

Covid Data Analysis

Nate Sema

2025-06-16

In this project, I analyzed COVID-19 data from 01-01-2020 to 04-30-2021 using datasets from Our World In Data: <https://ourworldindata.org/covid-deaths>. The goal was to look at COVID-19 cases, deaths, and vaccinations to answer key questions about the global impact of the pandemic. I explored how many total cases and deaths each country reported, examined daily new cases in the United States over time, and compared countries to see which had higher death rates relative to total cases. I also looked at the relationship between vaccinations and total cases to see if higher vaccination numbers were linked to fewer infections. Additionally, I compared daily new cases among the United States, India, and Brazil, and to account for population differences, I analyzed deaths per million people to provide a fairer comparison across countries of different sizes.

```
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    3.5.2      v tibble    3.3.0
## v lubridate  1.9.4      v tidyr     1.3.1
## v purrr      1.0.4
## -- 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)
library(scales)
```

```
##
## Attaching package: 'scales'
##
## The following object is masked from 'package:purrr':
##
##   discard
##
## The following object is masked from 'package:readr':
##
##   col_factor
```

```
library(tinytex)
```

```
#loading the datasets
deaths <- read_csv('CovidDeaths.csv')
```

```
## Rows: 85171 Columns: 26
## -- Column specification -----
## Delimiter: ","
## chr (4): iso_code, continent, location, date
## dbl (22): population, total_cases, new_cases, new_cases_smoothed, total_deat...
##
## 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.
```

```
vaccinations <- read_csv('CovidVaccinations.csv')
```

```
## Rows: 85171 Columns: 37
## -- Column specification -----
## Delimiter: ","
## chr (5): iso_code, continent, location, date, tests_units
## dbl (32): new_tests, total_tests, total_tests_per_thousand, new_tests_per_th...
##
## 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 the data -----
```

```
#making sure all dates are converted from character to numerical
deaths$date <- mdy(deaths$date)
vaccinations$date <- mdy(vaccinations$date)
str(deaths$date)
```

```
## Date[1:85171], format: "2020-02-24" "2020-02-25" "2020-02-26" "2020-02-27" "2020-02-28" ...
```

```
#checking for missing data by checking for TRUE/FALSE for each cell (is.na), then counting how many TRUE
colSums(is.na(deaths))
```

```
##               iso_code               continent
##                0                4111
##               location               date
##                0                0
##               population            total_cases
##                549                2099
##               new_cases            new_cases_smoothed
##               2101                3102
##               total_deaths            new_deaths
##               11763            11605
##               new_deaths_smoothed    total_cases_per_million
##               3102                2548
##               new_cases_per_million    new_cases_smoothed_per_million
##               2550                3546
##               total_deaths_per_million    new_deaths_per_million
##               12199            12041
##               new_deaths_smoothed_per_million    reproduction_rate
##               3546                16229
##               icu_patients    icu_patients_per_million
##               76487            76487
```

```
##           hosp_patients      hosp_patients_per_million
##           74357              74357
##       weekly_icu_admissions weekly_icu_admissions_per_million
##           84382              84382
##       weekly_hosp_admissions weekly_hosp_admissions_per_million
##           83876              83876
```

```
colSums(is.na(vaccinations))
```

```
##           iso_code              continent
##           0              4111
##           location              date
##           0              0
##           new_tests              total_tests
##           46226              46519
##       total_tests_per_thousand      new_tests_per_thousand
##           46519              46226
##           new_tests_smoothed      new_tests_smoothed_per_thousand
##           40546              40546
##           positive_rate              tests_per_case
##           42267              42860
##           tests_units              total_vaccinations
##           39092              75797
##           people_vaccinated      people_fully_vaccinated
##           76427              78740
##           new_vaccinations      new_vaccinations_smoothed
##           77217              70079
##       total_vaccinations_per_hundred      people_vaccinated_per_hundred
##           75797              76427
##       people_fully_vaccinated_per_hundred new_vaccinations_smoothed_per_million
##           78740              70079
##           stringency_index      population_density
##           12964              5897
##           median_age      aged_65_older
##           8465              9341
##           aged_70_older      gdp_per_capita
##           8895              8125
##           extreme_poverty      cardiovasc_death_rate
##           32722              7537
##           diabetes_prevalence      female_smokers
##           6392              24343
##           male_smokers      handwashing_facilities
##           25240              46164
##       hospital_beds_per_thousand      life_expectancy
##           14324              4338
##           human_development_index
##           7654
```

```
#Filtering out any rows of data where location AND date do not exist
deaths_clean <- deaths %>%
  filter(!is.na(location) & !is.na(date))

vaccinations_clean <- vaccinations %>%
```

```

filter(!is.na(location) & !is.na(date))

#joining the two data sets to see covid deaths and vaccinations together for each place and date.

covid_data <- inner_join(deaths_clean, vaccinations_clean,
                        by = c('location', 'date'))

head(covid_data)

