

EV Power - Lab 4 Project Report

Project 4: EV Power - Colin Choe

Part 0: libraries

```
library(knitr)
knitr::opts_chunk$set(dev = "ragg_png", dpi = 200, fig.width = 7, fig.height =
4.3, fig.align = "center")
library(tidyverse)
```

```
— Attaching core tidyverse packages ————— tidyverse 2.0.0
—
✓ dplyr      1.1.4      ✓ readr      2.1.5
✓ forcats    1.0.0      ✓ stringr    1.5.2
✓ 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
```

```
library(readr)
library(dplyr)
library(ggplot2)
library(stringr)
```

Part 1: Defining Research Question

Chosen Question: Which states increased their renewable energy use most from 2022 to 2023?

Part 2: Data Preparation and Cleaning

```
renew_2022 <- read_csv("data/renew-use-2022.csv")
```

Rows: 260 Columns: 3

	Column	specification
--	--------	---------------

```
Delimiter: ","
chr (3): State, Energy_Source, Renewable_Use_2022
```

```
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.
```

```
renew_2023 <- read_csv("data/renew-use-2023.csv")
```

```
Rows: 260 Columns: 3
— Column specification
```

```
Delimiter: ","
chr (3): State, Energy_Source, Renewable_Use_2023
```

```
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.
```

```
clean_numbers <- function(x) {
  x |>
    str_replace_all("[^0-9.]", "") |> # keep only digits and periods
    as.numeric()
}

names(renew_2022) <- names(renew_2022) |>
  tolower() |>
  str_replace_all("[- ]", "_")

names(renew_2023) <- names(renew_2023) |>
  tolower() |>
  str_replace_all("[- ]", "_")

renew_2022 <- renew_2022 %>%
  rename(state = state, source = energy_source, renewable_use =
renewable_use_2022) %>%
  mutate(
    state = str_to_upper(state),
    source = str_to_title(source),
    renewable_use = clean_numbers(renewable_use)
  )

renew_2023 <- renew_2023 %>%
  rename(state = state, source = energy_source, renewable_use =
renewable_use_2023) %>%
  mutate(
```

```

state = str_to_upper(state),
source = str_to_title(source),
renewable_use = clean_numbers(renewable_use)
)

```

Part 3: Joining / Pivoting Datasets for Analysis

```

renew_2022_state <- renew_2022 %>%
  group_by(state) %>%
  summarise(total_renew_2022 = sum(renewable_use, na.rm = TRUE))

head(renew_2022_state, 10)

```

```

# A tibble: 10 × 2
  state total_renew_2022
  <chr>         <dbl>
1 AK           10410
2 AL          232035
3 AR           90824
4 AZ          101214
5 CA          880995
6 CO          114918
7 CT           49084
8 DC           2623
9 DE           7402
10 FL         304605

```

```

renew_2023_state <- renew_2023 %>%
  group_by(state) %>%
  summarise(total_renew_2023 = sum(renewable_use, na.rm = TRUE))

head(renew_2023_state, 10)

```

```

# A tibble: 10 × 2
  state total_renew_2023
  <chr>         <dbl>
1 AK           10088
2 AL          222189
3 AR           87277
4 AZ          108445
5 CA          1065179
6 CO          115062
7 CT           48983
8 DC           2796

```

```

9 DE      8040
10 FL     286307

```

```

renew_compare <- renew_2022_state %>%
  inner_join(renew_2023_state, by = "state")

renew_compare <- renew_compare %>%
  mutate(
    absolute_change = total_renew_2023 - total_renew_2022,
    percent_change = ((total_renew_2023 - total_renew_2022) / total_renew_2022)
    * 100
  )

renew_compare <- renew_compare %>%
  arrange(desc(absolute_change))

head(renew_compare, 10)

