

EV Power - Lab 4 Project Report

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Example Solution 1

Introduction

The importance of protecting the environment never fails to becoming the focus of human beings living in modern age. The mission of seeking ways to improve the environmental condition is a crucial role for researchers in whatever area of study. For us Statistician (or at least Statistic Student), what we can contribute is using our professional knowledge of data manipulation and proposing insightful conclusions based on data.

In particular, this project asks us to look into the energy data of every states in the US from 2021 to 2023. From EDA, we are able to demystify some common misconception and ferret out overlooked features. In this report, after cleaning the dirty and chaotic data sets and joining them together, I will answer mainly three research questions:

1. How has the share of renewable energy changed from 2021–2023 across states?
2. What is the share of electricity that comes from clean sources by state?
3. Are EV registrations concentrated in states with cleaner energy mixes?

Starting this project late until the last day, I spent the whole afternoon finishing the data cleaning, and the whole evening drawing out figures and formulating the report. Clearly, the heavy workload of the projects and problem sheets far exceed my expectations of a 3 unit course, on par with the workload of notoriously 4 unit CS61B, or cosplaying the formidable Stat215a. Even so, it was a great pleasure of doing this project, where I takes a lot. With the assistance of Gemini AI, I was amazed by the power of R language in drawing out such an incredible map filled with my data. Thanks for Professor Andrew Bray and all the GSIs that gave us this valuable practice opportunity.

Part 0: libraries

Part 1: Defining Research Question

See Introduction.

Part 2: Data Preparation and Cleaning

Before cleaning, we have the raw data as below.

	State	Energy_Source	Renewable_Use_2021
1	AK	Biomass	≈3153

2	AK	Geothermal	186 MMBtu
3	AK	Hydropower	5763 about
4	AK	Solar Energy	~45
5	AK	Wind Energy	451 USD
6	AL	Biomass	198543 est.

	State	Energy_Source	Renewable_Use_2022
1	AK	Biomass	≈3846
2	AK	Geothermal	\$186
3	AK	Hydropower	\$5846
4	AK	Solar Energy	~57
5	AK	Wind Energy	\$475
6	AL	Biomass	193932 USD

	State	Energy_Source	Renewable_Use_2023
1	AK	Biomass	3404 kWh
2	AK	Geothermal	186.0
3	AK	Hydropower	6051
4	AK	Solar Energy	67
5	AK	Wind Energy	380
6	AL	Biomass	189040 kWh

	Total.energy.average.price..dollars.per.million.Btu...
1	'''
2	State,2021,2022,2023
3	AK,\$20.03 per MMBtu,\$27.33,\$23.84 est.
4	AL,about 17.85 USD,23.37 USD,≈21.11
5	AR,\$18.42,\$23.84 per MMBtu,\$21.76
6	AZ,≈25.07,31.72 USD,about 30.28

	Energy_Source	AK	AL	AR	AZ	CA	CO	CT		
1	Coal	18694	309791	216123	160299	28244	252442	2880		
2	Natural Gas†	395590	739891	360545	484962	2172757	509970	305184		
3	Petroleum (BTU)	261094	583042	328271	606862	2959389	497788	284788		
4	nuclear	0	480115	141372	329868	171842	0	179551		
5	total_renewable_energy	9597	239817	89714	99266	810020	103955	49306		
	DC	DE	FL	GA	HI	IA	ID	IL	IN	KS
KY										
1	0	4542	200193	203870	12566	264419	3051	522809	753557	219031
	548443									
2	28336	82708	1591864	773889	133	383424	135176	1088485	869328	291797
	365875									
3	18439	113641	1748346	922503	223014	408385	188263	1136797	712427	339006
	584011									