```

```

## # A tibble: 6 x 61
##   iso_code.x continent.x location    date      population total_cases new_cases
##   <chr>      <chr>      <chr>    <date>      <dbl>      <dbl>      <dbl>
## 1 AFG      Asia      Afghanistan 2020-02-24  38928341      1          1
## 2 AFG      Asia      Afghanistan 2020-02-25  38928341      1          0
## 3 AFG      Asia      Afghanistan 2020-02-26  38928341      1          0
## 4 AFG      Asia      Afghanistan 2020-02-27  38928341      1          0
## 5 AFG      Asia      Afghanistan 2020-02-28  38928341      1          0
## 6 AFG      Asia      Afghanistan 2020-02-29  38928341      1          0
## # i 54 more variables: new_cases_smoothed <dbl>, total_deaths <dbl>,
## #   new_deaths <dbl>, new_deaths_smoothed <dbl>, total_cases_per_million <dbl>,
## #   new_cases_per_million <dbl>, new_cases_smoothed_per_million <dbl>,
## #   total_deaths_per_million <dbl>, new_deaths_per_million <dbl>,
## #   new_deaths_smoothed_per_million <dbl>, reproduction_rate <dbl>,
## #   icu_patients <dbl>, icu_patients_per_million <dbl>, hosp_patients <dbl>,
## #   hosp_patients_per_million <dbl>, weekly_icu_admissions <dbl>, ...

```

Questions to answer: 1: How many total cases and deaths were there per country? 2: What did the daily new cases look like over time for the United States? 3: Which countries had higher death rates relative to total cases? 4: Looking at vaccinations vs cases to see if countries with more vaccinations had fewer cases. 5: How does the US compare with India, and Brazil on daily new cases 6: Since total deaths alone can be misleading as big countries will have bigger totals, what would deaths per million look like?

```

# Question 1: How many total cases and deaths were there per country?

total_cases <- covid_data %>%
  filter(!location %in% c("World", "Europe", "European Union",
                        "Asia", "South America", "North America",
                        "Africa", "Oceania")) %>% #filtering world/continent totals
  group_by(location) %>%
  summarise(
    total_cases = max(total_cases, na.rm = TRUE),
    total_deaths = max(total_deaths, na.rm = TRUE)
  ) %>%
  slice_max(total_cases, n = 10) #Displaying the top 10

```

```

## Warning: There were 48 warnings in 'summarise()'.
## The first warning was:
## i In argument: 'total_cases = max(total_cases, na.rm = TRUE)'.
## i In group 6: 'location = "Anguilla"'.
## Caused by warning in 'max()':
## ! no non-missing arguments to max; returning -Inf
## i Run 'dplyr::last_dplyr_warnings()' to see the 47 remaining warnings.

```

```
print(total_cases)
```

```
## # A tibble: 10 x 3
##   location      total_cases total_deaths
##   <chr>          <dbl>      <dbl>
## 1 United States  32346971    576232
## 2 India         19164969    211853
## 3 Brazil        14659011    403781
## 4 France        5677835     104675
## 5 Turkey        4820591      40131
## 6 Russia        4750755     108290
## 7 United Kingdom 4432246     127775
## 8 Italy          4022653     120807
## 9 Spain         3524077      78216
## 10 Germany       3405365      83097
```

Insight:

