

EV Power - Lab 4 Project Report

Part 0: libraries

```
library(sf)
library(rnaturalearth)
library(maps)
library(dplyr)
library(tidyr)
library(stringr)
library(ggplot2)
```

Part 1: Defining Research Question

Chosen Question: How has the share of renewable energy changed from 2021–2023 across states?
What is the share of electricity that comes from clean sources by state?

Part 2: Data Preparation and Cleaning

```
renew2021 <- read.csv("data/renew-use-2021.csv")
renew2022 <- read.csv("data/renew-use-2022.csv")
renew2023 <- read.csv("data/renew-use-2023.csv")
total2021 <- read.csv("data/total-use-2021.csv")
total2022 <- read.csv("data/total-use-2022.csv")
total2023 <- read.csv("data/total-use-2023.csv")
```

```
renew2021 <- renew2021|>
  mutate(Renewable_Use_2021 = as.numeric(str_extract(Renewable_Use_2021, "\
\d+\\.?\d*")))
renew2022 <- renew2022|>
  mutate(Renewable_Use_2022 = as.numeric(str_extract(Renewable_Use_2022, "\
\d+\\.?\d*")))
renew2023 <- renew2023|>
  mutate(Renewable_Use_2023 = as.numeric(str_extract(Renewable_Use_2023, "\
\d+\\.?\d*")))
```

```
total2021 <- total2021|>
  mutate(Energy_Source = str_replace_all(Energy_Source, "[^A-Za-z]", ""))
total2022$Energy_Source <- total2021$Energy_Source
total2023$Energy_Source <- total2021$Energy_Source
```

```

standardize_state <- function(df) {
  df <- df |>
    mutate(State = str_to_upper(State))
  return(df)
}
renew2021 <- standardize_state(renew2021)
renew2022 <- standardize_state(renew2022)
renew2023 <- standardize_state(renew2023)

```

Part 3: Joining / Pivoting Datasets for Analysis

```

renew_2 <- full_join(x=renew2021, y=renew2022, by=c("State", "Energy_Source"))
renew_all <- full_join(x=renew_2, y=renew2023, by=c("State", "Energy_Source"))
head(renew_all)

```

	State	Energy_Source	Renewable_Use_2021	Renewable_Use_2022	Renewable_Use_2023
1	AK	Biomass	3153	3846	3404
2	AK	Geothermal	186	186	186
3	AK	Hydropower	5763	5846	6051
4	AK	Solar Energy	45	57	67
5	AK	Wind Energy	451	475	380
6	AL	Biomass	198543	193932	189040

Part 4: Mapping Visualization

```

us_states <- ne_states(country = "United States of America", returnclass =
"sf")
renew_state_summary <- renew_all |>
  group_by(State) |>
  summarize(Renewable_Use_2021 = sum(Renewable_Use_2021, na.rm = TRUE),
Renewable_Use_2023 = sum(Renewable_Use_2023, na.rm = TRUE))
us_joined <- us_states |>
  left_join(renew_state_summary, by = join_by(postal == State))|>
  mutate(Change = Renewable_Use_2023 - Renewable_Use_2021)
renew_state_summary

```

```

# A tibble: 52 × 3
  State Renewable_Use_2021 Renewable_Use_2023
  <chr>          <dbl>          <dbl>
1 AK              9598             10088
2 AL          239816             222189
3 AR              89714             87277
4 AZ              99266             108445
5 CA          810020             1065179

```

```

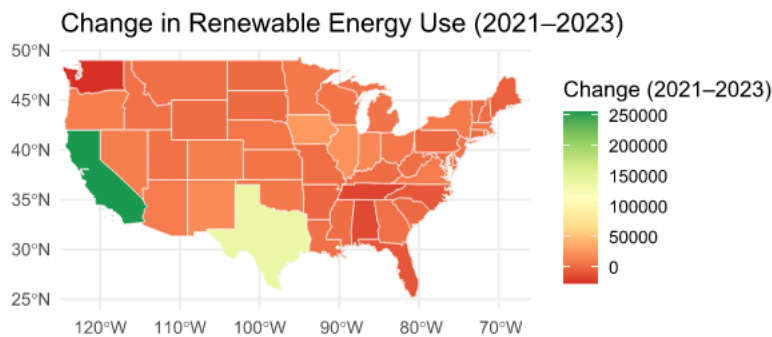
6 CO      103956      115062
7 CT      49306      48983
8 DC       2487       2796
9 DE       7151       8040
10 FL     297290     286307
# i 42 more rows

```

```

ggplot(us_joined) +
  geom_sf(aes(fill = Change), color = "white") +
  scale_fill_distiller(
    palette = "RdYlGn",
    direction = 1,
    name = "Change (2021–2023)") +
  labs(title = "Change in Renewable Energy Use (2021–2023)") +
  coord_sf(xlim = c(-125, -66), ylim = c(24, 50), expand = FALSE) +
  theme_minimal()

```



The map shows that renewable energy use has gone up in most states between 2021 and 2023, but the change is not even everywhere as when some states in the West and Northeast show bigger growth in renewables, some Southern and Midwestern states have smaller changes or almost none. This might be because of differences in how much each state invests in renewable projects or how heavily they rely on other energy sources like coal or gas. When looking at the color scale, we can easily see states like California and Texas are leading in renewable energy growth and fossil-fuel-heavy regions (e.g., parts of the South) lag behind.