

Problem set 4

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In the next problem set, we plan to explore the relationship between COVID-19 death rates and vaccination rates across US states by visually examining their correlation. This analysis will involve gathering COVID-19 related data from the CDC's API and then extensively processing it to merge the various datasets. Since the population sizes of states vary significantly, we will focus on comparing rates rather than absolute numbers. To facilitate this, we will also source population data from the US Census to accurately calculate these rates.

In this problem set we will learn how to extract and wrangle data from the data US Census and CDC APIs.

```
pkgs <- c(
  "httr2",      # HTTP requests and API interactions
  "tidyverse", # Data manipulation, visualization, and analysis
  "janitor",    # Data cleaning and formatting
  "jsonlite",  # JSON data parsing
  "lubridate"   # Date and time manipulation
)

to_install <- setdiff(pkgs, rownames(installed.packages()))
if (length(to_install)) {
  install.packages(to_install, dependencies = TRUE)
}

invisible(lapply(pkgs, require, character.only = TRUE))
```

Loading required package: httr2

Loading required package: tidyverse

— Attaching core tidyverse packages — tidyverse 2.0.0 —

✓ dplyr	1.1.4	✓ readr	2.1.5
✓ forcats	1.0.1	✓ stringr	1.5.2
✓ ggplot2	4.0.0	✓ tibble	3.3.0
✓ lubridate	1.9.4	✓ tidyr	1.3.1
✓ purrr	1.1.0		

— 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

Loading required package: janitor

Attaching package: 'janitor'

The following objects are masked from 'package:stats':

chisq.test, fisher.test

Loading required package: jsonlite

Attaching package: 'jsonlite'

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

flatten

1. Get an API key from the US Census at https://api.census.gov/data/key_signup.html. You can't share this public key. But your code has to run on a TFs computer. Assume the TF will have a file in their working directory named `census-key.R` with the following one line of code:

```
census_key <- "A_CENSUS_KEY_THAT_WORKS"
```

Write a first line of code for your problem set that defines `census_key` by running the code in the file `census-key.R`.

```
source("census_key.R")
```

2. The [US Census API User Guide](#) provides details on how to leverage this valuable resource. We are interested in vintage population estimates for years 2021 and 2022. From the documentation we find that the *endpoint* is:

```
url <- "https://api.census.gov/data/2021/pep/population"
```

Use the **httr2** package to construct the following GET request.

```
https://api.census.gov/data/2021/pep/population?  
get=POP_2020,POP_2021,NAME&for=state:*&key=YOURKEYHERE
```

Create an object called `request` of class `httr2_request` with this URL as an endpoint. Hint: Print out `request` to check that the URL matches what we want.

```
library(httr2)  
  
?request()  
?req_url_query  
request <- request(url) |>  
  req_url_query(  
    get = "POP_2020,POP_2021,NAME",
```

```
`for` = "state:*",
key = census_key
)
```

```
request
```

```
<httr2_request>
```

```
GET https://api.census.gov/data/2021/pep/population?
```

```
get=POP_2020%2CP0P_2021%2CNAME&for=state%3A%2A&key=09e560e2345990d79e92cf06a9a99288a8f910af
```

```
Body: empty
```

```
# <httr2_request>
# GET https://api.census.gov/data/2021/pep/population?get=POP_2020%2CP0P_2021%2CNAME&for=
# Body: empty

# Looks correct: %2C = , %3A = : %2A = *
```

3. Make a request to the US Census API using the `request` object. Save the response to an object named `response`. Check the response status of your request and make sure it was successful. You can learn about *status codes* [here](#).

```
?req_perform
# Make the request and store the response
response <- req_perform(request)

# Check status code (200 = OK)
status <- resp_status(response)
if (status != 200) {
  stop(paste("Request failed with status", status))
} else {
  message("Success: status ", status)
}
```

```
Success: status 200
```

