# **EV Adoption, Electricity Prices, and Renewable Energy Trends**

Ella Proctor

2025-10-29

```
options(repos = c(CRAN = "https://cloud.r-project.org"))
```

### **Example Solution 1**

### Part 0: libraries

library(tidyr)
library(stringr)

```
setwd("~/Documents/stat133/ev-power-ellaproctor/data")
options(repos = c(CRAN = "https://cloud.r-project.org"))
library(tidyverse)
-- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
v dplyr
        1.1.4 v readr 2.1.5
v forcats 1.0.1 v stringr 1.5.2
v ggplot2 4.0.0 v tibble 3.3.0
v lubridate 1.9.4 v tidyr 1.3.1
v purrr
         1.1.0
-- Conflicts ----- tidyverse_conflicts() --
x dplyr::filter() masks stats::filter()
x dplyr::lag()
                  masks stats::lag()
i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become
library(readr)
library(dplyr)
```

```
library(ggthemes)
library(maps)

Attaching package: 'maps'
The following object is masked from 'package:purrr':
    map

library(sf)
```

Linking to GEOS 3.13.0, GDAL 3.8.5, PROJ 9.5.1; sf\_use\_s2() is TRUE

### Part 1: Defining Research Question

Chosen Question: Are there more EV registrations in states with higher average electricity prices?

Some other questions: Do states with more volatile energy prices invest more in renewable sources over time? Do coastal states tend to have more EV registration and renewable energy adoption? Has EV adoption or renewable energy use grown faster?

### Part 2: Data Preparation and Cleaning

```
renew22 <- read_csv("renew-use-2022.csv")</pre>
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.
renew23 <- read_csv("renew-use-2023.csv")</pre>
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.
total21 <- read_csv("total-use-2021.csv")</pre>
Rows: 5 Columns: 53
-- Column specification ------
Delimiter: ","
chr (1): Energy_Source
dbl (52): AK, AL, AR, AZ, CA, CO, CT, DC, DE, FL, GA, HI, IA, ID, IL, IN, KS...
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.
total22 <- read_csv("total-use-2022.csv")</pre>
Rows: 5 Columns: 53
-- Column specification ------
Delimiter: ","
chr (1): Energy_Source
dbl (52): AK, AL, AR, AZ, CA, CO, CT, DC, DE, FL, GA, HI, IA, ID, IL, IN, KS...
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.
```

```
total23 <- read_csv("total-use-2023.csv")</pre>
Rows: 5 Columns: 53
-- Column specification -----
Delimiter: ","
chr (1): Energy_Source
dbl (52): AK, AL, AR, AZ, CA, CO, CT, DC, DE, FL, GA, HI, IA, ID, IL, IN, KS...
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.
evs <- read_csv("ev-registrations-by-state-2023.csv", skip = 1, show_col_types = FALSE)
New names:
* `` -> `...1`
* `` -> `...2`
price <- read_csv("av-energy-price-2021-2023.csv")</pre>
Rows: 54 Columns: 1
-- Column specification ------
Delimiter: ","
chr (1): Total energy average price, dollars per million Btu,,,
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 names <- function(df) {</pre>
 names(df) <- names(df) %>%
   str_to_lower() %>%
   str_replace_all("[^a-z0-9]+", "_") %>%
   str_replace_all("_$", "")
 return(df)
}
renew21 <- clean_names(renew21)</pre>
renew22 <- clean_names(renew22)</pre>
renew23 <- clean_names(renew23)</pre>
total21 <- clean_names(total21)</pre>
```

```
total22 <- clean_names(total22)</pre>
total23 <- clean_names(total23)</pre>
evs <- clean_names(evs)</pre>
price <- clean_names(price)</pre>
names (renew21)
[1] "state"
                            "energy_source"
                                                   "renewable_use_2021"
names(evs)
[1] "_1" "_2"
standardize_states <- function(df) {</pre>
  if ("state" %in% names(df)) {
    df <- df %>%
      mutate(
         state = str_to_title(state),
        state = str_trim(state)
  }
  return(df)
}
renew21 <- standardize_states(renew21)</pre>
renew22 <- standardize_states(renew22)</pre>
renew23 <- standardize_states(renew23)</pre>
total21 <- standardize_states(total21)</pre>
total22 <- standardize_states(total22)</pre>
total23 <- standardize_states(total23)</pre>
evs <- standardize_states(evs)</pre>
price <- standardize_states(price)</pre>
renew21 <- renew21 %>% mutate(year = 2021)
renew22 <- renew22 %>% mutate(year = 2022)
renew23 <- renew23 %>% mutate(year = 2023)
total21 <- total21 %>% mutate(year = 2021)
total22 <- total22 %>% mutate(year = 2022)
total23 <- total23 %>% mutate(year = 2023)
```