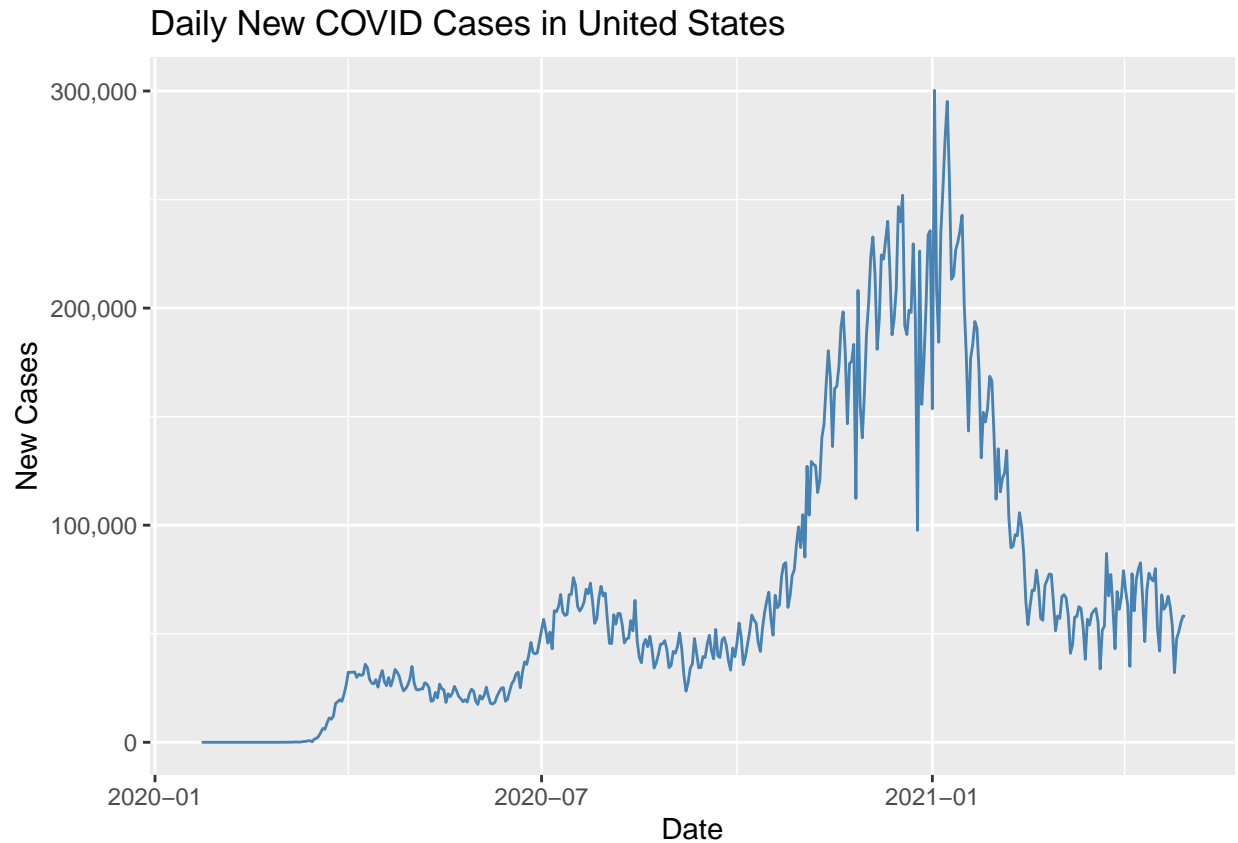
- At the time the data was collected, The United States had the highest cases (32,346,971) and the highest deaths (576,232).

```
# Question 2: What did the daily new cases look like over time for the United States?
```

```
country_data <- deaths_clean %>%
  filter(location == "United States")

ggplot(country_data, aes(x = date, y = new_cases)) +
  geom_line(color = 'steelblue') +
  labs(title = 'Daily New COVID Cases in United States',
       x = 'Date',
       y = 'New Cases') +
  scale_y_continuous(labels = comma)
```

```
## Warning: Removed 1 row containing missing values or values outside the scale range
## ('geom_line()').
```



Insight:

- Shows multiple waves of infection.
- The largest peak was during the winter months (Nov–Jan).
- After this peak, daily new cases decreased but did not drop to early-pandemic levels.

#Question 3: Which countries had higher death rates relative to total cases?

```
library(ggplot2)
```

```
country_totals <- deaths_clean %>%
  group_by(location) %>%
  summarise(total_cases = max(total_cases, na.rm = TRUE),
            total_deaths = max(total_deaths, na.rm = TRUE))
```

