# ANALYSIS OF EV STATIONS ACROSS THE U.S.



Department of Applied Data Science, San Jose State University

Business Intelligence & Data Visualization

#### **Business Problem**

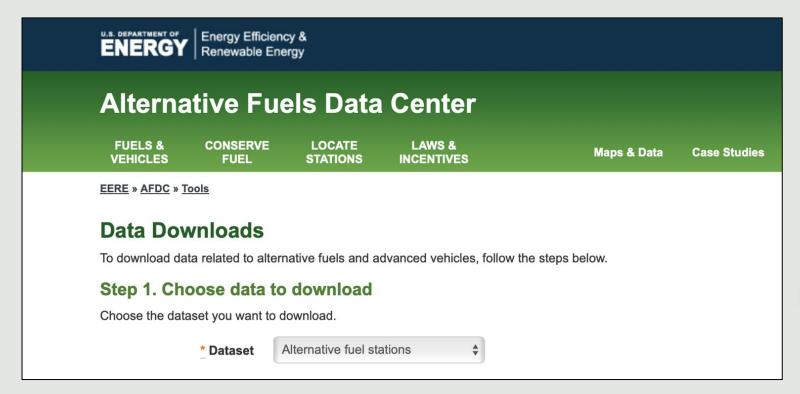


- Drive growth by targeting underserved regions based on datadriven trends.
- Identify gaps in sustainability efforts and promote the adoption of eco-friendly energy solutions in the EV charging network.
- Are there enough fast charging stations to meet the growing demand?
- Are charging stations dominated by a particular category of owners?

#### **Dataset Used**



- Source: U.S. Department of Energy Alternative Fuels Data Center.
- URL: <a href="https://afdc.energy.gov">https://afdc.energy.gov</a>
- Records Count: 78,837



### **Team Plan**

- Data collection and cleaning.
- EDA and feature engineering.
- Dashboard design and visualization.
- Advanced analytics.
- Aesthetic Refinements.

#### **EDA**



- Null Value Check for Key Columns
- Understanding the data distribution
- Duplicate Check
- Colab file link:

https://colab.research.google.com/drive/1oG bJiAL5pFXMYQX18lJ Mk7y0tL6lrln?usp=sharing#scrollTo=faJbRYog4uvo

# **Null Value Check for Key Columns**



Missing Data Overview:			
		Missing Percentage	
EV On-Site Renewable Source	78907	99.485595	
EV Level1 EVSE Num	78617	99.119965	
Federal Agency Name	78290	98.707685	
Federal Agency ID	78290	98.707685	
Federal Agency Code	78290	98.707685	
Access Detail Code	74822	94.335246	
Expected Date	73700	92.920633	
Cards Accepted	71518	90.169577	
Restricted Access	66811	84.235012	
EV DC Fast Count	66799	84.219883	
EV Pricing	63782	80.416063	
Facility Type	59921	75.548131	
Owner Type Code	58318	73.527076	
EV Level2 EVSE Num	11389	14.359201	
EV Network Web	9639	12.152808	
Access Davs Time	7986	10.068713	_
Open Date	441	0.556011	
Date Last Confirmed	156	0.196684	
Street Address	34	0.042867	
EV Workplace Charging	11	0.013869	
City	4	0.005043	
Station Name	3	0.003782	
Status Code	0	0.000000	
Updated At	0	0.000000	
State	0	0.000000	
Access Code	0	0.000000	
Longitude	0	0.000000	
Latitude	0	0.000000	
ID	0	0.000000	
Geocode Status	0	0.000000	
Fuel Type Code	0	0.000000	
EV Network	0	0.000000	
EV Connector Types	0	0.000000	
Country	0	0.000000	
ZIP	0	0.000000	
		<del>-</del>	

# Data Distribution (1)

Descri	ptive Statistics for Numerical	Columns:
	EV Level2 EVSE Num Latitu	ude Longitude \
count	79315.000000 79315.0000	000 79315.000000
mean	2.551291 37.8700	988 -96.199288
min	1.000000 18.0098	354 -162.286348
25%	2.000000 34.0433	346 -117.881102
50%	2.000000 38.5964	130 -91.068251
75%	2.551291 41.5916	-78.656218
max	338.000000 64.8524	166 -65.756678
std	3.357435 5.0417	741 19.380892
	Date Last Confirmed	ID /
count	79159	79315.000000
mean	2024-09-07 19:18:28.346492416	220069.654038
min	2019-12-12 00:00:00	1517.000000
25%	2024-10-11 00:00:00	165452.500000
50%	2024-12-02 00:00:00	205580.000000
75%	2024-12-02 00:00:00	308834.500000
max	2024-12-02 00:00:00	372237.000000
std	NaN	85900.785636
	•	EV Workplace Charging
count	78874	79315.000000
mean	2021-03-28 11:03:17.631665408	0.019268
min	1995-08-30 00:00:00	0.000000
25%	2020-06-12 00:00:00	0.000000
50%	2021-10-23 00:00:00	0.000000
75%	2023-06-16 00:00:00	0.000000
max	2025-01-15 00:00:00	1.000000
std	NaN	0.137455