colnames(evs) <- c("state", "ev\_registrations\_raw")</pre>

```
evs <- evs %>%
  mutate(
    state = str_to_title(state),
    ev_registrations = str_replace_all(ev_registrations_raw, "[^0-9]", ""),
    ev_registrations = as.numeric(ev_registrations)
) %>%
  select(state, ev_registrations)

# checks state names match across datasets
setdiff(unique(renew21$state), unique(evs$state))

[1] "Ak" "Al" "Ar" "Az" "Ca" "Co" "Ct" "Dc" "De" "Fl" "Ga" "Hi" "Ia" "Id" "Il"
[16] "In" "Ks" "Ky" "La" "Ma" "Md" "Me" "Mi" "Mn" "Mo" "Ms" "Mt" "Nc" "Nd" "Ne"
[31] "Nh" "Nj" "Nm" "Nv" "Ny" "Oh" "Ok" "Or" "Pa" "Ri" "Sc" "Sd" "Tn" "Tx" "Ut"
[46] "Va" "Vt" "Wa" "Wi" "Wv" "Wy" "Us"
```

### Part 3: Joining / Pivoting Datasets for Analysis

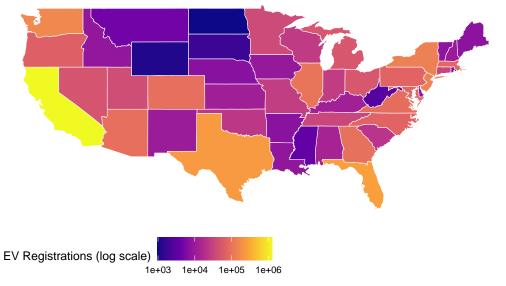
```
setwd("~/Documents/stat133/ev-power-ellaproctor/data")
evs <- read_csv("ev-registrations-by-state-2023.csv", skip = 1)</pre>
New names:
Rows: 53 Columns: 2
-- Column specification
------ Delimiter: "," chr
(2): \ldots 1, \ldots 2
i Use `spec()` to retrieve the full column specification for this data. i
Specify the column types or set `show_col_types = FALSE` to quiet this message.
* `` -> `...1`
* `` -> `...2`
colnames(evs) <- c("state", "ev_registrations_raw")</pre>
evs <- evs %>%
 mutate(
   state = str_to_title(state),
   ev_registrations = str_replace_all(ev_registrations_raw, "[^0-9]", ""),
   ev_registrations = as.numeric(ev_registrations)
```

```
) %>%
  select(state, ev_registrations)
head(evs)
# A tibble: 6 x 2
  state
         ev_registrations
  <chr>
                        <dbl>
1 State
                           NA
2 Alabama
                        13047
3 Alaska
                        2697
4 Arizona
                        89798
5 Arkansas
                         7108
6 California
                      1256646
renew21 <- renew21 %>%
 mutate(
   state = str_to_upper(state),
   renewable_use_2021 = str_replace_all(renewable_use_2021, "[^0-9.]", ""),
   renewable_use_2021 = as.numeric(renewable_use_2021)
  )
total21_long <- total21 %>%
 pivot_longer(
    cols = -energy_source,
   names_to = "state",
   values_to = "total_use_2021"
 ) %>%
 mutate(
    state = str_to_upper(state),
   total_use_2021 = as.numeric(total_use_2021)
  )
renew_state_2021 <- renew21 %>%
  group_by(state) %>%
  summarise(total_renewable_2021 = sum(renewable_use_2021, na.rm = TRUE))
total_2021_summary <- total21_long %>%
  filter(str_detect(energy_source, regex("total_renewable_energy", ignore_case = TRUE))) %>%
  select(state, total_use_2021)
```

```
combined 2021 <- left_join(total 2021 summary, renew_state 2021, by = "state") %%
 mutate(renewable_share_2021 = (total_renewable_2021 / total_use_2021) * 100)
combined 2021 <- combined 2021 %>%
 mutate(state = state.name[match(state, state.abb)]) %>%
 filter(!is.na(state)) %>%
 left_join(evs, by = "state") %>%
 mutate(
   region = case_when(
      state %in% c("California", "Oregon", "Washington") ~ "West Coast",
      state %in% c("New York", "Massachusetts", "Maine",
                   "New Jersey", "Maryland", "Virginia",
                   "Florida", "Georgia") ~ "East Coast",
     TRUE ~ "Inland"
   )
 )
glimpse(combined 2021)
```