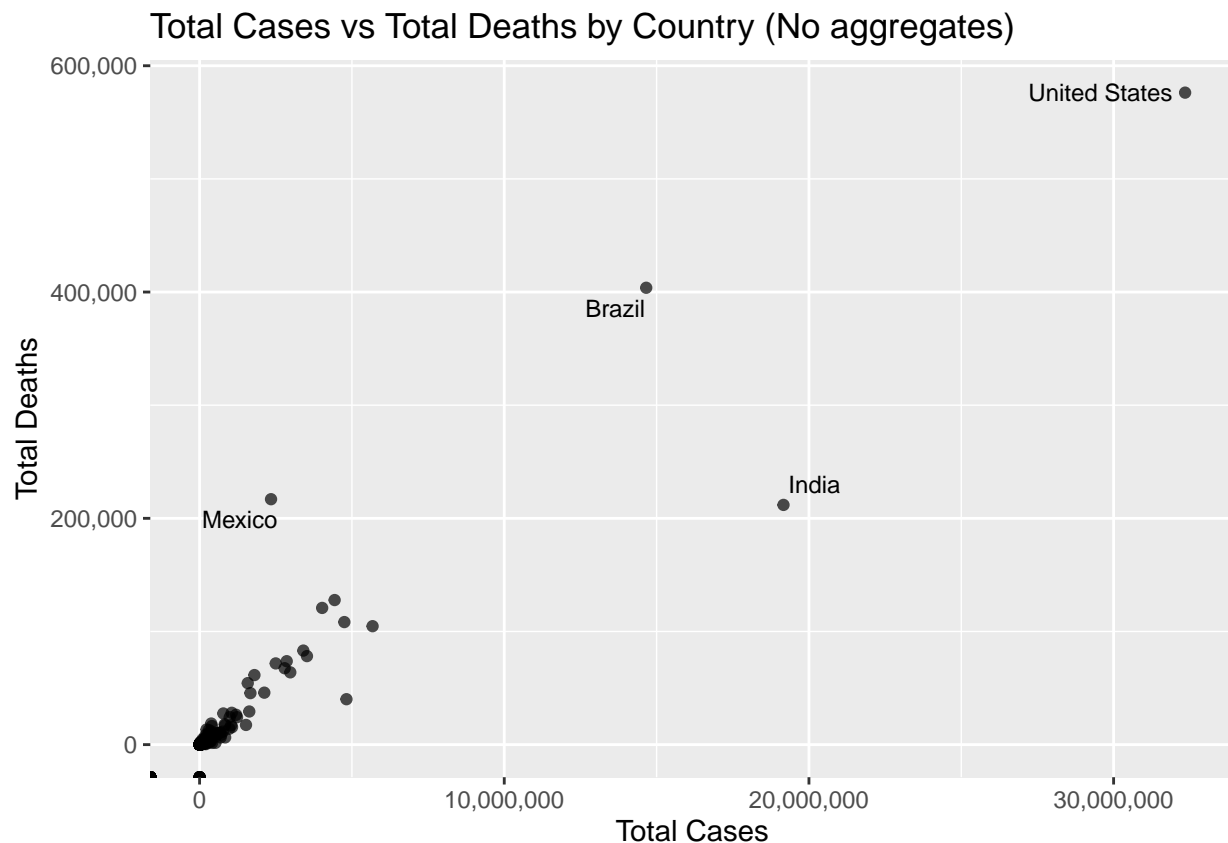
```
## Warning: There were 48 warnings in 'summarise()'.
## The first warning was:
## i In argument: 'total_cases = max(total_cases, na.rm = TRUE)'.
## i In group 7: 'location = "Anguilla"'.
## Caused by warning in 'max()':
## ! no non-missing arguments to max; returning -Inf
## i Run 'dplyr::last_dplyr_warnings()' to see the 47 remaining warnings.
```

```
# filtering out continents/global totals to focus on countries

country_totals_filtered <- country_totals %>%
  filter(!location %in% c('World', 'Europe', 'European Union',
                          'Asia', 'South America', 'North America',
                          'Africa', 'Oceania'))

# plotting filtered data
ggplot(country_totals_filtered, aes(x = total_cases, y = total_deaths, label = location)) +
  geom_point(alpha = 0.7) +
  geom_text_repel(size = 3) +
  labs(title = "Total Cases vs Total Deaths by Country (No aggregates)",
       x = "Total Cases",
       y = "Total Deaths") +
  scale_x_continuous(labels = scales::comma) +