4. Use a function from the **httr2** package to determine the content type of your response.

```
resp_content_type(response)
```

```
[1] "application/json"
```

5. Use just one line of code and one function to extract the data into a matrix. Hints: 1) Use the `resp_body_json` function. 2) The first row of the matrix will be the variable names and this OK as we will fix in the next exercise.

```
?resp_body_json
population <- resp_body_json(response, simplifyVector = TRUE, simplifyDataFrame = FALSE,
```

6. Examine the `population` matrix you just created. Notice that 1) it is not tidy, 2) the column types are not what we want, and 3) the first row is a header. Convert `population` to a tidy dataset. Remove the state ID column and change the name of the column with state names to `state_name`. Add a column with state abbreviations called `state`. Make sure you assign the abbreviations for DC and PR correctly. Hint: Use the `janitor` package to make the first row the header.

```
library(tidyverse)
library(janitor)
head(population)
```

```
      [,1]      [,2]      [,3]      [,4]
[1,] "POP_2020" "POP_2021" "NAME"      "state"
[2,] "3962031"  "3986639"  "Oklahoma"  "40"
[3,] "1961455"  "1963692"  "Nebraska"  "31"
[4,] "1451911"  "1441553"  "Hawaii"    "15"
[5,] "887099"   "895376"   "South Dakota" "46"
[6,] "6920119"  "6975218"  "Tennessee" "47"
```

```
?janitor
population <- population |>
  as.data.frame(stringsAsFactors = FALSE) |>
  janitor::row_to_names(1) |>      # first row -> header
  janitor::clean_names() |>
  as_tibble() |>
  select(-state) |>               # drop state FIPS ID
  rename(state_name = name) |>
  mutate(
    pop_2020 = as.numeric(pop_2020),
    pop_2021 = as.numeric(pop_2021),
    state = case_when(
      state_name == "District of Columbia" ~ "DC",
      state_name == "Puerto Rico" ~ "PR",
      TRUE ~ state.abb[match(state_name, state.name)]
    )
  ) |>
  relocate(state, state_name, pop_2020, pop_2021)

head(population)
```