4	0	0	307811	354085	0	0	0	1011555	0	89426
0										
5	2487	7150	297291	289113	20134	389787	74428	224106	157324	135551
71744										
	LA	MA	MD	ME	MI	MN	MO	MS	MT	NC
ND										
1	95856	0	69186	1588	436203	179055	616413	64446	122765	222501
361811										
2	1862349	404301	299282	57233	950364	523812	293633	576903	87105	637553
191168										
3	1840835	503312	433791	163991	814081	561731	607276	384328	176686	884299
168682										
4	179886	0	156369	0	358114	147286	44766	122771	0	449675
0										
5	135905	75370	52732	95141	194075	216113	88879	66134	56334	196973
92653										
	NE	NH	NJ	NM	NV	NY	OH	OK	OR	PA
1	216298	3259	12586	133228	35910	5370	575920	131695	1303	485193
2	191008	60116	697019	285809	305212	1359437	1294814	745911	305665	1868137
3	237214	142030	749892	262885	286548	1237451	1028000	517408	317322	1047658
4	71758	102789	293494	0	0	325141	182330	0	0	791587
5	158275	38479	70039	62210	63647	263977	146858	177087	225544	179589
	RI	SC	SD	TN	TX	UT	VA	VT	WA	WI
1	0	162628	21589	225784	968401	276159	68603	0	36943	286760
2	105473	349990	96787	413554	4773076	274420	699927	13801	384769	561076
3	76464	508147	119505	713210	6783182	304823	795296	72241	711662	533390
4	0	560782	0	368461	419363	0	297972	0	88764	103979
5	11798	143796	127382	135841	654199	36050	174615	21430	394052	145936
	WY	US								
1	376971	10548957								
2	161580	31688203								
3	146274	35250685								
4	0	8130913								
5	37734	7646167								

	Energy_Source	AK	AL	AR	AZ	CA	CO	CT	DC
1	coal Consumption	18615	297654	211724	154007	30049	233256	0	0
2	Natural-Gas	437916	787300	398099	468038	2131372	524890	307212	30174
3	petroleum (btu)	263335	578431	327813	594859	3017944	538413	302881	18000
4	Nuclear Energy†	0	442093	149654	333738	183814	0	172018	0
5	total_renewables	10410	232035	90825	101215	880995	114917	49084	2622
	DE	FL	GA	HI	IA	ID	IL	IN	KS
1	1846	171953	180888	7680	227866	1881	496983	719238	226712
2	89674	1659544	809618	159	434374	141924	1134781	913401	318779
3	112026	1815529	940579	241994	423592	190635	1138141	699235	346852
4	0	321468	356001	0	0	0	1032989	0	93844
5	7402	304605	293237	20471	421784	78406	248541	170986	151788
									77517

	LA	MA	MD	ME	MI	MN	MO	MS	MT	NC	
1	96914	0	61932	1269	423504	184517	566940	66214	131345	163029	
2	2087166	432442	310133	62559	1087716	535010	322547	617855	93971	747187	
3	1663129	529154	411842	166724	820709	568916	606374	383366	177009	906477	
4	168889	0	154742	0	271788	153546	92724	89856	0	445547	
5	138209	80700	51255	93867	206811	229769	95312	66614	60644	198165	
	ND	NE	NH	NJ	NM	NV	NY	OH	OK	OR	
1	369340	223571	3864	6199	138077	35835	6143	539587	106855	1066	
2	198986	199260	60176	755048	301279	302315	1403401	1422175	772405	297591	
3	170390	237556	149025	769751	255571	303234	1321362	1031807	521629	315400	
4	0	58702	114108	295875	0	0	280133	175806	0	0	
5	96024	168382	39863	73187	77286	72734	269884	155282	189654	237768	
	PA	RI	SC	SD	TN	TX	UT	VA	VT	WA	WI
1	435540	0	150973	24769	204725	932569	237870	67739	0	42238	232501
2	1936985	93829	361249	98288	440017	5007366	287076	665869	14046	381886	622144
3	1108074	78260	495616	118593	707095	6582173	322387	798162	71534	725931	535483
4	795783	0	568055	0	372319	434709	0	294606	0	102929	105285
5	182051	13264	145328	129978	116472	751680	37369	185638	22009	418470	150890
	WV	WY	US								
1	536642	390303	9885694								
2	281657	172450	33361871								
3	201769	145723	35330835								
4	0	0	8061020								
5	28391	42079	8107353								

