US 2020-01-30      0
## 10 Autauga Alabama        US      Autauga, Alabama, US 2020-01-31      0
```

```
US_deaths <- US_deaths %>%
  pivot_longer(cols = -(UID:Population),
               names_to = "date",
               values_to = "deaths") %>%
  select(Admin2:deaths) %>%
  mutate(date=mdy(date)) %>%
  select(-c(Lat,Long_))

head(US_deaths, n=10)
```

```
## # A tibble: 10 x 7
##   Admin2 Province_State Country_Region Combined_Key Population date
##   <chr>    <chr>          <chr>          <chr>      <dbl> <date>
## 1 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-22
## 2 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-23
## 3 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-24
## 4 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-25
## 5 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-26
## 6 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-27
## 7 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-28
## 8 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-29
## 9 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-30
## 10 Autauga Alabama        US      Autauga, Alabama~ 55869 2020-01-31
## # i 1 more variable: deaths <dbl>
```

```
#join two US datasets together
US <- US_cases %>%
  full_join(US_deaths)
```

```
## Joining with 'by = join_by(Admin2, Province_State, Country_Region,
## Combined_Key, date)'
```

```
#check join
head(US, n=10)
```

```
## # A tibble: 10 x 8
##   Admin2 Province_State Country_Region Combined_Key date       cases Population
##   <chr>   <chr>           <chr>         <chr>      <date>    <dbl>      <dbl>
## 1 Autau~ Alabama        US           Autauga, Al~ 2020-01-22    0      55869
## 2 Autau~ Alabama        US           Autauga, Al~ 2020-01-23    0      55869
## 3 Autau~ Alabama        US           Autauga, Al~ 2020-01-24    0      55869
## 4 Autau~ Alabama        US           Autauga, Al~ 2020-01-25    0      55869
## 5 Autau~ Alabama        US           Autauga, Al~ 2020-01-26    0      55869
## 6 Autau~ Alabama        US           Autauga, Al~ 2020-01-27    0      55869
## 7 Autau~ Alabama        US           Autauga, Al~ 2020-01-28    0      55869
## 8 Autau~ Alabama        US           Autauga, Al~ 2020-01-29    0      55869
## 9 Autau~ Alabama        US           Autauga, Al~ 2020-01-30    0      55869
## 10 Autau~ Alabama        US           Autauga, Al~ 2020-01-31    0      55869
## # i 1 more variable: deaths <dbl>
```

```
#add population data to global dataset
```

```
global <- global %>%
  unite("Combined_Key",
        c(Province_State, Country_Region),
        sep=" ",
        na.rm=TRUE,
        remove=FALSE)
head(global, n=10)
```

```
## # A tibble: 10 x 6
##   Combined_Key Province_State Country_Region date       cases deaths
##   <chr>         <chr>           <chr>      <date>    <dbl>  <dbl>
## 1 Afghanistan <NA>            Afghanistan 2020-02-24    5      0
## 2 Afghanistan <NA>            Afghanistan 2020-02-25    5      0
## 3 Afghanistan <NA>            Afghanistan 2020-02-26    5      0
## 4 Afghanistan <NA>            Afghanistan 2020-02-27    5      0
## 5 Afghanistan <NA>            Afghanistan 2020-02-28    5      0
## 6 Afghanistan <NA>            Afghanistan 2020-02-29    5      0
## 7 Afghanistan <NA>            Afghanistan 2020-03-01    5      0
## 8 Afghanistan <NA>            Afghanistan 2020-03-02    5      0
## 9 Afghanistan <NA>            Afghanistan 2020-03-03    5      0
## 10 Afghanistan <NA>            Afghanistan 2020-03-04    5      0
```

```
#retrieve lookup table URL from John Hopkins Website
```

```
uid_lookup_url <- "https://raw.githubusercontent.com/CSSEGISandData/COVID-19/master/csse_covid_19_data/"
uid <- read_csv(uid_lookup_url) %>%
  select(-c(Lat, Long_, Combined_Key, code3, iso2, iso3, Admin2))
```