  scale_y_continuous(labels = scales::comma)
```

```
## Warning: ggrepel: 207 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```



Insight:

- United States: Is on the far right and highest up indicating it had the highest total cases and highest total deaths.

- Brazil: Has high deaths compared to its total cases, showing a high fatality count for its caseload.
- Mexico: Has fewer cases than Brazil or India but a high death count — suggesting a higher death rate per reported case.

Question 4: Looking at vaccinations vs cases to see if countries with more vaccinations had fewer cases

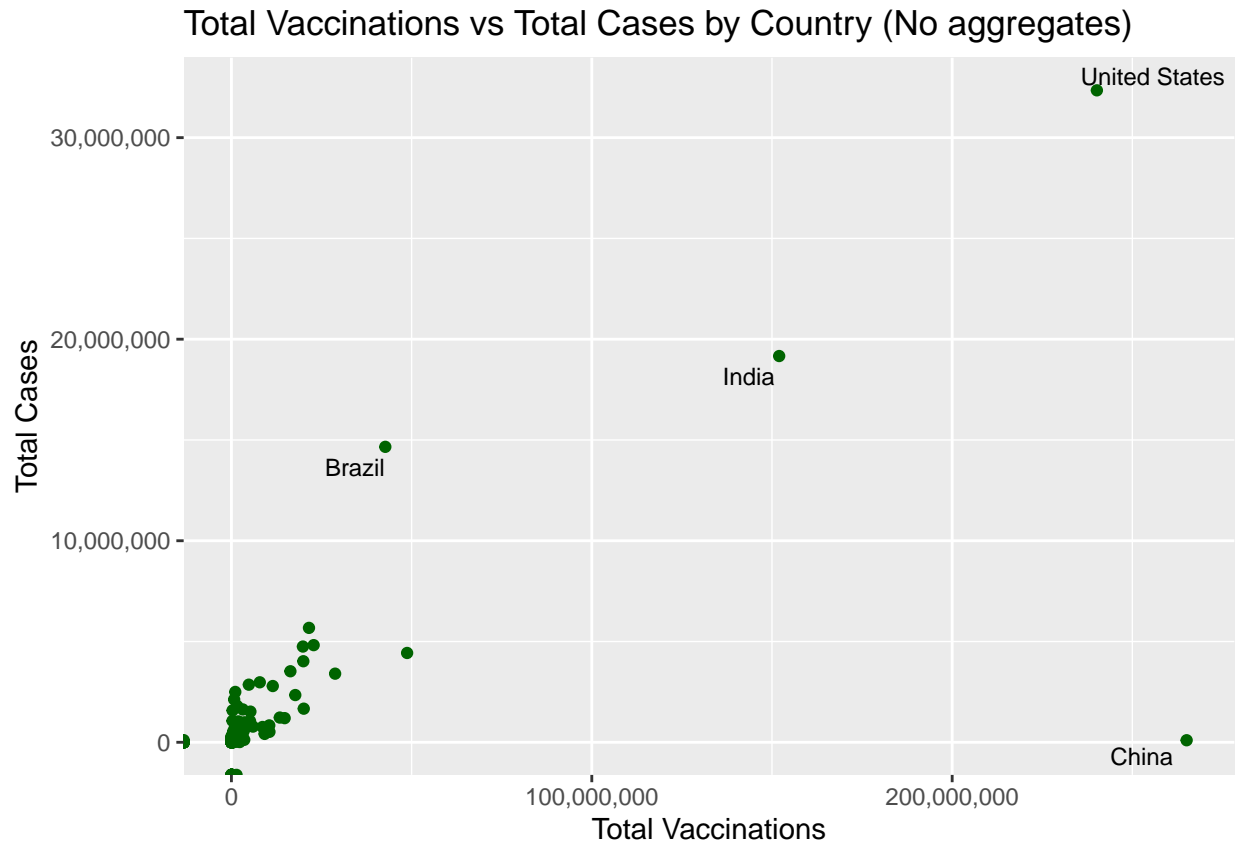
```
vacc_cases <- covid_data %>%
  filter(!location %in% c("World", "Europe", "European Union",
                        "Asia", "South America", "North America",
                        "Africa", "Oceania")) %>% #filtering world/continent totals
  group_by(location) %>%
  summarise(
    total_vaccinations = max(total_vaccinations, na.rm = TRUE),
    total_cases = max(total_cases, na.rm = TRUE)
  )
```