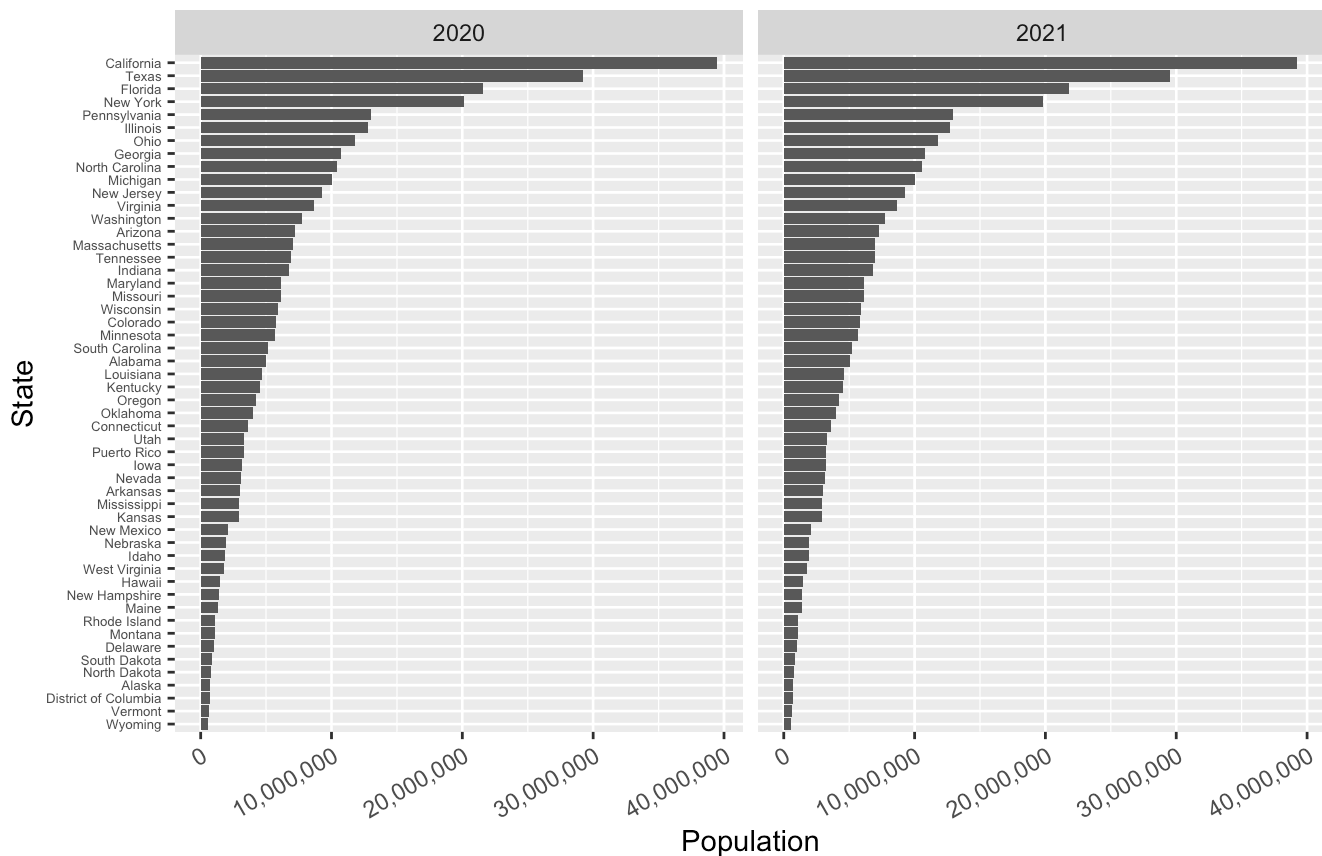
```
# A tibble: 6 × 4
  state state_name  pop_2020 pop_2021
<chr> <chr>         <dbl>    <dbl>
1 OK    Oklahoma       3962031  3986639
2 NE    Nebraska       1961455  1963692
3 HI    Hawaii         1451911  1441553
4 SD    South Dakota   887099   895376
5 TN    Tennessee     6920119  6975218
6 NV    Nevada         3114071  3143991
```

7. As a check, make a barplot of states' 2021 and 2022 populations. Show the state names in the y-axis ordered by population size. Hint: You will need to use `reorder` and use `facet_wrap`.

```
library(tidyr)

population |>
  dplyr::select(state_name, pop_2020, pop_2021) |>
  tidyr::pivot_longer(c(pop_2020, pop_2021),
                      names_to = "year", values_to = "population") |>
  dplyr::mutate(year = as.integer(sub("^pop_", "", year))) |>
  dplyr::group_by(year) |>
  dplyr::mutate(state_order = reorder(state_name, population)) |>
  dplyr::ungroup() |>
  ggplot2::ggplot(ggplot2::aes(x = state_order, y = population)) +
  ggplot2::geom_col() +
  ggplot2::coord_flip() +
  ggplot2::facet_wrap(~ year) +
  ggplot2::scale_y_continuous(labels = scales::label_comma()) +
  ggplot2::labs(
    title = "State Populations, 2020 vs 2021",
    x = "State", y = "Population",
    caption = "Source: US Census PEP"
  ) +
  ggplot2::theme(axis.text.y = element_text(size=5),
                 axis.text.x = element_text(angle = 30, hjust = 1))
```