```
## Rows: 4321 Columns: 12
## -- Column specification -----
## Delimiter: ","
## chr (7): iso2, iso3, FIPS, Admin2, Province_State, Country_Region, Combined_Key
## dbl (5): UID, code3, Lat, Long_, Population
##
## 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.
```

```
#join uid onto global dataset
```

```
global <- global %>%
  left_join(uid, by=c("Province_State", "Country_Region")) %>%
  select(-c(UID, FIPS)) %>%
  select(Province_State, Country_Region, date, cases, deaths, Population, Combined_Key)

head(global, n=10)
```

```
## # A tibble: 10 x 7
##   Province_State Country_Region date      cases deaths Population Combined_Key
##   <chr>          <chr>      <date>    <dbl>  <dbl>      <dbl> <chr>
## 1 <NA>          Afghanistan 2020-02-24      5      0    38928341 Afghanistan
## 2 <NA>          Afghanistan 2020-02-25      5      0    38928341 Afghanistan
## 3 <NA>          Afghanistan 2020-02-26      5      0    38928341 Afghanistan
## 4 <NA>          Afghanistan 2020-02-27      5      0    38928341 Afghanistan
## 5 <NA>          Afghanistan 2020-02-28      5      0    38928341 Afghanistan
## 6 <NA>          Afghanistan 2020-02-29      5      0    38928341 Afghanistan
## 7 <NA>          Afghanistan 2020-03-01      5      0    38928341 Afghanistan
## 8 <NA>          Afghanistan 2020-03-02      5      0    38928341 Afghanistan
## 9 <NA>          Afghanistan 2020-03-03      5      0    38928341 Afghanistan
## 10 <NA>         Afghanistan 2020-03-04      5      0    38928341 Afghanistan
```

In the data cleaning and transformation, we removed unnecessary columns for our analysis like latitude and longitude. We combined the cases into deaths per date into one variable called global. Combined_key puts together the county and the state for US cases.

Visualizing Data

```
#First, group US dataset by state, region, and date
US_by_state <- US %>%
  group_by(Province_State, Country_Region, date) %>%
  summarize(cases = sum(cases), deaths = sum(deaths),
            Population = sum(Population)) %>%
  mutate(deaths_per_mill = deaths * 1000000 / Population) %>%
  select(Province_State, Country_Region, date, cases, deaths, deaths_per_mill, Population) %>%
  ungroup()
```

```
## 'summarise()' has grouped output by 'Province_State', 'Country_Region'. You can
## override using the '.groups' argument.
```

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

```
## # A tibble: 10 x 7
##   Province_State Country_Region date       cases deaths deaths_per_mill
##   <chr>           <chr>       <date>     <dbl>  <dbl>         <dbl>
## 1 Alabama        US         2020-01-22      0      0             0
## 2 Alabama        US         2020-01-23      0      0             0
## 3 Alabama        US         2020-01-24      0      0             0
## 4 Alabama        US         2020-01-25      0      0             0
## 5 Alabama        US         2020-01-26      0      0             0
## 6 Alabama        US         2020-01-27      0      0             0
## 7 Alabama        US         2020-01-28      0      0             0
## 8 Alabama        US         2020-01-29      0      0             0
## 9 Alabama        US         2020-01-30      0      0             0
## 10 Alabama       US         2020-01-31      0      0             0
## # i 1 more variable: Population <dbl>
```

#look at total for US, group by Country/Region and date

```
US_totals <- US_by_state %>%
  group_by(Country_Region, date) %>%
  summarize(cases = sum(cases), deaths = sum(deaths),
            Population = sum(Population)) %>%
  mutate(deaths_per_mill = deaths * 1000000 / Population) %>%
  select(Country_Region, date, cases, deaths, deaths_per_mill, Population) %>%
  ungroup()
```

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

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