	Energy_Source			AK	AL	AR	AZ	CA	CO	CT
1	coal_usage			18414	224926	180262	137885	28746	204826	0
2	NaturalGas			448087	775747	399566	537151	2154533	525446	304924
3	petroleum (BTU)			270391	565754	327465	599712	2996168	514174	292864
4	nuclear-energy †			0	476392	156492	329474	185192	0	142873
5	total renewable-energy			10087	222189	87277	108445	1065179	115061	48981
	DC	DE	FL	GA	HI	IA	ID	IL	IN	KS
KY										
1	0	338	129387	177521	0	201276	1144	342683	613533	184614
481815										
2	26236	84387	1673836	787361	152	446677	154150	1101064	921814	309427
369986										
3	17292	110721	1835394	980546	251676	404172	189553	1134461	695709	345807
584722										
4	0	0	312935	390663	0	0	0	1019691	0	107675
0										
5	2795	8041	286306	291462	21046	414801	77128	245703	172891	140268
72603										
	LA	MA	MD	ME	MI	MN	MO	MS	MT	NC
1	58224	0	30349	1295	287490	148968	442901	49606	130059	153784
2	2055504	386946	304669	61045	1104234	536789	316512	630107	96777	662302
3	1620038	525647	429784	177091	810789	567072	612625	378072	173283	900241

```

4 127634      0 156610      0 292615 124626 95947 122807      0 442493
5 138982 81559 53711 89444 198459 223864 90412 67305 58470 186804
   ND      NE      NH      NJ      NM      NV      NY      OH      OK      OR
1 325716 195602 1838      0 75182 29284 4823 413577 63787 652
2 220768 206276 59589 721282 337083 301655 1346622 1448857 860217 327164
3 169307 233599 147387 787262 251686 296155 1341811 1009729 515440 313013
4      0 72391 99658 296162      0      0 287690 169392      0      0
5 92154 164502 38988 74408 80278 74878 272967 153083 185378 236062
   PA      RI      SC      SD      TN      TX      UT      VA      VT      WA
WI
1 307604      0 162323 22246 202367 805600 174315 46785      0 49523
219995
2 1937041 112499 346881 99752 396870 5284670 298976 655997 13001 403038
565025
3 1132958 76844 507146 114623 702827 6752349 324640 807547 70235 718277
525386
4 787083      0 581365      0 396522 425186      0 310037      0 88163
101204
5 178035 13579 142486 126540 115678 791211 39674 183979 22209 365956
150965
   WV      WY      US
1 472309 366098 8169673
2 309019 181395 33609104
3 206969 143944 35460356
4      0      0 8098974
5 28370 38474 8187317

```

```

electric.vehicle.registrations_by_state..2023.      X
1
2      STATE Count-EVs
3      Alabama      #13047
4      Alaska      ~2697
5      Arizona      89798
6      Arkansas      7108 EVs

```

After performing the monumental efforts on data cleaning, I am able to get the frames as below.

```

State_abb 2021 2022 2023      State
1      AK 20.03 20.03 20.03      Alaska
2      AL 17.85 17.85 17.85      Alabama
3      AR 18.42 18.42 18.42      Arkansas
4      AZ 25.07 25.07 25.07      Arizona
5      CA 28.44 28.44 28.44      California
6      CO 20.64 20.64 20.64      Colorado

```

	State	State_abb	count
1	Alabama	AL	13047
2	Alaska	AK	2697
3	Arizona	AZ	89798
4	Arkansas	AR	7108
5	California	CA	1256646
6	Colorado	CO	90083

	State_abb	Source	Price_2021
1	AK	Biomass	3153
2	AK	Geothermal	186
3	AK	Hydropower	5763
4	AK	Solar Energy	45
5	AK	Wind Energy	451
6	AL	Biomass	198543

	State_abb	Source	Price_2022
1	AK	Biomass	3846
2	AK	Geothermal	186
3	AK	Hydropower	5846
4	AK	Solar Energy	57
5	AK	Wind Energy	475
6	AL	Biomass	193932

	State_abb	Source	Price_2023
1	AK	Biomass	3404
2	AK	Geothermal	186
3	AK	Hydropower	6051
4	AK	Solar Energy	67
5	AK	Wind Energy	380
6	AL	Biomass	189040

```
# A tibble: 6 × 3
  Energy_Source State_abb Price_2021
  <fct>         <chr>         <int>
1 Coal         AK             18694
2 Coal         AL             309791
3 Coal         AR             216123
4 Coal         AZ             160299
5 Coal         CA             28244
6 Coal         CO             252442
```

```
# A tibble: 6 × 3
  Energy_Source State_abb Price_2022
```

	<fct>	<chr>	<int>
1	Coal	AK	18615
2	Coal	AL	297654
3	Coal	AR	211724
4	Coal	AZ	154007
5	Coal	CA	30049
6	Coal	CO	233256

```
# A tibble: 6 × 3
  Energy_Source State_abb Price_2023
  <fct>         <chr>         <int>
1 Coal         AK          18414
2 Coal         AL         224926
3 Coal         AR         180262
4 Coal         AZ         137885
5 Coal         CA          28746
6 Coal         CO         204826
```

Part 3: Joining / Pivoting Datasets for Analysis

This part will be done along with the next part on EDA.