State Populations, 2020 vs 2021



Source: US Census PEP

8. The following URL:

```
url <- "https://github.com/datasciencelabs/2025/raw/refs/heads/main/data/regions.json"
```

points to a JSON file that lists the states in the 10 Public Health Service (PHS) defined by CDC. We want to add these regions to the `population` dataset. To facilitate this create a data frame called `regions` that has two columns `state_name`, `region`, `region_name`. One of the regions has a long name. Change it to something shorter.

```
library(jsonlite)
library(purrr)
url <- "https://github.com/datasciencelabs/2025/raw/refs/heads/main/data/regions.json"
# regions <- use jsonlit JSON parser
# regions <- convert list to data frame. You can use map_df in purrr package

regions <- jsonlite::fromJSON(url, simplifyVector = FALSE) |>
  purrr::map_dfr(~ tibble(
    region      = as.integer(.x$region[[1]]),
    region_name = as.character(.x$region_name),
    state_name  = unlist(.x$states, use.names = FALSE)
  )) |>
  dplyr::mutate(region_name = dplyr::if_else(region == 2, "NY-NJ-PR-VI", region_name)) |>
  dplyr::filter(state_name %in% c(state.name, "District of Columbia", "Puerto Rico")) |>
```

```
dplyr::arrange(region, state_name)

table(regions$region_name)
```

Central Plains	Mid-Atlantic	Midwest	Mountain States
4	6	6	6
New England	NY-NJ-PR-VI	Pacific	Pacific Northwest
6	3	4	4
South Central	Southeast		
5	8		

9. Add a region and region name columns to the `population` data frame.

```
population <- population |>
  dplyr::left_join(regions, by = "state_name")

head(population)
```

```
# A tibble: 6 × 6
  state state_name  pop_2020 pop_2021 region region_name
<chr> <chr>      <dbl>    <dbl> <int> <chr>
1 OK    Oklahoma    3962031  3986639     6 South Central
2 NE    Nebraska    1961455  1963692     7 Central Plains
3 HI    Hawaii      1451911  1441553     9 Pacific
4 SD    South Dakota  887099   895376     8 Mountain States
5 TN    Tennessee    6920119  6975218     4 Southeast
6 NV    Nevada       3114071  3143991     9 Pacific
```

10. From reading <https://data.cdc.gov/> we learn the endpoint <https://data.cdc.gov/resource/pwn4-m3yp.json> provides state level data from SARS-COV2 cases. Use the **httr2** tools you have learned to download this into a data frame. Is all the data there? If not, comment on why.

```
api <- "https://data.cdc.gov/resource/pwn4-m3yp.json"
# Make the request
resp <- request(api) |>
  req_user_agent("bst260-pset04/1.0") |>
  req_perform()

# Check status/content
stopifnot(resp_status(resp) == 200,
  grepl("json", resp_content_type(resp), ignore.case = TRUE))

# Parse to a data frame
cases_raw <- resp_body_json(resp, simplifyVector = TRUE) |> as_tibble()

dim(cases_raw)
```

```
[1] 1000 10
```

```
names(cases_raw)
```

```
[1] "date_updated"      "state"              "start_date"
[4] "end_date"          "tot_cases"          "new_cases"
[7] "tot_deaths"        "new_deaths"         "new_historic_cases"
[10] "new_historic_deaths"
```

```
head(cases_raw, 3)
```

```
# A tibble: 3 × 10
  date_updated      state start_date end_date tot_cases new_cases tot_deaths
  <chr>            <chr> <chr>    <chr>    <chr>    <chr>    <chr>
1 2023-02-23T00:00:00... AZ    2023-02-1... 2023-02... 2434631.0 3716.0    33042.0
2 2022-12-22T00:00:00... LA    2022-12-1... 2022-12... 1507707.0 4041.0    18345.0
3 2023-02-23T00:00:00... GA    2023-02-1... 2023-02... 3061141.0 5298.0    42324.0
# i 3 more variables: new_deaths <chr>, new_historic_cases <chr>,
#   new_historic_deaths <chr>
```

```
head(cases_raw)
```

```
# A tibble: 6 × 10
  date_updated      state start_date end_date tot_cases new_cases tot_deaths
  <chr>            <chr> <chr>    <chr>    <chr>    <chr>    <chr>
1 2023-02-23T00:00:00... AZ    2023-02-1... 2023-02... 2434631.0 3716.0    33042.0
2 2022-12-22T00:00:00... LA    2022-12-1... 2022-12... 1507707.0 4041.0    18345.0
3 2023-02-23T00:00:00... GA    2023-02-1... 2023-02... 3061141.0 5298.0    42324.0
4 2023-03-30T00:00:00... LA    2023-03-2... 2023-03... 1588259.0 2203.0    18858.0
5 2023-02-02T00:00:00... LA    2023-01-2... 2023-02... 1548508.0 5725.0    18572.0
6 2023-03-23T00:00:00... LA    2023-03-1... 2023-03... 1580709.0 1961.0    18835.0
# i 3 more variables: new_deaths <chr>, new_historic_cases <chr>,
#   new_historic_deaths <chr>
```