```
## # A tibble: 10 x 6
##   Country_Region date       cases deaths deaths_per_mill Population
##   <chr>         <date>     <dbl>  <dbl>         <dbl>     <dbl>
## 1 US           2020-01-22      1      1         0.00300 332875137
## 2 US           2020-01-23      1      1         0.00300 332875137
## 3 US           2020-01-24      2      1         0.00300 332875137
## 4 US           2020-01-25      2      1         0.00300 332875137
## 5 US           2020-01-26      5      1         0.00300 332875137
## 6 US           2020-01-27      5      1         0.00300 332875137
## 7 US           2020-01-28      5      1         0.00300 332875137
## 8 US           2020-01-29      6      1         0.00300 332875137
## 9 US           2020-01-30      6      1         0.00300 332875137
## 10 US          2020-01-31      8      1         0.00300 332875137
```

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

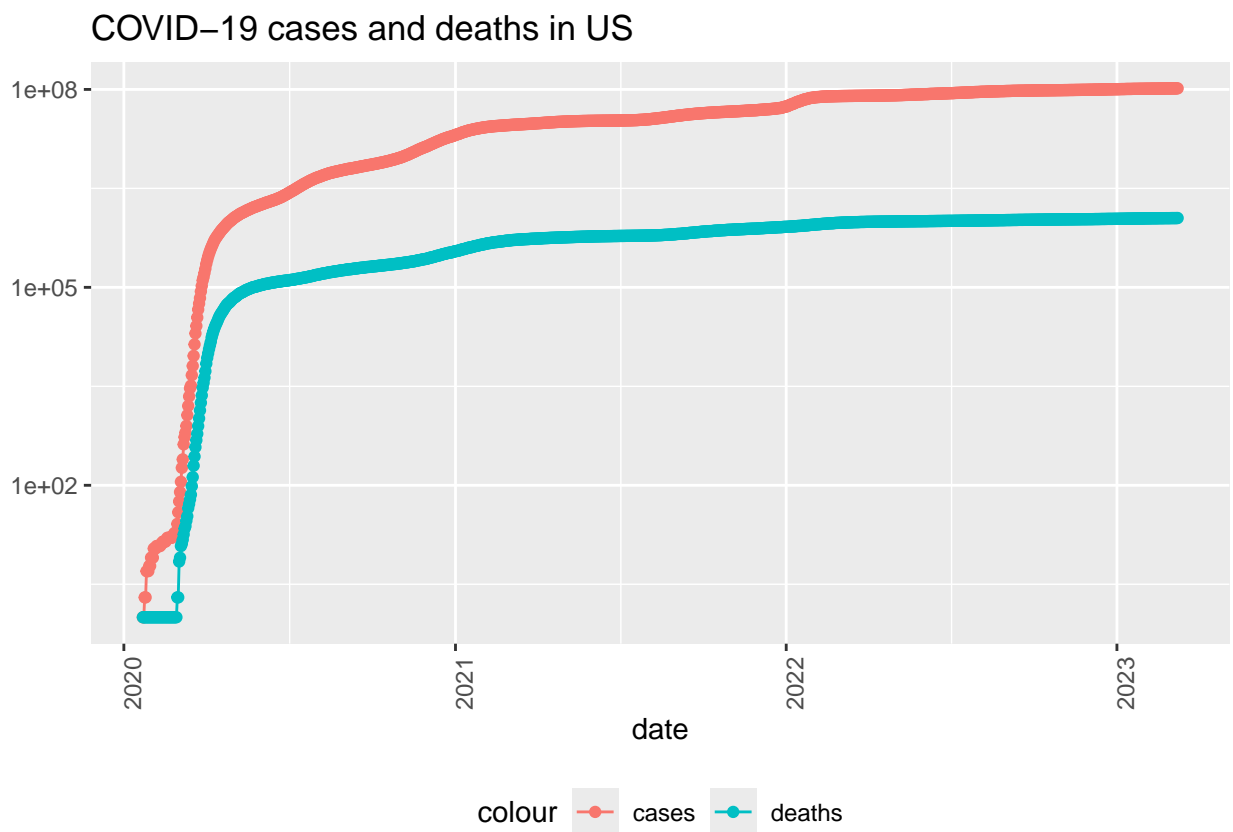
```
## # A tibble: 10 x 6
##   Country_Region date       cases deaths deaths_per_mill Population
##   <chr>         <date>     <dbl>  <dbl>         <dbl>     <dbl>
```