```
## Warning: There were 40 warnings in 'summarise()'.
## The first warning was:
## i In argument: 'total_vaccinations = max(total_vaccinations, na.rm = TRUE)'.
## i In group 21: 'location = "Benin"'.
## Caused by warning in 'max()':
## ! no non-missing arguments to max; returning -Inf
## i Run 'dplyr::last_dplyr_warnings()' to see the 39 remaining warnings.
```

#plotting filtered data

```
ggplot(vacc_cases, aes(x = total_vaccinations, y = total_cases, label = location)) +
  geom_point(color = "darkgreen") +
  geom_text_repel(size = 3) +
  labs(title = "Total Vaccinations vs Total Cases by Country (No aggregates)",
       x = "Total Vaccinations",
       y = "Total Cases") +
  scale_x_continuous(labels = scales::comma) +
  scale_y_continuous(labels = scales::comma)
```

```
## Warning: ggrepel: 207 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```

Countries to the right, have given more total vaccine doses while countries higher up have more total COVID cases over the whole pandemic. As expected, big population countries naturally had higher totals.

Insight:

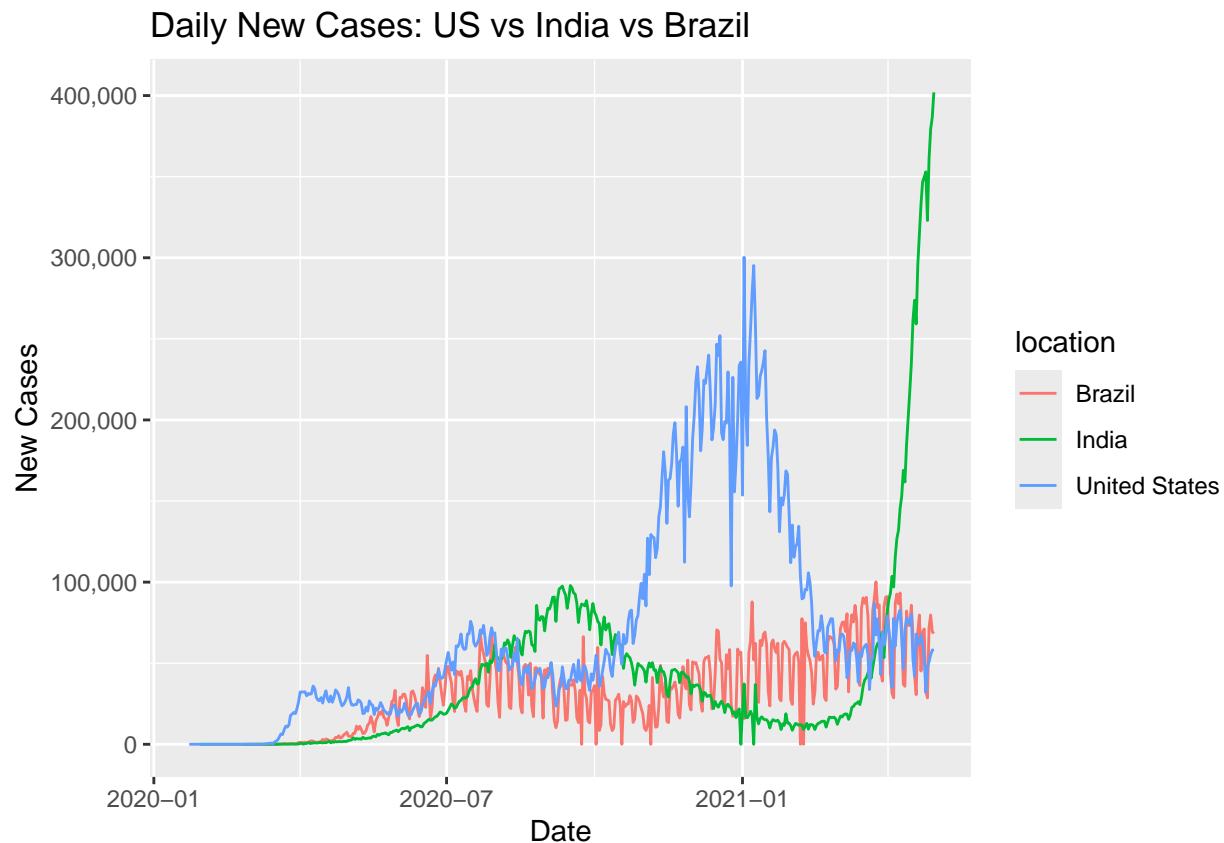
- China: Far right which means they had the most doses given by far. This could be a result of their huge population but since they were not very high in total cases compared to the US/India, it suggests lower per-capita infection, tied to aggressive vaccination campaign and stricter control.
- US & India: Were high on both axes had very large outbreaks AND large vaccination campaigns.
- France, Brazil: Mid-range vaccinations but also high total cases.

Question 5: How does the US compare with India, and Brazil on daily new cases

```
multi_country <- deaths_clean %>%
  filter(location %in% c('United States', 'India', 'Brazil'))

ggplot(multi_country, aes(x = date, y = new_cases, colour = location)) +
  geom_line() +
  labs(title = 'Daily New Cases: US vs India vs Brazil',
       x = 'Date',
       y = 'New Cases') +
  scale_y_continuous(labels = comma)
```

```
## Warning: Removed 1 row containing missing values or values outside the scale range
## ('geom_line()').
```



Insight: - The US had the highest peaks earlier (winter 2020–2021) but India's spike at the end becomes very steep — showing how cases surged rapidly there.

