

EV Power - Lab 4 Project Report

Part 0: libraries

```
library(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    ✓ tidyrr    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(stringr)
library(dplyr)
library(ggplot2)
library(sf)
```

```
Warning: package 'sf' was built under R version 4.5.2
```

```
Linking to GEOS 3.13.1, GDAL 3.11.4, PROJ 9.7.0; sf_use_s2() is TRUE
```

```
library(maps)
```

```
Warning: package 'maps' was built under R version 4.5.2
```

```
Attaching package: 'maps'
```

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

```
map
```

Part 1: Defining Research Question

Chosen Question: Do states with more EV registrations tend to have cheaper electricity?

Part 2: Data Preparation and Cleaning

```
# fixing messy data from csv: av-energy  
  
# read in the messy CSV (skip extra rows if needed)  
energy_cost_raw <- read_csv("data/av-energy-price-2021-2023.csv", skip = 2)
```

```
Rows: 52 Columns: 1  
— Column specification
```

```
Delimiter: ","  
chr (1): State,2021,2022,2023
```

```
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.
```

```
energy_cost_raw
```

```
# A tibble: 52 × 1  
  `State,2021,2022,2023`  
  <chr>  
1 AK,$20.03 per MMBtu,$27.33,$23.84 est.  
2 AL,about 17.85 USD,23.37 USD,≈21.11  
3 AR,$18.42,$23.84 per MMBtu,$21.76  
4 AZ,≈25.07,31.72 USD,about 30.28  
5 CA,$28.44,$37.35,$35.72 per MMBtu  
6 CO,20.64 USD,≈25.85,23.85  
7 CT,about $25.85,$33.15,$32.32 est.  
8 DC,≈25.67,$30.84,about 32.28 USD  
9 DE,$21.83,$27.74 per MMBtu,$26.70  
10 FL,≈22.53,29.35 USD,$28.12  
# i 42 more rows
```

```
# Split the one column into four depending on where there is a comma  
energy_cost_raw <- separate(  
  energy_cost_raw,
```

```

    col = 1,
    into = c("State", "2021", "2022", "2023"),
    sep = ","
)

# preview data
head(energy_cost_raw)

```

```

# A tibble: 6 × 4
  State `2021`      `2022`      `2023`
  <chr> <chr>       <chr>       <chr>
1 AK   $20.03 per MMBtu $27.33     $23.84 est.
2 AL   about 17.85 USD  23.37 USD   ≈21.11
3 AR   $18.42          $23.84 per MMBtu $21.76
4 AZ   ≈25.07          31.72 USD     about 30.28
5 CA   $28.44          $37.35      $35.72 per MMBtu
6 CO   20.64 USD       ≈25.85      23.85

```

```

# clean the table using regex
energy_cost <- energy_cost_raw |>
  mutate(across(
    c(`2021`, `2022`, `2023`),
    ~ as.numeric(str_extract(.x, "\\\d+\\.?\\d*")))
  ))
head(energy_cost)

```

```

# A tibble: 6 × 4
  State `2021` `2022` `2023`
  <chr>  <dbl>   <dbl>   <dbl>
1 AK      20.0    27.3   23.8
2 AL      17.8    23.4   21.1
3 AR      18.4    23.8   21.8
4 AZ      25.1    31.7   30.3
5 CA      28.4    37.4   35.7
6 CO      20.6    25.8   23.8

```

```

# fixing messy data from csv: messy-ev-registrations
ev_reg <- read_csv("data/ev-registrations-by-state-2023.csv", skip = 2)

```

```

Rows: 52 Columns: 2
— Column specification —
Delimiter: ","
chr (2): STATE, Count-EVs