```
## 1 US      2023-02-28 103443455 1119917      3364.  332875137
## 2 US      2023-03-01 103533872 1120897      3367.  332875137
## 3 US      2023-03-02 103589757 1121658      3370.  332875137
## 4 US      2023-03-03 103648690 1122165      3371.  332875137
## 5 US      2023-03-04 103650837 1122172      3371.  332875137
## 6 US      2023-03-05 103646975 1122134      3371.  332875137
## 7 US      2023-03-06 103655539 1122181      3371.  332875137
## 8 US      2023-03-07 103690910 1122516      3372.  332875137
## 9 US      2023-03-08 103755771 1123246      3374.  332875137
## 10 US     2023-03-09 103802702 1123836      3376.  332875137
```

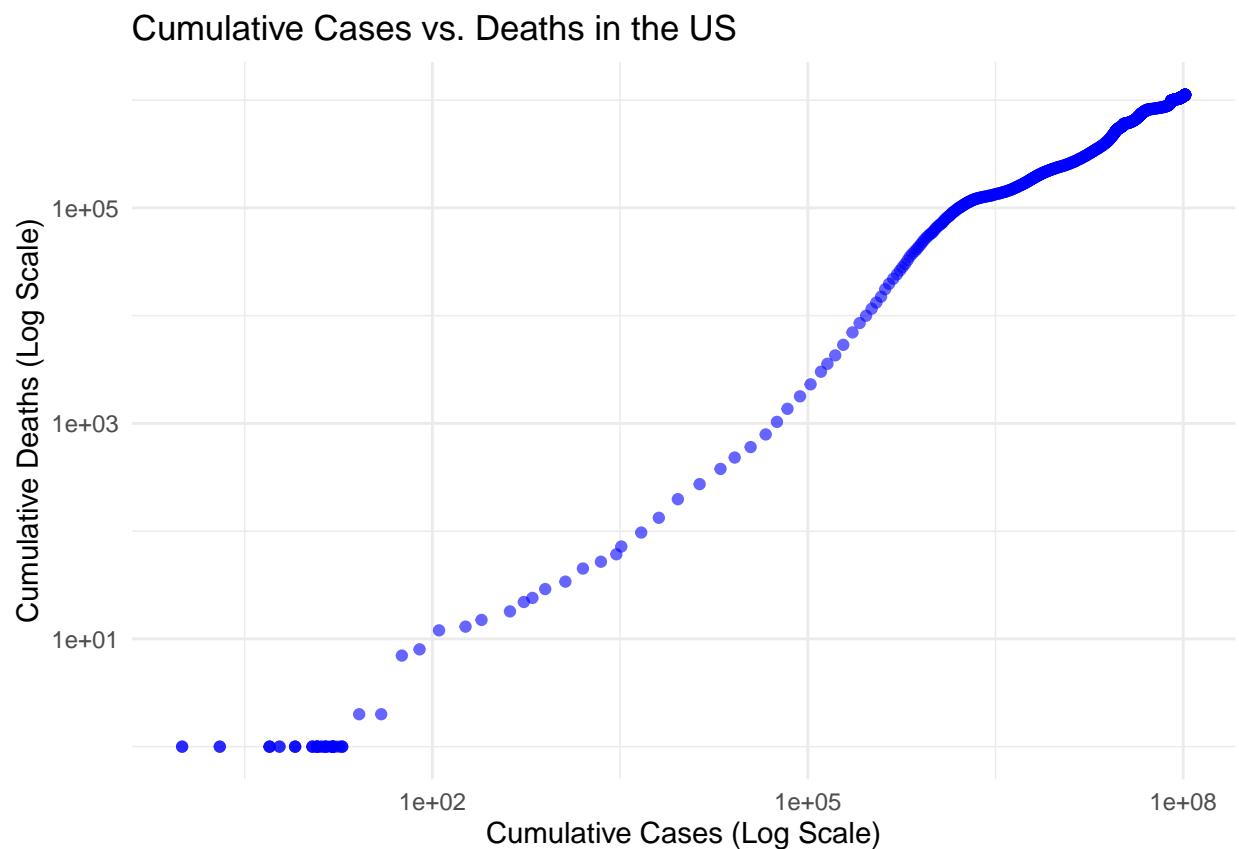
```
#plot US data
```