Part 4: Mapping Visualization

Question 1: How has the share of renewable energy changed from 2021–2023 across states?

To answer this question, I firstly join the three *cleaned_renew* data frames, named and shown as q1. I then adjust this joint data frame into a format better for future plotting, named and shown as analysis_q1.

	State_abb	Source	Price_2023	Price_2021	Price_2022
1	AK	Biomass	3404	3153	3846
2	AK	Geothermal	186	186	186
3	AK	Hydropower	6051	5763	5846
4	AK	Solar Energy	67	45	57
5	AK	Wind Energy	380	451	475
6	AL	Biomass	189040	198543	193932

```
# A tibble: 6 × 6
  State_abb Source      Year Energy_Amount Total_Energy_Per_Year Share
  <chr>      <fct>      <dbl>      <dbl>          <dbl> <dbl>
1 AK        Biomass    2023         3404          10088 0.337
2 AK        Biomass    2021         3153           9598 0.329
3 AK        Biomass    2022         3846          10410 0.369
4 AK        Geothermal  2023          186          10088 0.0184
```

5	AK	Geothermal	2021	186	9598	0.0194
6	AK	Geothermal	2022	186	10410	0.0179

I then use the ggplot to generate a multi-facets bar plot, where each facet represents a state, in which shows the change in proportion of renewable energy across the three years.

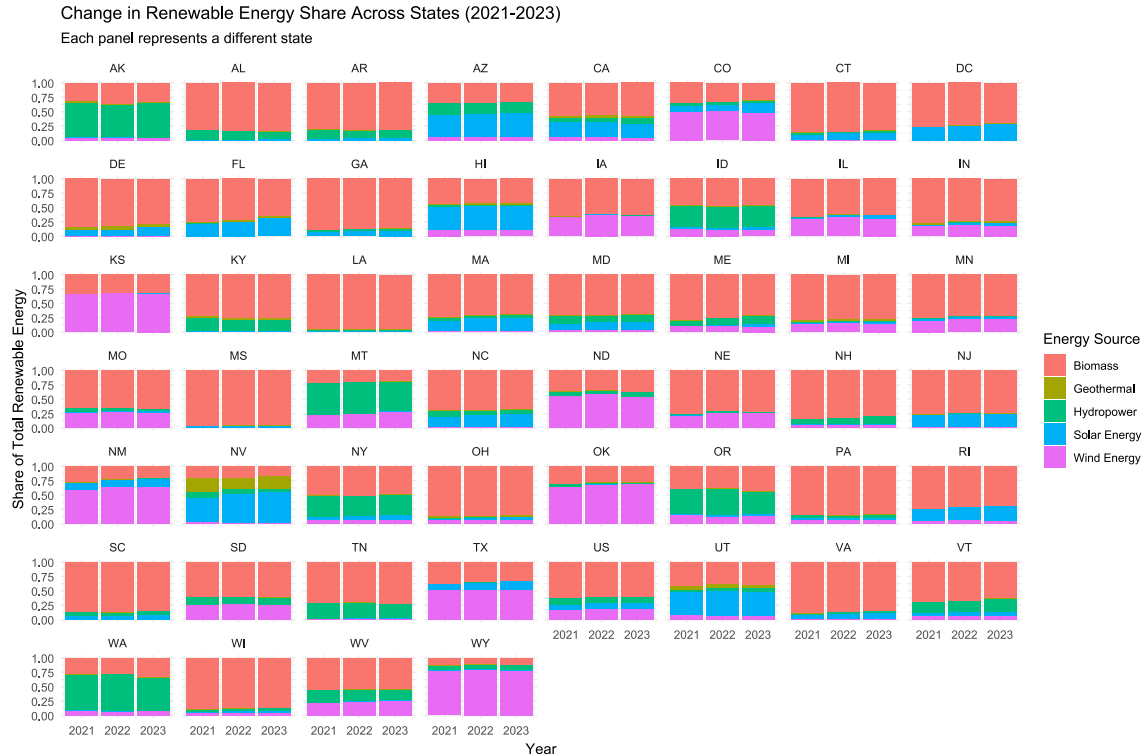
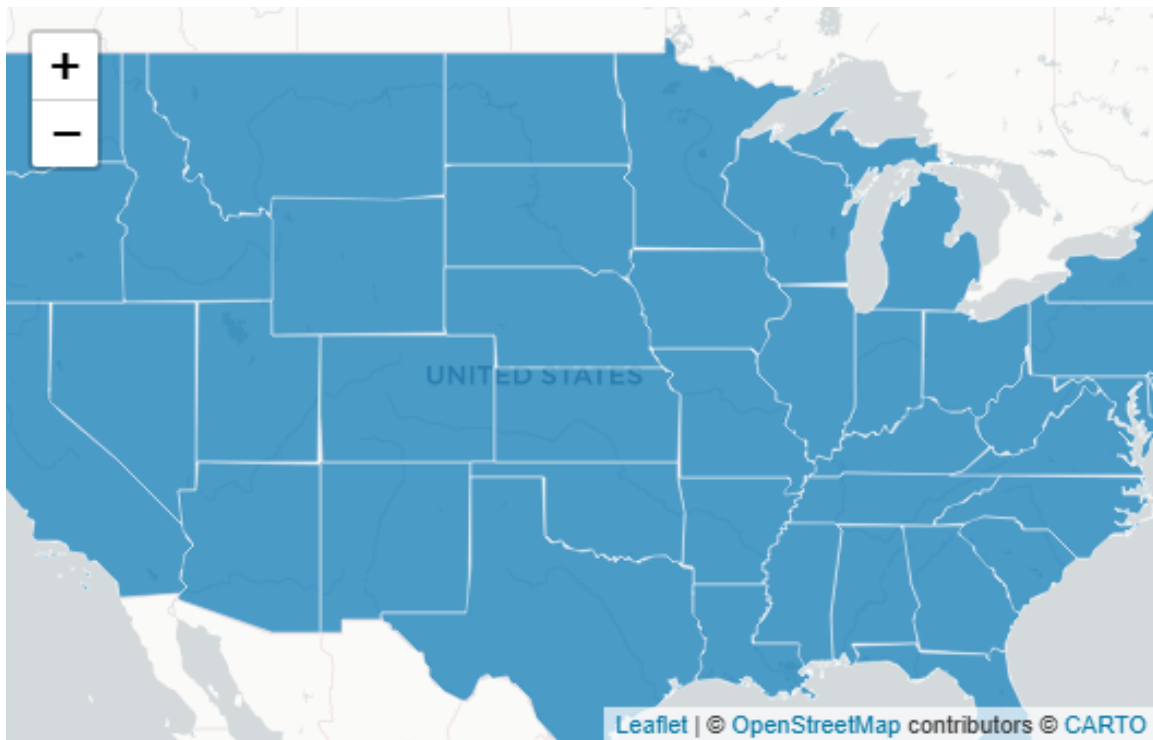