## **Data Distribution (2)**

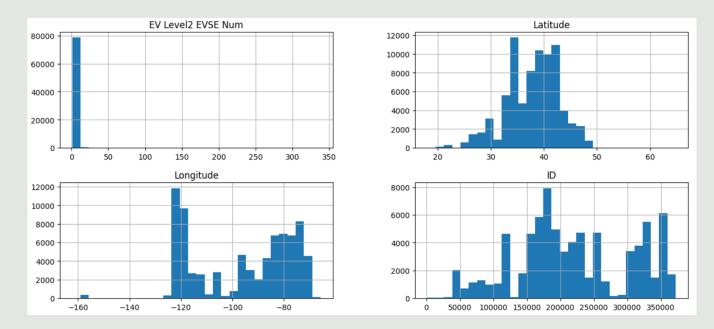
```
import matplotlib.pyplot as plt

# Plot histograms for all numerical columns

df_cleaned.select_dtypes(include=['float64', 'int64']).hist(figsize=(15, 10), bins=30)

plt.suptitle("Histograms for Numerical Columns", fontsize=16)

plt.show()
```





## **Duplicates Check**



```
# Check for duplicate rows
print(f"Number of duplicate rows before removal: {df_cleaned.duplicated().sum()}")

# Remove duplicate rows
df_no_duplicates = df_cleaned.drop_duplicates()

# Display results
print(f"Shape before duplicate removal: {df_cleaned.shape}")
print(f"Shape after duplicate removal: {df_no_duplicates.shape}")

Number of duplicate rows before removal: 0
Shape before duplicate removal: (79315, 22)
Shape after duplicate removal: (79315, 22)
```

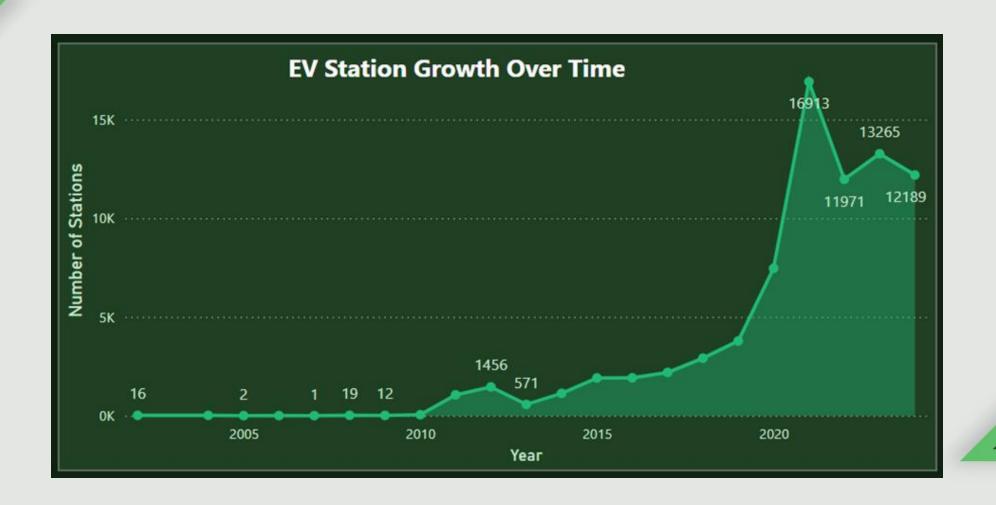
## **Analysis Performed**

(T)

- Trend Analysis over Time.
- Access Type Distribution.
- Top States by EV Station Utilization
- Availability of Fast Chargers.
- Classification of Key Stakeholders.
- ML Forecasting of Future Expansion.