```
US_totals %>%
  filter(cases > 0) %>%
  ggplot(aes(x=date, y=cases)) +
  geom_line(aes(color="cases")) +
  geom_point(aes(color="cases")) +
  geom_line(aes(y=deaths, color="deaths")) +
  geom_point(aes(y=deaths, color="deaths")) +
  scale_y_log10() +
  theme(legend.position="bottom",
        axis.text.x=element_text(angle=90)) +
  labs(title = "COVID-19 cases and deaths in US", y=NULL)
```



#New Visualization 1: Cumulative Cases vs. Deaths (Scatter Plot with Log Scale)
This scatter plot will show the relationship between cumulative cases and cumulative deaths in the US
We'll use a log scale to visualize a wide range of values effectively.

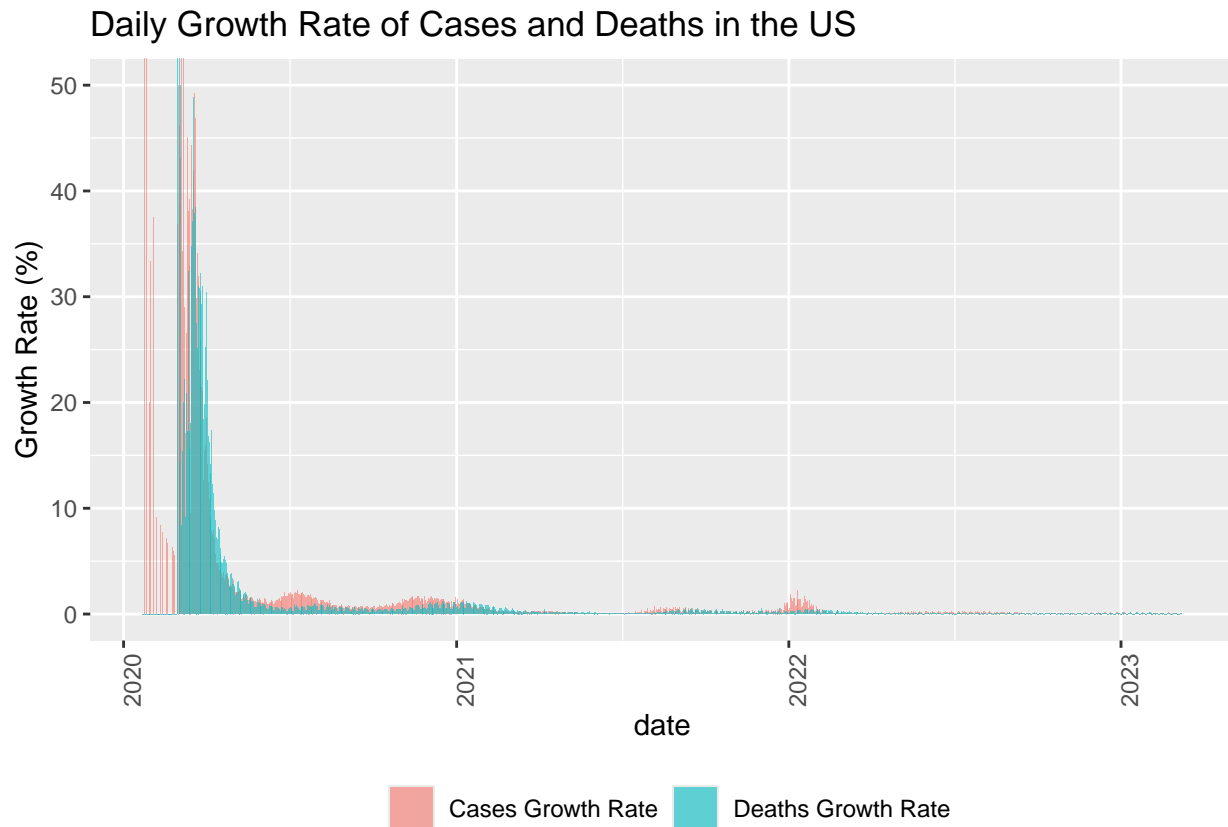
```
US_totals %>%
  filter(cases > 0, deaths > 0) %>%
  ggplot(aes(x = cases, y = deaths)) +
  geom_point(alpha = 0.6, color = "blue") +
  scale_x_log10() +
  scale_y_log10() +
  labs(title = "Cumulative Cases vs. Deaths in the US",
       x = "Cumulative Cases (Log Scale)",
       y = "Cumulative Deaths (Log Scale)") +
  theme_minimal()
```



#New Visualization 2: Cases and Deaths Growth Rate (Bar Chart)
#This bar chart tracks the daily percentage growth rates of both cases and deaths, helping to highlight