```

```

# A tibble: 10 × 5
  state total_renew_2022 total_renew_2023 absolute_change percent_change
<chr>      <dbl>          <dbl>          <dbl>          <dbl>
1 CA          880995        1065179        184184          20.9
2 US          8107355        8187317         79962           0.986
3 TX          751680         791210         39530           5.26
4 AZ          101214         108445          7231           7.14
5 NY          269883         272968          3085           1.14
6 NM           77285          80278           2993           3.87
7 MD           51256          53711           2455           4.79
8 UT           37369          39675           2306           6.17
9 NV           72735          74879           2144           2.95
10 IN          170987         172891          1904           1.11

```

Part 4: Mapping Visualization

```
library(sf)
```

```
Linking to GEOS 3.13.0, GDAL 3.8.5, PROJ 9.5.1; sf_use_s2() is TRUE
```

```
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(maps)
```

Attaching package: 'maps'

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

map

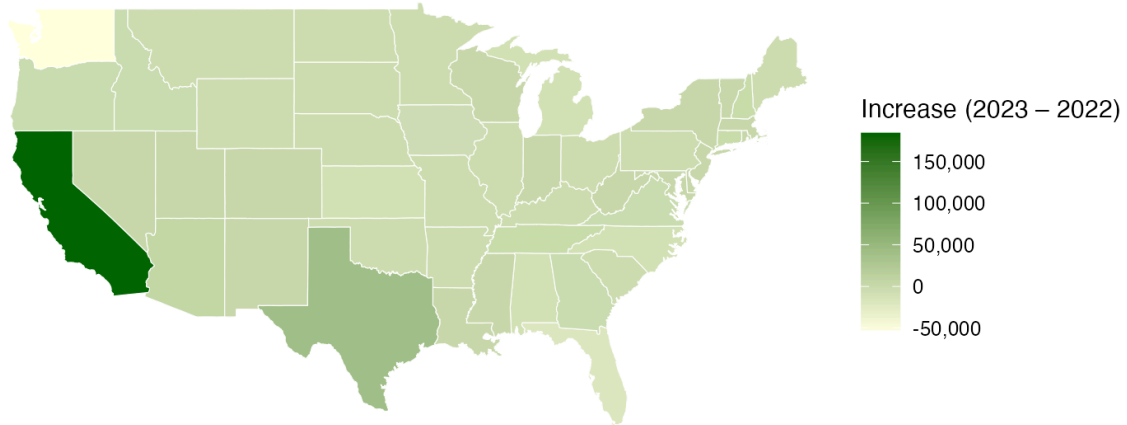
```
states_sf <- st_as_sf(map("state", plot = FALSE, fill = TRUE))

renew_compare_clean <- renew_compare %>%
  mutate(state_name = tolower(state.name[match(toupper(state), state.abb)]))

map_data <- states_sf %>%
  left_join(renew_compare_clean, by = c("ID" = "state_name"))

ggplot(map_data) +
  geom_sf(aes(
    fill = absolute_change,
    color = "white",
    linewidth = 0.2) +
  scale_fill_gradient(
    low = "lightyellow",
    high = "darkgreen",
    name = "Increase (2023 - 2022)",
    labels = comma) +
  labs(
    title = "Change in Renewable Energy Use by State (2022 to 2023)",
    caption = "Data: renew-use-2022.csv and renew-use-2023.csv"
  ) +
  theme_void() +
  theme(
    legend.position = "right",
    plot.title = element_text(size = 12, face = "bold")
  )
```

Change in Renewable Energy Use by State (2022 to 2023)



Data: renew-use-2022.csv and renew-use-2023.csv

Part 5: Analysis

Analysis: The map shows that renewable energy use increased the most in states like California, which clearly stands out from the rest of the country. Most other states saw smaller or steady changes, especially in the Midwest and South. This suggests that while renewable energy is growing overall, most of the progress is driven by a few key states with stronger clean energy policies. It helps visualize where that growth is happening and how uneven it still is across the U.S.