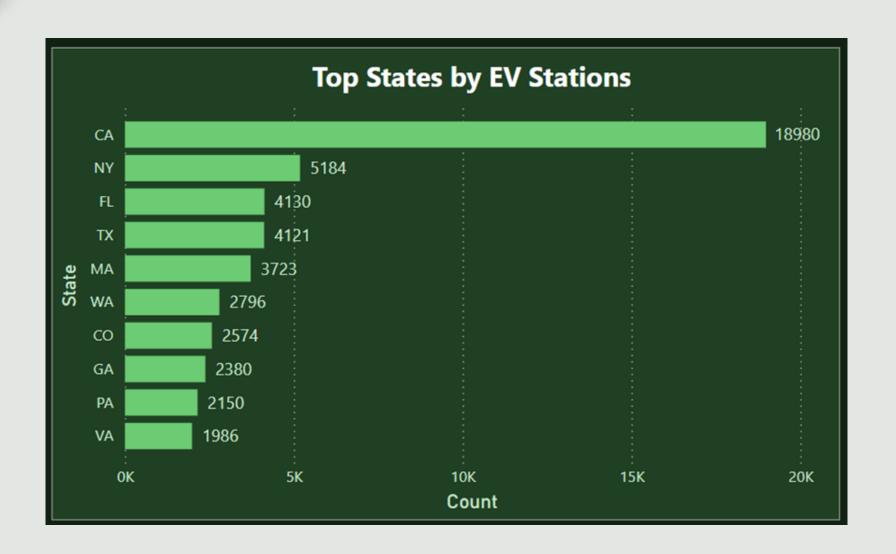
## **Area Chart**





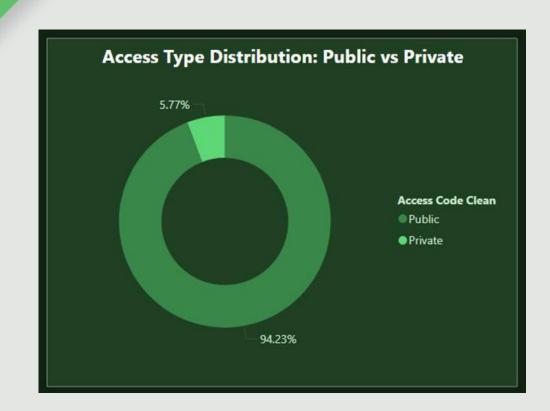


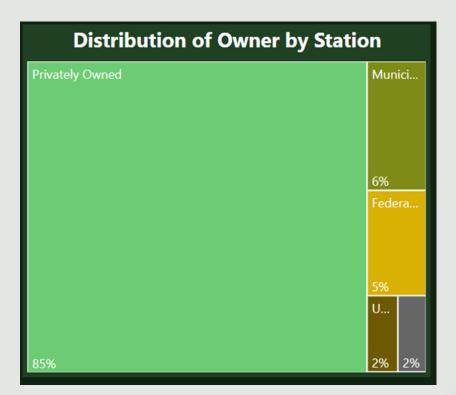




## **Donut Chart & Tree Map**

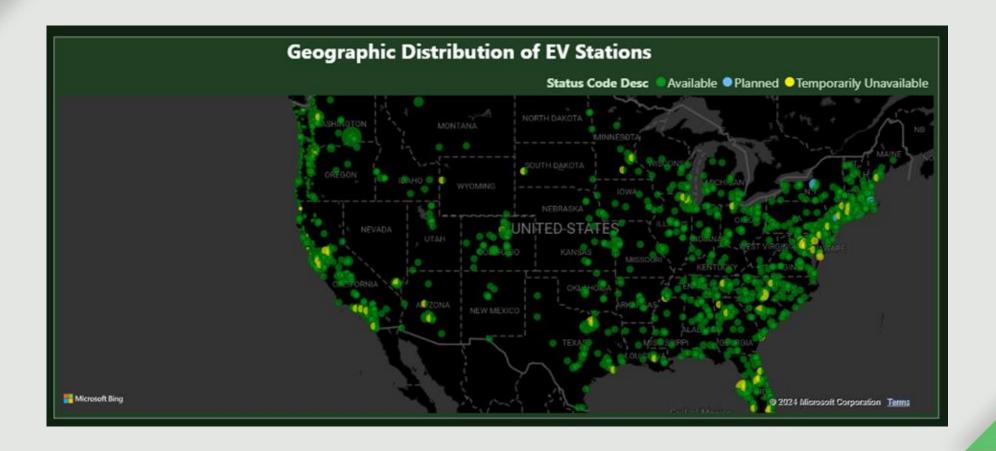






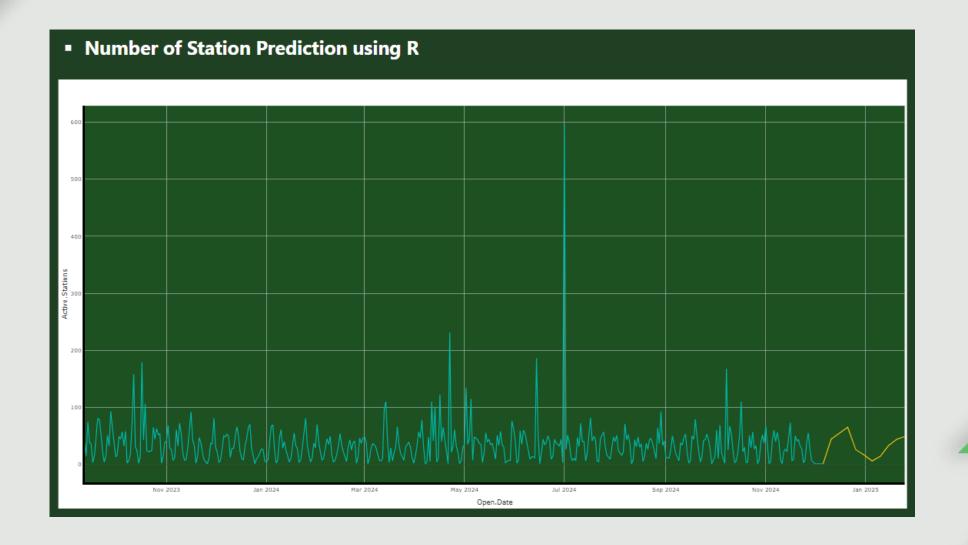