```
US_totals %>%
  filter(cases > 0, deaths > 0) %>%
  mutate(growth_rate_cases = (cases - lag(cases)) / lag(cases) * 100,
         growth_rate_deaths = (deaths - lag(deaths)) / lag(deaths) * 100) %>%
  replace_na(list(growth_rate_cases = 0, growth_rate_deaths = 0)) %>%
  ggplot(aes(x = date)) +
  geom_bar(aes(y = growth_rate_cases, fill = "Cases Growth Rate"), stat = "identity", alpha = 0.6) +
  geom_bar(aes(y = growth_rate_deaths, fill = "Deaths Growth Rate"), stat = "identity", alpha = 0.6) +
```

```
coord_cartesian(ylim = c(0, 50)) + # Adjust y-axis limits to zoom in
labs(title = "Daily Growth Rate of Cases and Deaths in the US",
     y = "Growth Rate (%)",
     fill = "") +
theme(axis.text.x = element_text(angle = 90),
     legend.position = "bottom")
```



In the first plot of ‘COVID-19 cases and deaths in the US’, we see that cases and deaths follow similar trends with a large increase around March/April 2020 and continue to increase but somewhat level-out through the beginning of 2023. In the new visualization, Cumulative Cases vs Deaths in the US, it shows the relationship between cumulative cases and cumulative deaths in the US which seems to be a positive linear relationship. In the second new visualization, Daily Growth Rate of Cases and Deaths in the US, this bar chart tracks the daily percentage growth rates of both cases and deaths, helping to highlight periods of rapid growth or slowdown. We see large growth rates in the beginning of 2020 and a few other spikes in the beginning of 2021 and 2022.

Further Transformation and Analysis

```
US_by_state <- US_by_state %>%
  mutate(new_cases = cases - lag(cases),
         new_deaths = deaths - lag(deaths))

US_totals <- US_totals %>%
  mutate(new_cases = cases - lag(cases),
```

```

    new_deaths = deaths - lag(deaths))