```

```
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.
```

```
ev_reg
```

```
# A tibble: 52 × 2  
  STATE          `Count-EVs`  
  <chr>         <chr>  
1 Alabama       #13047  
2 Alaska        ~2697  
3 Arizona       89798  
4 Arkansas      7108 EVs  
5 California    1256646  
6 Colorado      90083  
7 Connecticut   EVs 31557  
8 Delaware      8435  
9 District of Columbia 8066  
10 Florida      254878  
# i 42 more rows
```

```
# cleaning ev_reg  
names(ev_reg) <- names(ev_reg) |>  
  str_trim() |>  
  str_to_lower() |>  
  str_replace_all("[ -_]", " ") |>  
  str_replace_all("\\\\s+", "_") |>  
  str_remove_all("\\\\(.*)\\\\)") # remove units like (2023)  
  
head(ev_reg)
```

```
# A tibble: 6 × 2  
  state      count_evs  
  <chr>     <chr>  
1 Alabama    #13047  
2 Alaska     ~2697  
3 Arizona    89798  
4 Arkansas   7108 EVs  
5 California 1256646  
6 Colorado   90083
```

```
ev_reg <- ev_reg |>
  mutate(count_evs = str_extract(count_evs, "\\d+") %>% as.numeric())
head(ev_reg)
```

```
# A tibble: 6 × 2
  state      count_evs
  <chr>        <dbl>
1 Alabama      13047
2 Alaska       2697
3 Arizona     89798
4 Arkansas     7108
5 California   1256646
6 Colorado     90083
```

```
# choosing the year 2023 because only have one year of data for EV
registrations

# state names are not equal, so must first change state names so they can be
joined
# adding DC to list of state names
state_fulls <- c(datasets::state.name, "District of Columbia")
state_fulls
```

```
[1] "Alabama"          "Alaska"           "Arizona"
[4] "Arkansas"         "California"       "Colorado"
[7] "Connecticut"       "Delaware"          "Florida"
[10] "Georgia"          "Hawaii"           "Idaho"
[13] "Illinois"         "Indiana"          "Iowa"
[16] "Kansas"           "Kentucky"         "Louisiana"
[19] "Maine"             "Maryland"          "Massachusetts"
[22] "Michigan"         "Minnesota"        "Mississippi"
[25] "Missouri"         "Montana"          "Nebraska"
[28] "Nevada"           "New Hampshire"    "New Jersey"
[31] "New Mexico"        "New York"          "North Carolina"
[34] "North Dakota"      "Ohio"              "Oklahoma"
[37] "Oregon"            "Pennsylvania"     "Rhode Island"
[40] "South Carolina"    "South Dakota"     "Tennessee"
[43] "Texas"              "Utah"              "Vermont"
[46] "Virginia"          "Washington"       "West Virginia"
[49] "Wisconsin"         "Wyoming"          "District of Columbia"
```

```
state_abbs <- c(datasets::state.abb, "DC")
state_abbs
```

```
[1] "AL" "AK" "AZ" "AR" "CA" "CO" "CT" "DE" "FL" "GA" "HI" "ID" "IL" "IN"
"IA"
[16] "KS" "KY" "LA" "ME" "MD" "MA" "MI" "MN" "MS" "MO" "MT" "NE" "NV" "NH"
"NJ"
[31] "NM" "NY" "NC" "ND" "OH" "OK" "OR" "PA" "RI" "SC" "SD" "TN" "TX" "UT"
"VT"
[46] "VA" "WA" "WV" "WI" "WY" "DC"
```

```
state_lookup <- tibble(state_fulls, state_abbs) # maybe
state_lookup # maybe
```

```
# A tibble: 51 × 2
  state_fulls state_abbs
  <chr>        <chr>
1 Alabama      AL
2 Alaska       AK
3 Arizona      AZ
4 Arkansas     AR
5 California   CA
6 Colorado     CO
7 Connecticut CT
8 Delaware    DE
9 Florida      FL
10 Georgia     GA
# i 41 more rows
```

```
# matching state abbreviations to full names and replacing full names with
abbrvs
matching <- match(energy_cost$State, state_abbs)
matching
```

```
[1]  2  1  4  3  5  6  7 51  8  9 10 11 15 12 13 14 16 17 18 21 20 19 22 23
25
[26] 24 26 33 34 27 29 30 31 28 32 35 36 37 38 39 40 41 42 43 44 46 45 47 49
48
[51] 50 NA
```

```
energy_cost$State <- ifelse(is.na(matching), ev_reg$State,
state_fulls[matching])
```

```
Error in ans[ypos] <- rep(yes, length.out = len)[ypos]: replacement has length
zero
```

```
head(energy_cost)
```

```
# A tibble: 6 × 4
  State `2021` `2022` `2023`
  <chr>   <dbl>   <dbl>   <dbl>
1 AK      20.0    27.3   23.8
2 AL      17.8    23.4   21.1
3 AR      18.4    23.8   21.8
4 AZ      25.1    31.7   30.3
5 CA      28.4    37.4   35.7
6 CO      20.6    25.8   23.8
```

```
# subsetting only 2023 year of energy costs
energy_cost_2023 <- energy_cost[,c(1,4)]  
  
# renaming columns
colnames(energy_cost_2023) <- c("state", "cost")
energy_cost_2023
```

```
# A tibble: 52 × 2
  state   cost
  <chr>   <dbl>
1 AK      23.8
2 AL      21.1
3 AR      21.8
4 AZ      30.3
5 CA      35.7
6 CO      23.8
7 CT      32.3
8 DC      32.3
9 DE      26.7
10 FL     28.1
# i 42 more rows
```

```
colnames(ev_reg) <- (c("state", "count_of_regs"))  
  
ev_reg
```

```
# A tibble: 52 × 2
  state           count_of_regs
  <chr>                  <dbl>
1 Alabama            13047
2 Alaska              2697
3 Arizona             89798
```

```

4 Arkansas          7108
5 California        1256646
6 Colorado           90083
7 Connecticut        31557
8 Delaware            8435
9 District of Columbia   8066
10 Florida           254878
# i 42 more rows