Figure 1: Change in Renewable Energy Share Across States (2021-2023)

However, since there are quite a lot of states and the change in proportion across years is small, it is quite hard to extract any understandable information from the figure. In this case, a better figure could be drawn by putting those data into the US map. With the help of Gemini, I finally able to create an interactive map which contains all the data in the previous multi-facets bar chart, which is so amazing.



Question 2: What is the share of electricity that comes from clean sources by state?

To answer this question, I join the three *cleaned_total_use* data frames named q2. Likewise, I reformat this new data frame, and calculate Share of the clean energy.

```
# A tibble: 6 × 6
  Energy_Source State_abb Year Energy_Amount Total_Energy_Per_Year Share
  <fct>         <chr>   <dbl>         <dbl>         <dbl> <dbl>
1 Total_Renewables AK      2021           9597         684975 0.0140
2 Total_Renewables AK      2022          10410         730276 0.0143
3 Total_Renewables AK      2023          10087         746979 0.0135
4 Total_Renewables AL      2021         239817        2352656 0.102
5 Total_Renewables AL      2022         232035        2337513 0.0993
6 Total_Renewables AL      2023         222189        2265008 0.0981
```

This time, we create another bar chart, which shows the proportion of electricity that comes from each state.

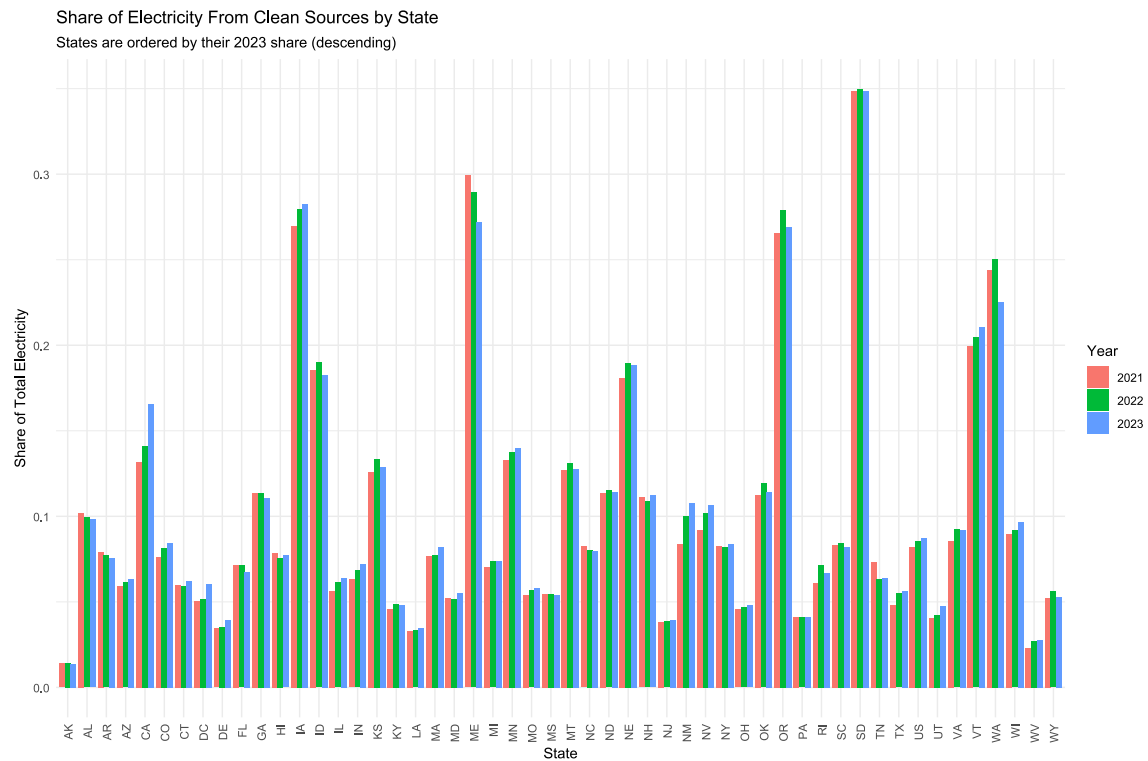
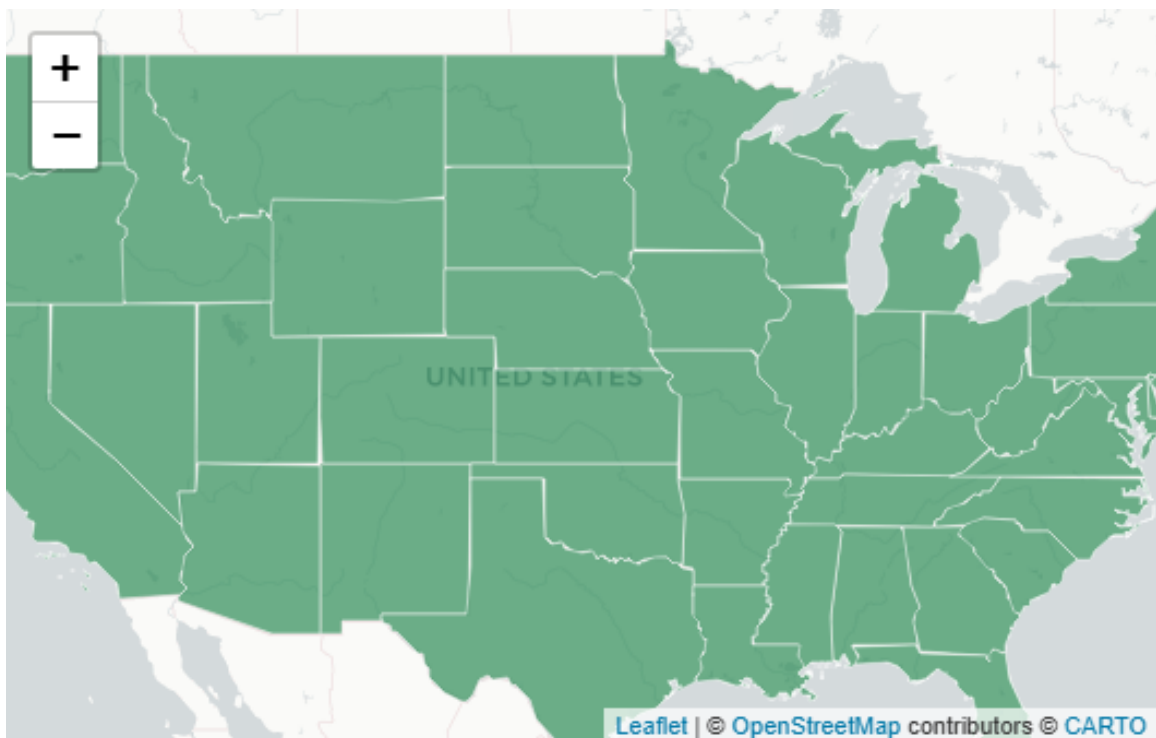