- Brazil had a more consistent high baseline, indicating ongoing spread without sharp spikes or deep valleys.
- Each country's curve reflects different outbreak timings and possibly different containment and reporting patterns.

*# Question 6: Since total deaths alone can be misleading as big countries will have bigger totals, what
Deaths per-capita (million) = total deaths / population * 1000000*

```
glimpse(deaths_clean)
```

```
## Rows: 85,171
## Columns: 26
## $ iso_code      <chr> "AFG", "AFG", "AFG", "AFG", "AFG", ~
## $ continent     <chr> "Asia", "Asia", "Asia", "Asia", "As~
## $ location      <chr> "Afghanistan", "Afghanistan", "Afgh~
## $ date          <date> 2020-02-24, 2020-02-25, 2020-02-26~
## $ population    <dbl> 38928341, 38928341, 38928341, 38928~
## $ total_cases   <dbl> 1, 1, 1, 1, 1, 1, 1, 1, 2, 4, 4, 4,~
## $ new_cases     <dbl> 1, 0, 0, 0, 0, 0, 0, 0, 1, 2, 0, 0,~
## $ new_cases_smoothed <dbl> NA, NA, NA, NA, NA, 0.143, 0.143, 0~
## $ total_deaths  <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ new_deaths    <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA,~
```

```
## $ new_deaths_smoothed      <dbl> NA, NA, NA, NA, NA, 0, 0, 0, 0, 0, ~
## $ total_cases_per_million  <dbl> 0.026, 0.026, 0.026, 0.026, 0.026, ~
## $ new_cases_per_million    <dbl> 0.026, 0.000, 0.000, 0.000, 0.000, ~
## $ new_cases_smoothed_per_million <dbl> NA, NA, NA, NA, NA, 0.004, 0.004, 0~
## $ total_deaths_per_million <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ new_deaths_per_million   <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ new_deaths_smoothed_per_million <dbl> NA, NA, NA, NA, NA, 0, 0, 0, 0, 0, ~
## $ reproduction_rate       <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ icu_patients             <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ icu_patients_per_million <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ hosp_patients            <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ hosp_patients_per_million <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ weekly_icu_admissions    <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ weekly_icu_admissions_per_million <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ weekly_hosp_admissions   <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
## $ weekly_hosp_admissions_per_million <dbl> NA, NA, NA, NA, NA, NA, NA, NA, NA, NA,~
```

```
countries_per_capita <- deaths_clean %>%
  filter(!location %in% c("World", "Europe", "European Union",
    "Asia", "South America", "North America",
    "Africa", "Oceania")) %>% #filtering world/continent totals
  group_by(location) %>%
  summarise(
    total_deaths = max(total_deaths, na.rm = TRUE),
    population = max(population, na.rm = TRUE)
  ) %>%
  mutate(
    deaths_per_million = (total_deaths / population) * 1000000
  )
```

```
## Warning: There were 30 warnings in 'summarise()'.
## The first warning was:
## i In argument: 'total_deaths = max(total_deaths, na.rm = TRUE)'.
## i In group 6: 'location = "Anguilla"'.
## Caused by warning in 'max()':
## ! no non-missing arguments to max; returning -Inf
## i Run 'dplyr::last_dplyr_warnings()' to see the 29 remaining warnings.
```

```
head(countries_per_capita)
```

```
## # A tibble: 6 x 4
##   location      total_deaths population deaths_per_million
##   <chr>          <dbl>      <dbl>          <dbl>
## 1 Afghanistan    2625    38928341          67.4
## 2 Albania         2394    2877800          832.
## 3 Algeria         3253   43851043          74.2
## 4 Andorra         125     77265          1618.
## 5 Angola          596   32866268          18.1
## 6 Anguilla       -Inf     15002          -Inf
```

```
#Plotting deaths per million in comparison to total cases
```

```

country_per_capita <- deaths_clean %>%
  filter(!location %in% c("World", "Europe", "European Union",
                          "Asia", "South America", "North America",
                          "Africa", "Oceania")) %>% #filtering world/continent totals

group_by(location) %>%
  summarise(
    total_cases = max(total_cases, na.rm = TRUE),
    total_deaths = max(total_deaths, na.rm = TRUE),
    population = max(population, na.rm = TRUE)
  ) %>%
  mutate(
    deaths_per_million = (total_deaths / population) * 1000000
  )

```

```

## Warning: There were 50 warnings in 'summarise()'.
## The first warning was:
## i In argument: 'total_cases = max(total_cases, na.rm = TRUE)'.
## i In group 6: 'location = "Anguilla"'.
## Caused by warning in 'max()':
## ! no non-missing arguments to max; returning -Inf
## i Run 'dplyr::last_dplyr_warnings()' to see the 49 remaining warnings.