We see exactly 1,000 rows. We should be seeing over 52×3 rows per state.

11. The reason you see exactly 1,000 rows is because CDC has a default limit. You can change this limit by adding `$limit=10000000000` to the request. Rewrite the previous request to ensure that you receive all the data. Then wrangle the resulting data frame to produce a data frame with columns `state`, `date` (should be the end date) and `cases`. Make sure the cases are numeric and the dates are in `Date` ISO-8601 format.

```
api <- "https://data.cdc.gov/resource/pwn4-m3yp.json"
cases <- request(api) |>
  req_url_query(`$limit` = 10000000000) |>
  req_user_agent("bst260-pset04/1.0") |>
  req_perform() |>
  resp_check_status() |>
  resp_body_json(simplifyVector = TRUE) |>
  tibble::as_tibble() |>
  dplyr::transmute(
```



```

    state = state,
    date = as.Date(end_date), # ISO-8601 Date
    cases = suppressWarnings(as.numeric(new_cases)) # numeric
  ) |>
dplyr::filter(
  state %in% c("DC", "PR"),
  !is.na(date), !is.na(cases)
)
head(cases)

```

A tibble: 6 × 3

	state	date	cases
	<chr>	<date>	<dbl>
1	AZ	2023-02-22	3716
2	LA	2022-12-21	4041
3	GA	2023-02-22	5298
4	LA	2023-03-29	2203
5	LA	2023-02-01	5725
6	LA	2023-03-22	1961

12. For 2020 and 2021, make a time series plot of cases per 100,000 versus time for each state. Stratify the plot by region name. Make sure to label you graph appropriately.