# **Geospatial Map**





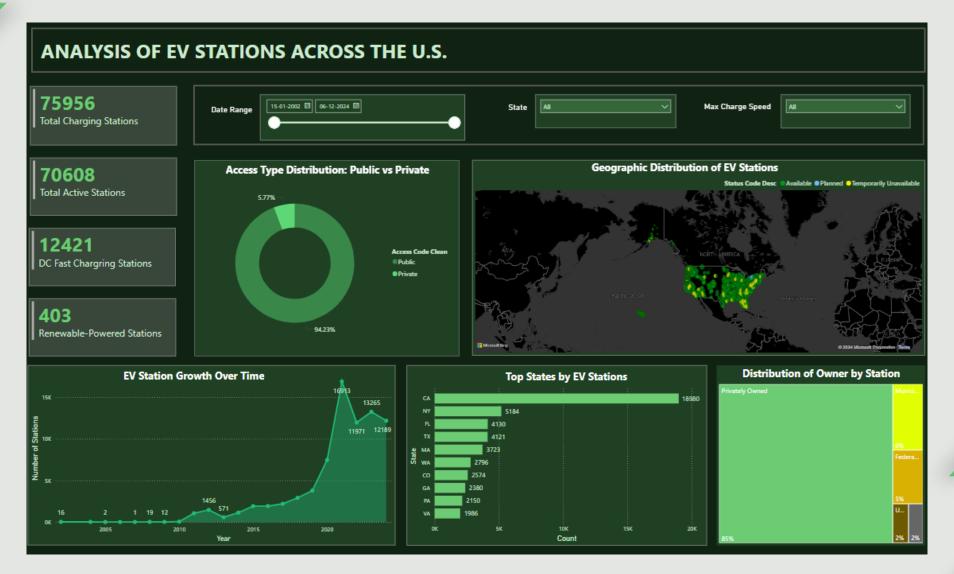












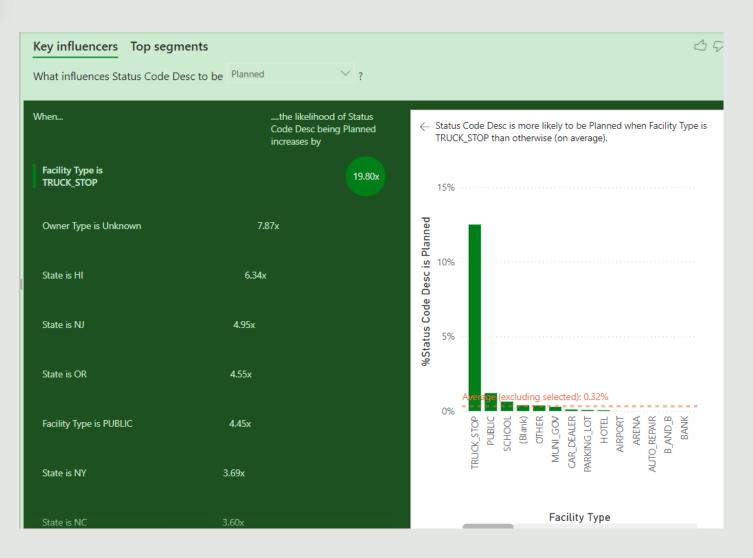
## Additional Features - Q & A



Which <u>state</u> has maximum <u>EV network stations</u> ?	
Showing results for <u>Top state</u> of <u>EV stations data by active stations of those EV stations data</u>	
State	
State  CA	