```

`energy_cost_2023`

```

# A tibble: 52 × 2
  state    cost
  <chr>   <dbl>
1 AK      23.8
2 AL      21.1
3 AR      21.8
4 AZ      30.3
5 CA      35.7
6 CO      23.8
7 CT      32.3
8 DC      32.3
9 DE      26.7
10 FL     28.1
# i 42 more rows

```

Part 3: Joining / Pivoting Datasets for Analysis

```

# Now that the tables have the same state names, join them:
ev_reg_to_energy_cost <- full_join(ev_reg, energy_cost_2023, by = "state")

#remove total column at bottom
ev_reg_to_energy_cost <- ev_reg_to_energy_cost[1:51,]
print(ev_reg_to_energy_cost, n=52)

```

```

# A tibble: 51 × 3
  state          count_of_regs   cost
  <chr>           <dbl>   <dbl>
1 Alabama          13047     NA
2 Alaska            2697     NA
3 Arizona          89798     NA
4 Arkansas          7108     NA
5 California        1256646  NA

```

6 Colorado	90083	NA
7 Connecticut	31557	NA
8 Delaware	8435	NA
9 District of Columbia	8066	NA
10 Florida	254878	NA
11 Georgia	92368	NA
12 Hawaii	25565	NA
13 Idaho	8501	NA
14 Illinois	99573	NA
15 Indiana	26101	NA
16 Iowa	9031	NA
17 Kansas	11271	NA
18 Kentucky	11617	NA
19 Louisiana	8150	NA
20 Maine	7377	NA
21 Maryland	72139	NA
22 Massachusetts	73768	NA
23 Michigan	50284	NA
24 Minnesota	37050	NA
25 Mississippi	3590	NA
26 Missouri	26861	NA
27 Montana	4608	NA
28 Nebraska	6920	NA
29 Nevada	47361	NA
30 New Hampshire	9861	NA
31 New Jersey	134753	NA
32 New Mexico	10276	NA
33 New York	131250	NA
34 North Carolina	70164	NA
35 North Dakota	959	NA
36 Ohio	50393	NA
37 Oklahoma	22843	NA
38 Oregon	64361	NA
39 Pennsylvania	70154	NA
40 Rhode Island	6396	NA
41 South Carolina	20873	NA
42 South Dakota	1675	NA
43 Tennessee	33221	NA
44 Texas	230125	NA
45 Utah	39998	NA
46 Vermont	7816	NA
47 Virginia	84936	NA
48 Washington	152101	NA
49 West Virginia	2758	NA
50 Wisconsin	24943	NA
51 Wyoming	1139	NA