```

```

ggplot(country_per_capita, aes(x = total_cases, y = deaths_per_million, label = location)) +
  geom_point(color = "purple") +
  geom_text_repel(size = 3) +
  labs(
    title = "Deaths per Million vs Total Cases by Country",
    x = "Total Cases",
    y = "Deaths per Million"
  ) +
  scale_x_continuous(labels = scales::comma) +
  scale_y_continuous(labels = scales::comma)

```

```

## Warning: Removed 1 row containing missing values or values outside the scale range
## ('geom_point()').

```

```

## Warning: Removed 1 row containing missing values or values outside the scale range
## ('geom_text_repel()').

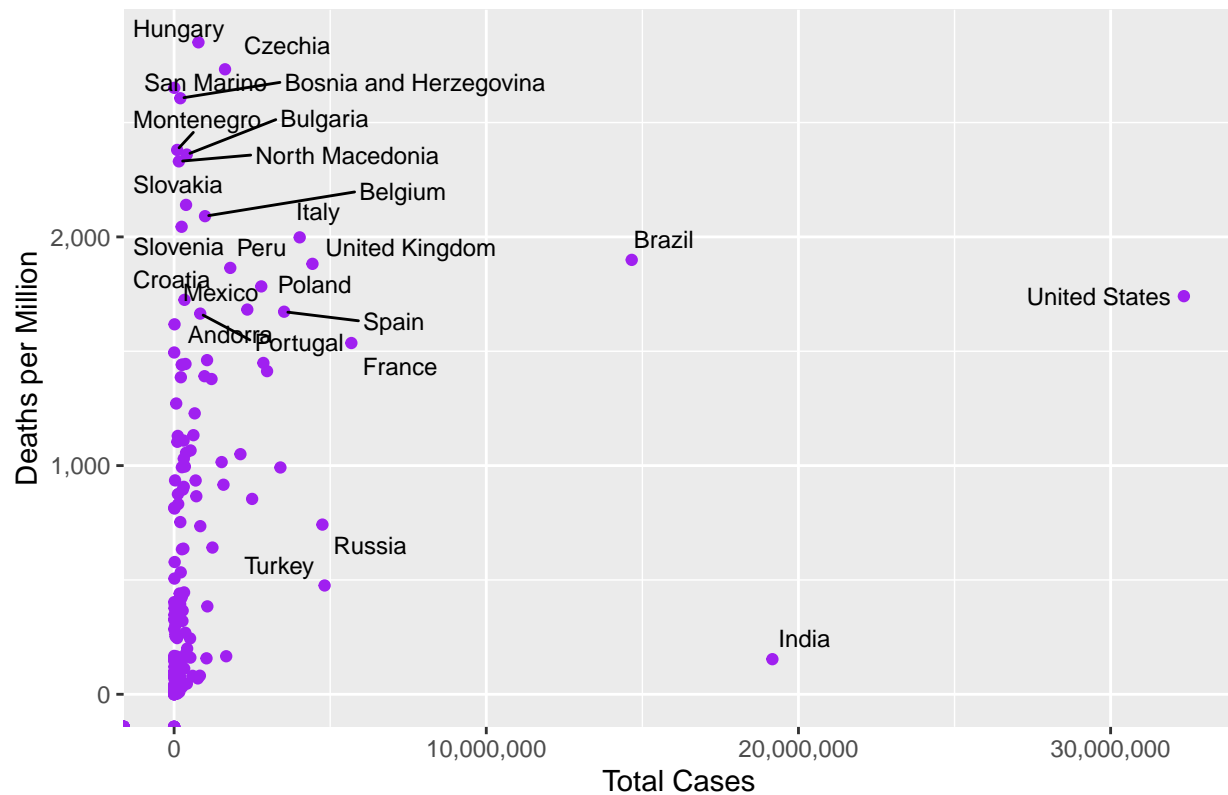
```

```

## Warning: ggrepel: 185 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps

```

Deaths per Million vs Total Cases by Country



Insight:

- Countries like Hungary and Czechia stand out for having very high deaths per million despite not having the largest total case counts indicating high severity.
- Large-population countries like India report very high case numbers but appear low on deaths per million, which could reflect underreporting, or other factors.
- The US and Brazil show both large outbreaks and high per-capita death tolls, indicating significant health impacts.