Figure 2: Share of Electricity From Clean Sources by State

Likewise, feed these data into the combination with the US map with the help of Gemini.

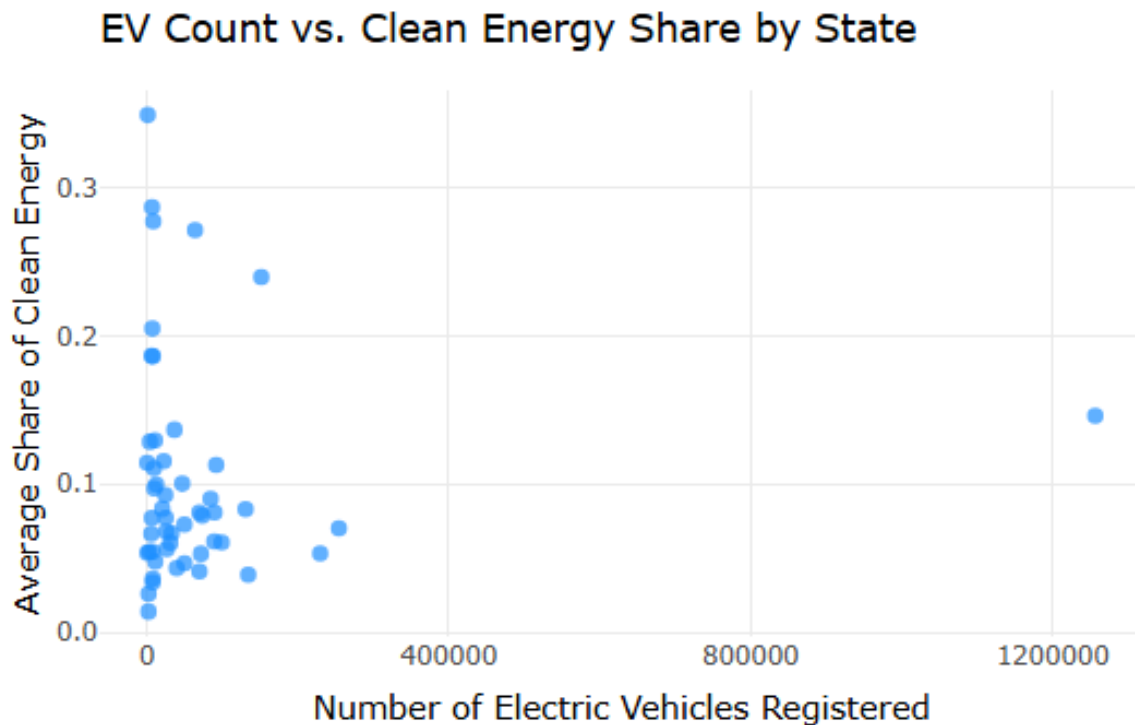


Question 3: Are EV registrations concentrated in states with cleaner energy mixes?

After getting some general senses of our data and the energy situations for every state, it would be helpful to get into some specific interrogations. Using our data from previous part, I join the data frame *q2* and *cleaned_ev_registrations*. Removing redundant columns, I get the following result.