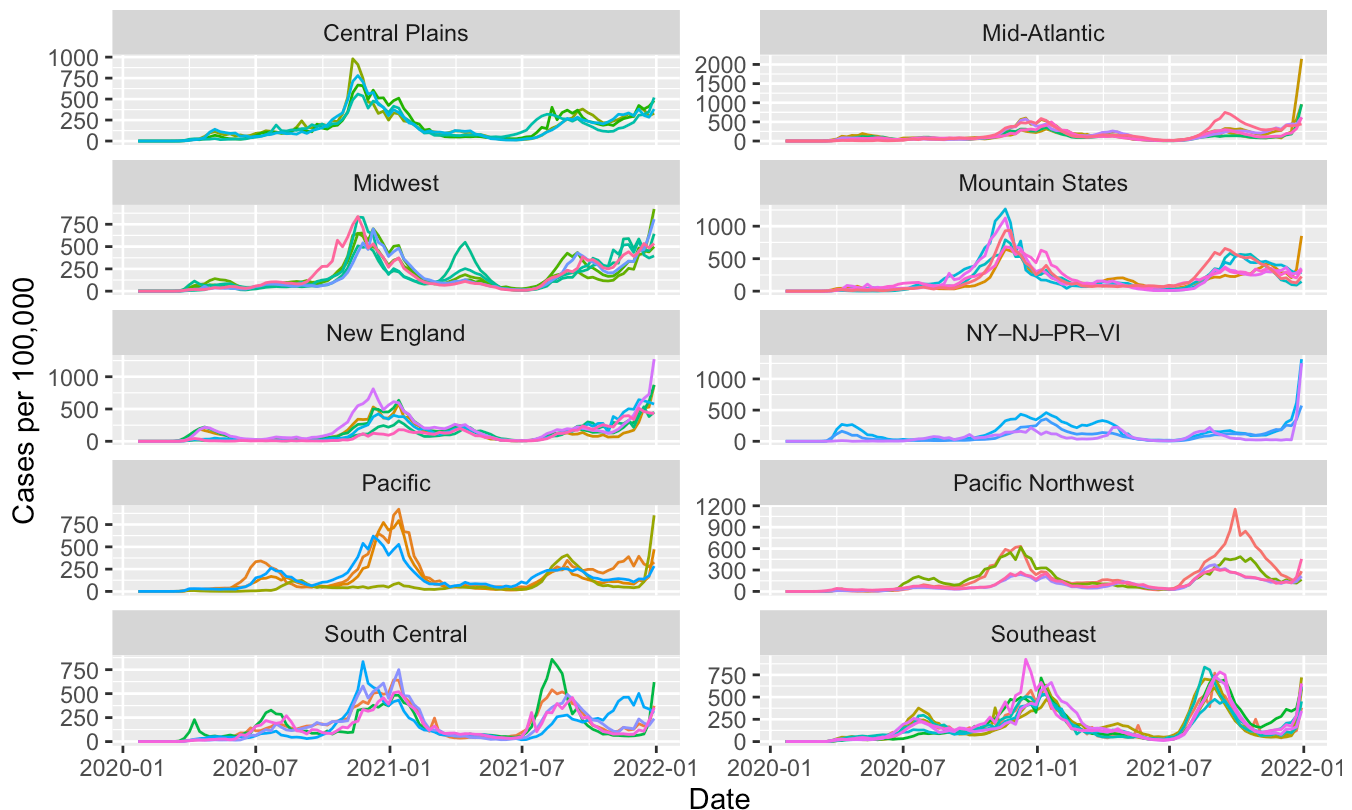
```

library(lubridate)
cases |>
  filter(date >= as.Date("2020-01-01"), date <= as.Date("2021-12-31")) %>%
  left_join(
    population |> select(state, state_name, pop_2020, pop_2021, region_name),
    by = "state"
  ) |>
  mutate(
    pop = if_else(year(date) == 2020, pop_2020, pop_2021),
    cases_per_100k = 1e5 * as.numeric(cases) / pop
  ) |>
  ggplot(aes(x = date, y = cases_per_100k, group = state, color = state)) +
  geom_line() +
  facet_wrap(~ region_name, ncol = 2, scales = "free_y") +
  guides(color = "none") +
  labs(
    title = "COVID-19 Cases per 100,000 by State (2020-2021)",
    subtitle = "Stratified by CDC PHS Region",
    x = "Date",
    y = "Cases per 100,000",
    caption = "Sources: CDC (pwn4-m3yp) and US Census PEP"
  )

```

COVID-19 Cases per 100,000 by State (2020–2021)

Stratified by CDC PHS Region



Sources: CDC (pwn4-m3yp) and US Census PEP

13. The dates in the `cases` dataset are stored as character strings. Use the **lubridate** package to properly parse the `date` column, then create a summary table showing the total COVID-19 cases by month and year for 2020 and 2021. The table should have columns for year, month (as month name), and total cases across all states. Order by year and month. Use the **knitr** package and `kable()` function to display the results as a formatted table.

```
library(knitr)

cases_monthly <- cases %>%
  mutate(date = if (inherits(date, "Date")) date else ymd(date)) %>%
  filter(year(date) %in% c(2020, 2021)) %>%
  group_by(
    year = year(date),
    month_num = month(date),
    month = month(date, label = TRUE, abbr = FALSE)
  ) %>%
  summarise(total_cases = sum(as.numeric(cases), na.rm = TRUE), .groups = "drop") %>%
  arrange(year, month_num) %>%
  select(year, month, total_cases)

kable(cases_monthly, caption = "Total COVID-19 Cases by Month and Year (2020–2021)")
```