Part 4: Mapping Visualization

```
length(energy_cost_2023$cost)
```

```
[1] 52
```

```
energy_cost_2023$cost[52]
```

```
[1] 23.59
```

```
us_average_cost <- mean(ev_reg_to_energy_cost$cost)  
us_average_cost
```

```
[1] NA
```

```
ev_reg_to_energy_cost <- ev_reg_to_energy_cost |>  
  mutate(energy_cost_vs_average = (cost / us_average_cost))
```

```
# mapping ratio of energy cost  
  
# create map of US states  
states_map <- st_as_sf(map("state", plot = FALSE, fill = TRUE))  
  
# Match state names: map names are lowercase  
ev_reg_to_energy_cost$state <- tolower(ev_reg_to_energy_cost$state)  
# Join map data  
ratio_map <- states_map |>  
  left_join(ev_reg_to_energy_cost, by = c("ID" = "state"))
```

```
# map of energy cost in states  
ggplot(ratio_map) +  
  geom_sf(aes(fill = energy_cost_vs_average), color = "white") +  
  scale_fill_gradient2(  
    low = "green", mid = "white", high = "red",  
    midpoint = 1  
  ) +  
  labs(  
    title = "Electricity Price by State Compared to Average",  
    subtitle = "Red = above US average, Green = below US average"  
  ) +  
  theme_minimal() +
```

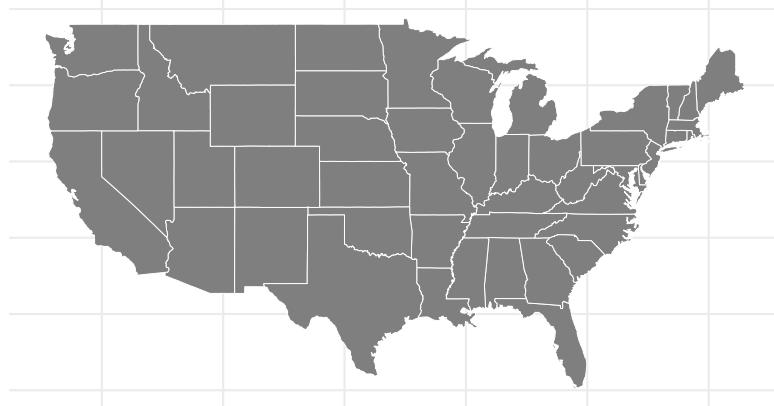
```

theme(
  axis.text = element_blank(),
  axis.ticks = element_blank()
)

```

Electricity Price by State Compared to Average

Red = above US average, Green = below US average



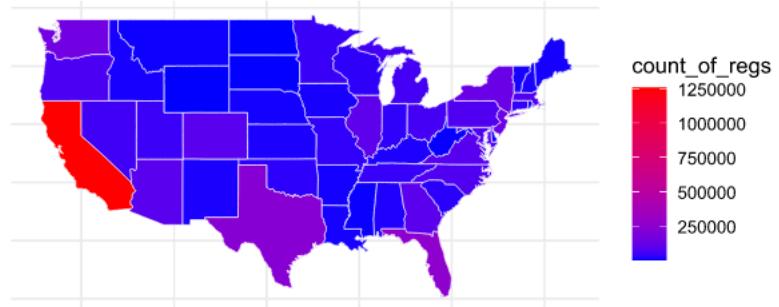
```

# map of number of EVs
ggplot(ratio_map) +
  geom_sf(aes(fill = count_of_regs), color = "white") +
  scale_fill_gradient(
    low = "blue", high = "red"
  ) +
  labs(
    title = "EV Registrations",
    subtitle = "Red = above US average, Blue = below US average"
  ) +
  theme_minimal() +
  theme(
    axis.text = element_blank(),
    axis.ticks = element_blank()
)

```

EV Registrations

Red = above US average, Blue = below US average



Part 5: Analysis

These graphs suggest a weak relationship between EV registrations and the cost of energy compared to the national average. This effect appears negligible since electricity is still cheaper than gasoline. Perhaps a more meaningful comparison would be EV registrations per state compared to that state's cost of gasoline.

However, there does seem to be a geographic relationship to EV registrations. The highest levels of EV registrations is at the states closest to the ocean, as well as states closer to the equator. California, a leader in renewable energy, stands out as leading the EV industry as well.