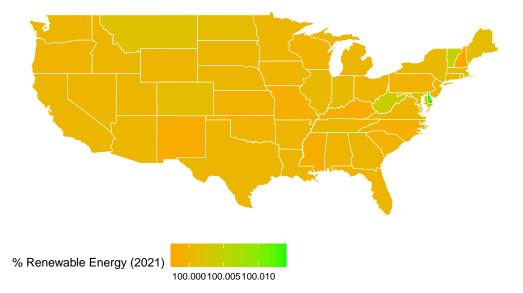
### Part 4: Mapping Visualization

### **Electric Vehicle Registrations by State (2023)**



Data: EV Power - stat 133

### Renewable Energy Share by State (2021)

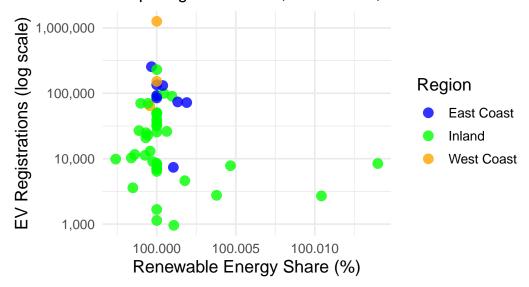


Data: EV Power - stat 133

```
#coast vs inland
ggplot(combined_2021, aes(x = renewable_share_2021, y = ev_registrations, color = region)) +
    geom_point(size = 3, alpha = 0.8) +
    scale_y_log10(labels = scales::comma) +
    scale_color_manual(values = c("blue", "green", "orange")) +
    labs(title = "EV Adoption vs Renewable Energy Share (2021-2023)",
        subtitle = "Comparing West Coast, East Coast, and Inland States",
        x = "Renewable Energy Share (%)",
        y = "EV Registrations (log scale)",
        color = "Region") +
    theme_minimal(base_size = 13)
```

### EV Adoption vs Renewable Energy Share (20)

Comparing West Coast, East Coast, and Inland States



Part 5: Final Deliverable

### 1. Introduction

This report explores how electric-vehicle (EV) adoption interacts with the electricity prices and renewable-energy use in different states. My guiding question is whether coastal states show stronger growth in EV adoption compared to inland regions.

### 2. Data and Maps

This analysis used three different datasets: EV Registrations (by state), Renewable Energy Use (energy consumption by source) and Total Energy Use (by state). Data cleaning inluded removing units and symbols, computing each state's renewable energy share, classifying regions into inland and different coasts, etc.

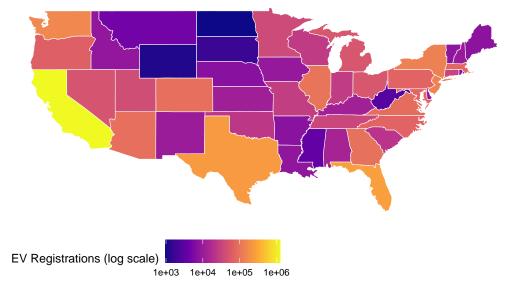
#### head(combined\_2021)

```
2 alabama
                     239817
                                           239816
                                                                 100.0
3 arkansas
                     89714
                                           89714
                                                                 100
                      99266
                                           99266
                                                                 100
4 arizona
5 california
                     810020
                                          810020
                                                                 100
6 colorado
                     103955
                                           103956
                                                                 100.
# i 2 more variables: ev_registrations <dbl>, region <chr>
```

### head(evs)

```
# A tibble: 6 x 2
 state
             ev_registrations
 <chr>>
                         <dbl>
1 State
                            NA
2 Alabama
                         13047
3 Alaska
                         2697
                         89798
4 Arizona
5 Arkansas
                         7108
6 California
                      1256646
```

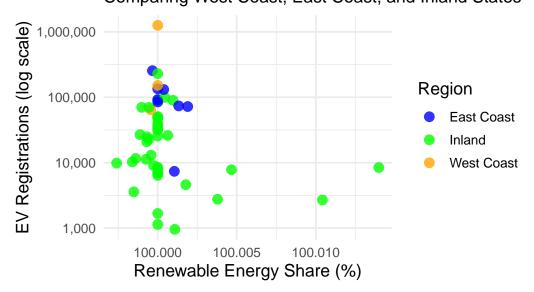
### **Electric Vehicle Registrations by State (2023)**



Data: EV Power - stat 133

```
ggplot(combined_2021, aes(x = renewable_share_2021, y = ev_registrations, color = region)) +
    geom_point(size = 3, alpha = 0.8) +
    scale_y_log10(labels = scales::comma) +
    scale_color_manual(values = c("blue", "green", "orange")) +
    labs(title = "EV Adoption vs Renewable Energy Share (2021-2023)",
        subtitle = "Comparing West Coast, East Coast, and Inland States",
        x = "Renewable Energy Share (%)",
        y = "EV Registrations (log scale)",
        color = "Region") +
    theme_minimal(base_size = 13)
```

## EV Adoption vs Renewable Energy Share (20) Comparing West Coast, East Coast, and Inland States



### 3. Analysis

The maps and scatterplot revels several trends, but the most important one to answer my question is that we can see high EV adoption clusters on the West Coast. Inland states tend to have lower EV registrations even when renewable-energy share is high (Iowa, South Dakota), suggesting that charging-network access and income levels influence adoption more. We can also see a positive correlation between renewable-energy share and EV registrations, which implies that states investing in clean energy infrastructure also foster consumer adoption of EVs. In terms of pricing, west coast states like California and Washington had high renewable-energy shares and high electricity prices which shows a correlation (people switch to renewables when electricity prices are high). Of course, with all of this data we cannot just assume causation without further digging, because there may be some confounding factors like cultures in different areas that affect these results.