Total COVID-19 Cases by Month and Year (2020–2021)

year	month	total_cases
2020	January	11
2020	February	68
2020	March	50335
2020	April	822648
2020	May	616691
2020	June	642552
2020	July	1977016
2020	August	1452393
2020	September	1401917
2020	October	1608932
2020	November	3887222
2020	December	6907540
2021	January	5649115
2021	February	2543964
2021	March	1928749
2021	April	1694189
2021	May	948953
2021	June	484817
2021	July	1120939
2021	August	3519407
2021	September	4960807
2021	October	2317854
2021	November	2289118
2021	December	5293391

14. The following URL provides additional COVID-19 data from the CDC in JSON format:

```
deaths_url <- "https://data.cdc.gov/resource/9bhhg-hcku.json"
```

Use **httr2** to download COVID-19 death data from this endpoint. Make sure to remove the default limit to get all available data. Create a clean dataset called **deaths** with columns **state**, **date**, and **deaths** (renamed from the original column name). Ensure dates are in proper Date format and deaths are numeric.

```
resp <- request(deaths_url) |>
  req_url_query(`$limit` = 10000000000) |>
```

```

req_perform()

stopifnot(resp_status(resp) == 200,
          grepl("json", resp_content_type(resp), ignore.case = TRUE))

raw_deaths <- resp_body_json(resp, simplifyVector = TRUE) |> as_tibble()

# Identify relevant columns robustly
state_col <- intersect(names(raw_deaths), c("state", "jurisdiction", "state_name"))[1]
date_col <- intersect(names(raw_deaths), c("end_date", "submission_date", "date", "as_o
# Prefer cumulative deaths if present; otherwise fall back to daily/new
deaths_col <- intersect(names(raw_deaths), c("tot_death", "total_deaths", "deaths", "new_

if (any(is.na(c(state_col, date_col, deaths_col)))) {
  stop("Required columns not found in deaths dataset. Columns present: ",
       paste(names(raw_deaths), collapse = ", "))
}

# Clean dataset: state, date (Date), deaths (numeric), get rid of the non-states
deaths <- raw_deaths |> mutate(state = state, date = as.Date(end_date), deaths = as.numeri

# "United States" %in% deaths$state
head(deaths)

```

A tibble: 6 × 3

	state	date	deaths
	<chr>	<date>	<dbl>
1	Alabama	2023-09-23	21520
2	Alabama	2023-09-23	19
3	Alabama	2023-09-23	46
4	Alabama	2023-09-23	142
5	Alabama	2023-09-23	267
6	Alabama	2023-09-23	416

15. Using the `deaths` dataset you created, make a bar plot showing the total COVID-19 deaths by state. Show only the top 10 states with the highest death counts. Order the bars from highest to lowest and use appropriate labels and title.