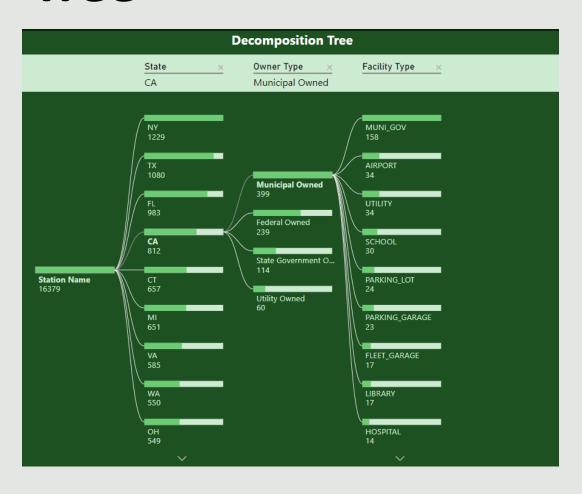






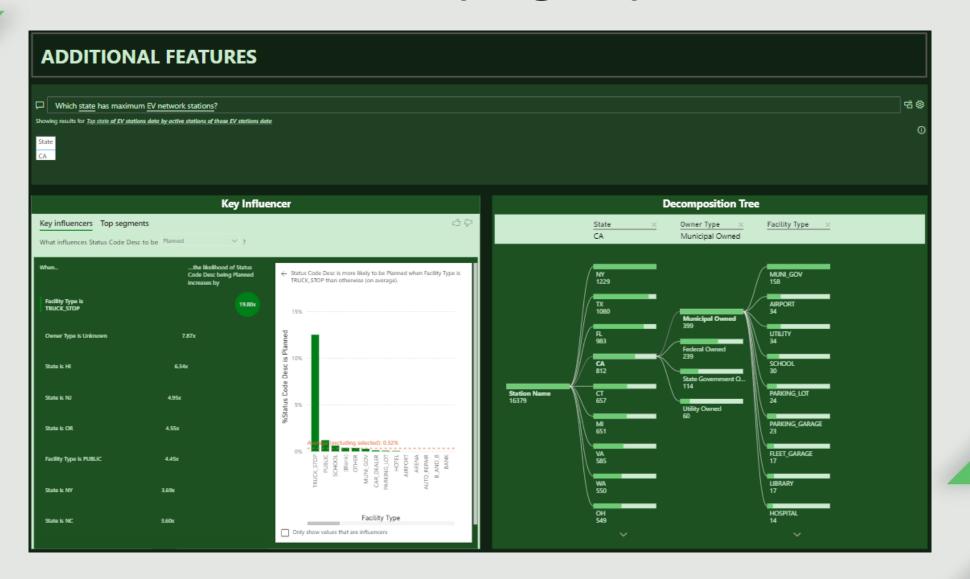
## Additional Features - Decomposition Tree





# **Dashboard View (Page 2)**





#### **Demonstration**



https://app.powerbi.com/groups/me/reports/2df8ca23-ccda-

44ce-bf88-

e7c20369b9b2/56dcbb3684d7de188dac?experience=power-bi

## **Advanced Features**

- DAX Metrics
- Q&A Feature
- Key Influencers
- Decomposition Tree

#### Conclusion



- <u>California leads</u> the U.S. in EV charging stations, showcasing its commitment to EV infrastructure.
- 95% of stations are publicly accessible, ensuring wide user access.
- EV charging stations have grown <u>steadily since 2010</u>, with a significant <u>surge after 2019</u>.
- Over 85% are privately owned, emphasizing the role of private entities in expansion.
- Only 18% offer DC Fast Charging, and less than 1% use renewable energy, highlighting opportunities for improvement in speed and sustainability.





- Integration with Real-Time Data to monitor station utilization and downtime.
- Use geo-spatial analysis and clustering techniques for recommending optimal locations for new stations.
- Incorporate user reviews for service improvements.



# Thank You!