#transform to group by state and deaths and cases per thousand
US_state_totals <- US_by_state %>%
  group_by(Province_State) %>%
  summarize(deaths = max(deaths), cases = max(cases),
            population = max(Population),
            cases_per_thou = 1000 * cases / population,
            deaths_per_thou = 1000 * deaths / population) %>%
  filter(cases >0, population >0)

#Lets look at the 10 states that had the least amount of deaths per thousand
US_state_totals %>%
  slice_min(deaths_per_thou, n=10) %>%
  select(deaths_per_thou, cases_per_thou, everything())

## # A tibble: 10 x 6
##   deaths_per_thou cases_per_thou Province_State deaths cases population
##   <dbl>          <dbl> <chr>          <dbl> <dbl>    <dbl>
## 1         0.611         150. American Samoa         34 8.32e3    55641
## 2         0.744         248. Northern Mariana Isl~         41 1.37e4    55144
## 3         1.21         231. Virgin Islands         130 2.48e4   107268
## 4         1.30         269. Hawaii         1841 3.81e5   1415872
## 5         1.49         245. Vermont          929 1.53e5    623989
## 6         1.55         293. Puerto Rico        5823 1.10e6   3754939
## 7         1.65         340. Utah          5298 1.09e6   3205958
## 8         2.01         415. Alaska          1486 3.08e5    740995
## 9         2.03         252. District of Columbia    1432 1.78e5    705749
## 10        2.06         253. Washington       15683 1.93e6   7614893

#Here are the 10 states that had the most amount of deaths per thousand
US_state_totals %>%
  slice_max(deaths_per_thou, n=10) %>%
  select(deaths_per_thou, cases_per_thou, everything())

## # A tibble: 10 x 6
##   deaths_per_thou cases_per_thou Province_State deaths cases population
##   <dbl>          <dbl> <chr>          <dbl> <dbl>    <dbl>
## 1         4.55         336. Arizona        33102 2443514   7278717
## 2         4.54         326. Oklahoma        17972 1290929   3956971
## 3         4.49         333. Mississippi     13370 990756    2976149
## 4         4.44         359. West Virginia    7960 642760    1792147
## 5         4.32         320. New Mexico       9061 670929    2096829
## 6         4.31         334. Arkansas        13020 1006883   3017804
## 7         4.29         335. Alabama         21032 1644533   4903185
## 8         4.28         368. Tennessee       29263 2515130   6829174
## 9         4.23         307. Michigan        42205 3064125   9986857
## 10        4.06         385. Kentucky        18130 1718471   4467673

```

In this analysis, we wanted to look at which states had the most and least deaths per thousand residents. The state/territory with the least amount of deaths per thousand residents was American Samoa and the state territory with the most amount of deaths per thousand residents was Arizona.

Modeling the Data

```
# Simple Linear Regression Model to predict deaths based on the number of cases in the US.

# Filter data to only include rows where both cases and deaths are greater than 0
US_lm_data <- US_totals %>%
  filter(cases > 0, deaths > 0)