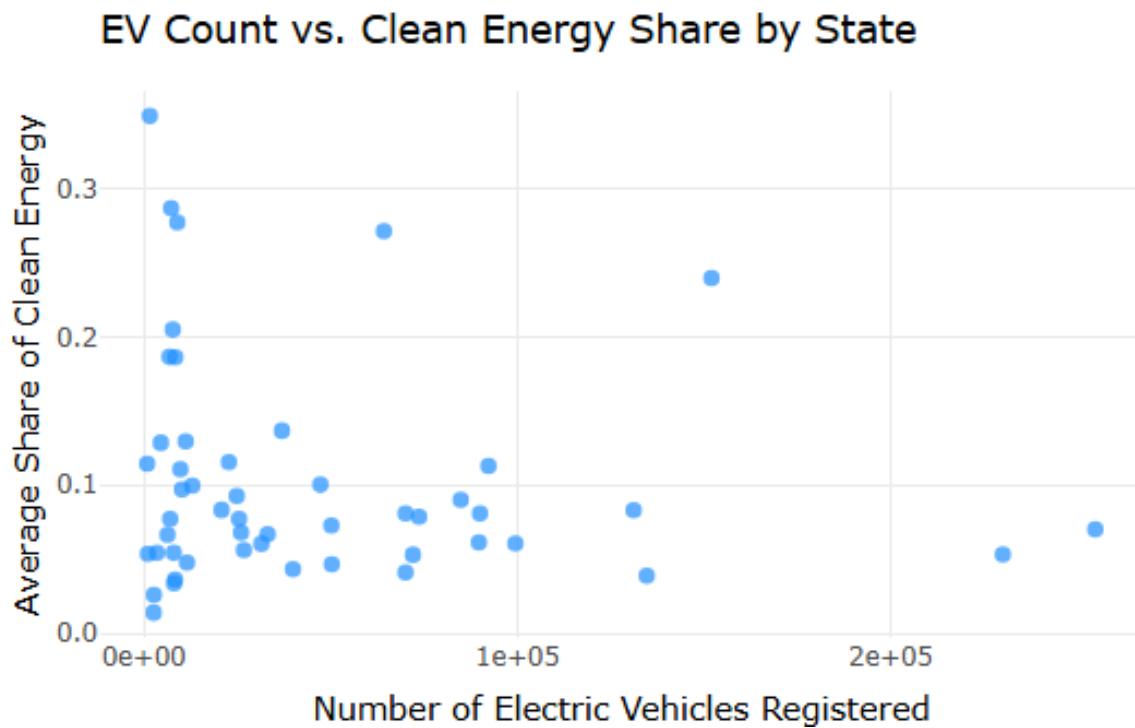
```
# A tibble: 6 × 3
  State_abb EV_count avg_share
  <chr>      <dbl>    <dbl>
1 AL         13047    0.0998
2 AK          2697    0.0139
3 AZ        89798    0.0612
4 AR          7108    0.0773
5 CA       1256646    0.146
6 CO         90083    0.0808
```

To visualize whether there is a relationship between clean energy proportion and EV registrations, I made a scatter plot with numbers of EVs against the average clean energy shares among 2021, 2022, 2023. To make it better for visualization, I make it interactively to show the name of the state for each point, with the help of Gemini.



Nevertheless, I'm facing the issue that the *EV_count* for California are far more than other states, which distorts the figure. Therefore, treating it as an outlier by removing it, I can replot the figures as above.

```
# A tibble: 6 × 3
  State_abb EV_count avg_share
  <chr>      <dbl>    <dbl>
1 AL        13047    0.0998
2 AK         2697    0.0139
3 AZ        89798    0.0612
4 AR         7108    0.0773
5 CO        90083    0.0808
6 CT        31557    0.0604
```



The figures give us a general sense on the relationship between these two variables, but to make more justifiable claim, we need the help of testing statistics. Below, I use a linear model to fit the two variables, and calculating some characteristic values as shown.

```
Call:
lm(formula = avg_share ~ EV_count, data = q3)

Residuals:
    Min       1Q   Median       3Q      Max
-0.08716 -0.04693 -0.02359  0.01357  0.24791

Coefficients:
            Estimate Std. Error t value Pr(>|t|)
            <dbl>      <dbl>    <dbl>  <dbl>
```

```

(Intercept) 1.010e-01 1.119e-02 9.028 5.35e-12 ***
EV_count    1.708e-08 5.893e-08 0.290 0.773
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.07434 on 49 degrees of freedom
(1 observation deleted due to missingness)
Multiple R-squared: 0.001712, Adjusted R-squared: -0.01866
F-statistic: 0.08402 on 1 and 49 DF, p-value: 0.7731

```

```

Pearson's product-moment correlation

data:  q3$avg_share and q3$EV_count
t = 0.28986, df = 49, p-value = 0.7731
alternative hypothesis: true correlation is not equal to 0
95 percent confidence interval:
 -0.2369117 0.3133837
sample estimates:
      cor
0.04137317

```

From the summary, we get a p-value larger than 0.05. Also, the correlation coefficient is very insignificant. Therefore, we cannot reject our null hypothesis, that there is no clear linear relationship between clean energy shares and number of EVs registered.