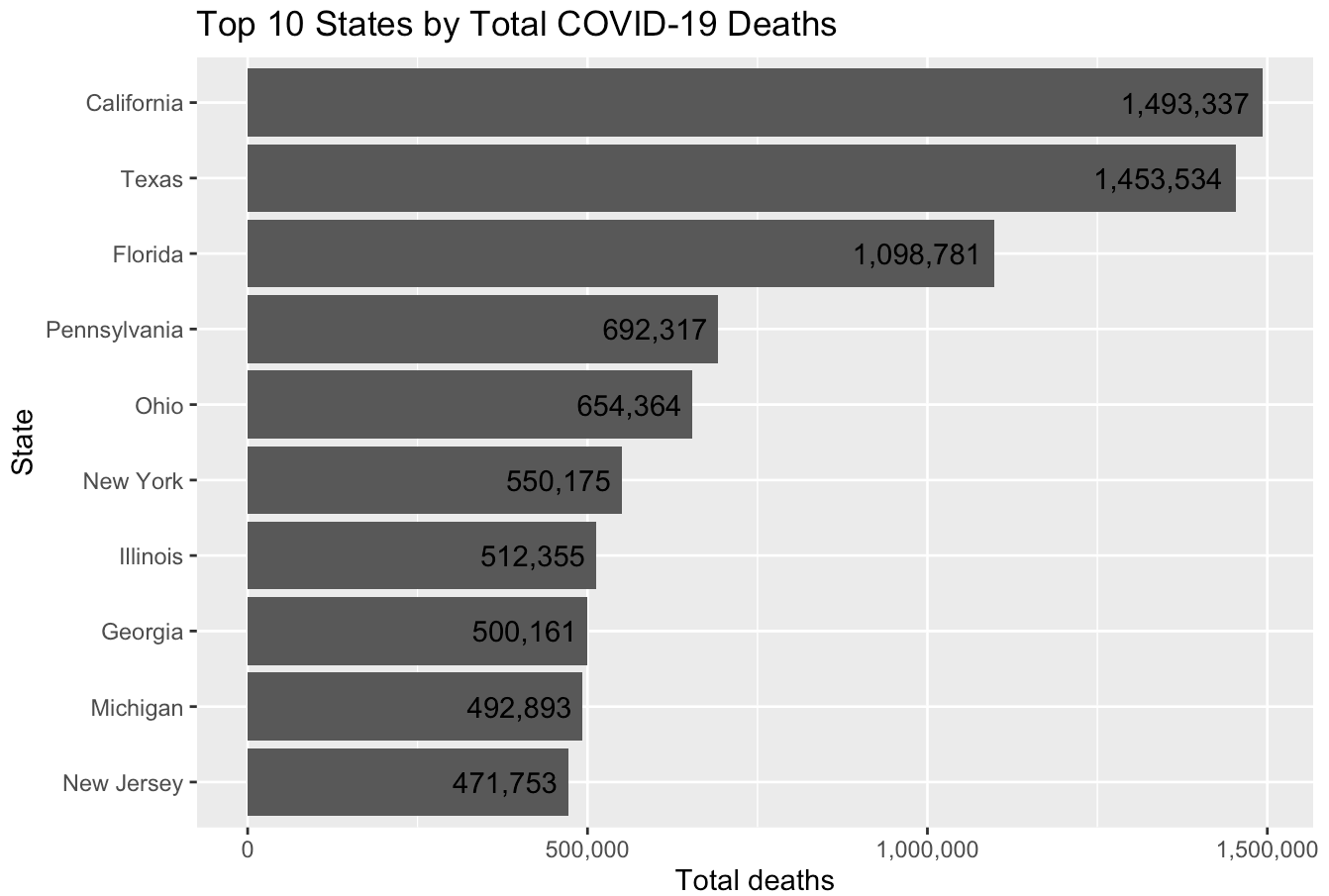
```

# Sum deaths by state, keep top 10
top10_deaths <- deaths |> group_by(state) |> summarise(total_deaths = sum(deaths, na.rm =

# Bar plot (highest to lowest)
ggplot(top10_deaths, aes(x = reorder(state, total_deaths), y = total_deaths)) +
  geom_col() +
  coord_flip() + geom_text(aes(label = scales::comma(total_deaths)),
  hjust = 1.1) +
  scale_y_continuous(labels = scales::label_comma()) +
  labs(
    title = "Top 10 States by Total COVID-19 Deaths",
    x = "State",
    y = "Total deaths",

```

```
caption = "Source: CDC (9bhg-hcku)"  
)
```



Source: CDC (9bhg-hcku)