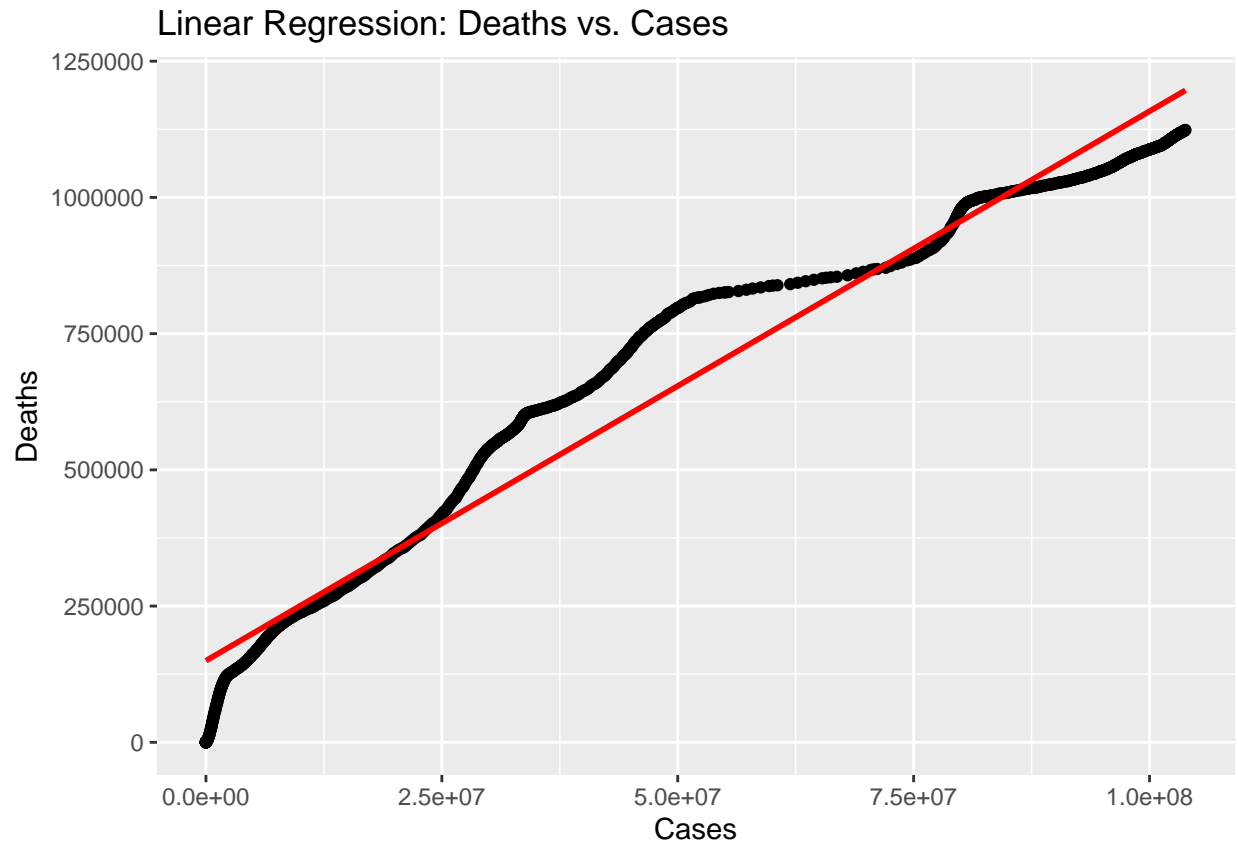
# Fit a simple linear regression model
model <- lm(deaths ~ cases, data = US_lm_data)

# Summary of the model
summary(model)
```

```
##
## Call:
## lm(formula = deaths ~ cases, data = US_lm_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -149805  -62016  -13343   89500  143891
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.498e+05  3.869e+03  38.72  <2e-16 ***
## cases       1.008e-02  6.498e-05  155.19  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 80090 on 1141 degrees of freedom
## Multiple R-squared:  0.9548, Adjusted R-squared:  0.9547
## F-statistic: 2.408e+04 on 1 and 1141 DF, p-value: < 2.2e-16
```

```
# Visualize the regression model
ggplot(US_lm_data, aes(x = cases, y = deaths)) +
  geom_point() +
  geom_smooth(method = "lm", se = FALSE, color = "red") +
  labs(title = "Linear Regression: Deaths vs. Cases", x = "Cases", y = "Deaths")
```

```
## 'geom_smooth()' using formula = 'y ~ x'
```



The modeled data is a simple linear regression model to predict deaths based on the number of cases in the US. The data was filtered to only include rows where both cases and deaths are greater than 0. We see that based on the total data given between early 2020 and March 2023, the model is predicting an increase in deaths. There is a positive linear relationship between the amount of cases and deaths. We can also see from the summary output that the p-value for the predictor variable, cases, is significant (< 0.05).