IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

Javier Vázquez
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Outline

- Executive Summary
- Introduction
- Methodology
- Results
- Conclusion
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Executive Summary

- Summary of methodologies

- Data collection

- Data wrangling

- EDA with data visualization

- EDA with SQL

- Intecartive map building with Folium

- Dashboard building with Plotly Dash

- Predictive analysis with classification

- Summary of all results

- EDA result

- Intereactive analytics

- Predictive analiytics

Introduction

- Project background and context

The commercial space age has arrived, making space travel more accessible. Companies like Virgin Galactic, Rocket Lab, and Blue Origin are pioneering various aspects of space travel and satellite services. Among these, SpaceX stands out with significant achievements such as sending spacecraft to the International Space Station, launching the Starlink satellite internet constellation, and conducting manned space missions. A key factor in SpaceX's success is the relatively low cost of its Falcon 9 rocket launches, priced at $62 million compared to other providers' $165 million. This cost efficiency is largely due to SpaceX's ability to reuse the first stage of its rockets.

- Problems you want to find answers

- Space Y, a new rocket company and aims to compete with SpaceX. However, Space Y faces a significant challenge: determining the cost of each rocket launch. Unlike SpaceX, which has mastered the art of reusing the first stage of its rockets, Space Y needs to predict whether the first stage of their rockets will land successfully. This prediction is crucial because it directly impacts the overall cost of the launch. So the aim of presentation to predict the succesfull missions to save costs, demages.

Section 1

Methodology

Methodology

Executive Summary

- Data collection methodology:
  - SpaceX Open source Rest API
- Perform data wrangling
  - Transforming categorical data using One Hot Encoding for machine learning algorithms and removing any empty or unecessary information from data set
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Logistic regression, K-nearest neighbors, Support Vector Machine and Decesion Tree

Data Collection

- 1. Request to the SpaceX API

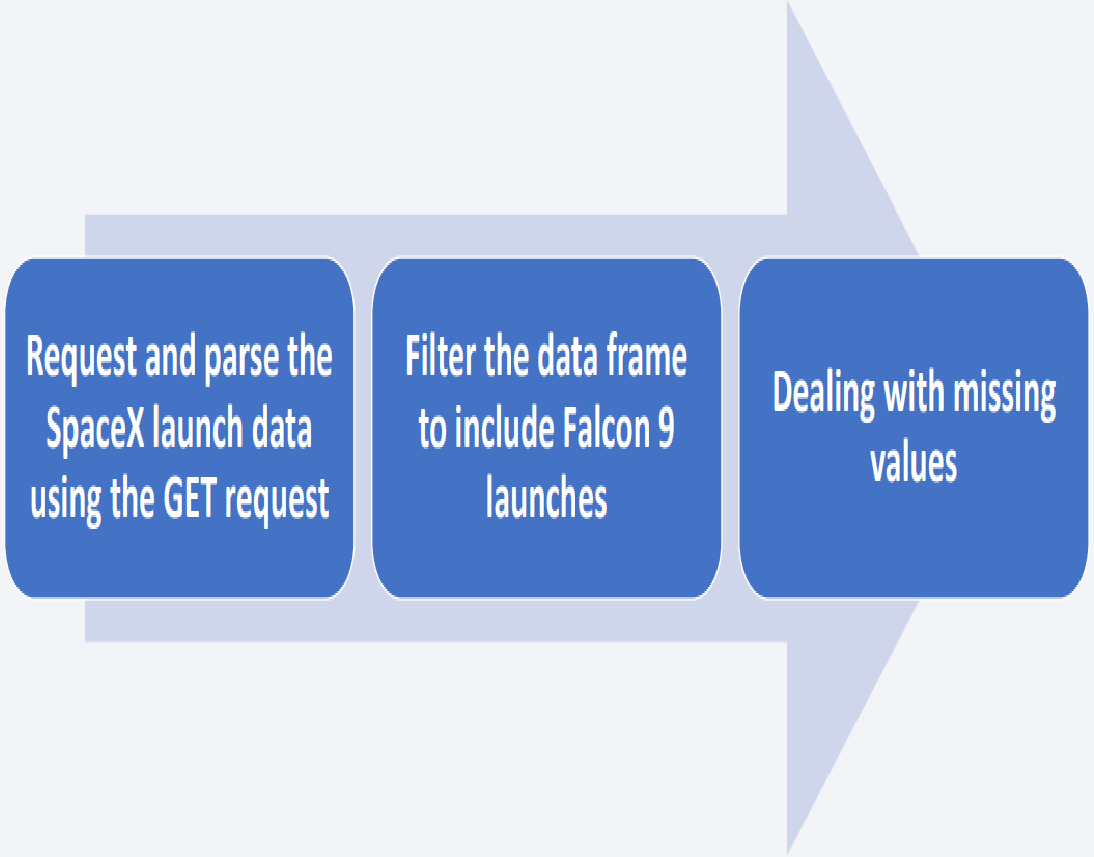
- Collected data from SpaceX’s open API
- Retrived and processed the data with GET request
- Ensure that only Falcon9 lanunches will be download
- Filled missing payload values with avarage values

- 2. Web Scrapping

- Requested past Falcon9 and Falcon Heavy launch data from Wikipedia link.
- Accessed to the Falcon9 launch pages via a direct Wikipedia link.
- Extracted all the column names from the HTML table
- Parsed and transformed the table into a Pandas data frame suitable for analysis.

Data Collection – SpaceX API

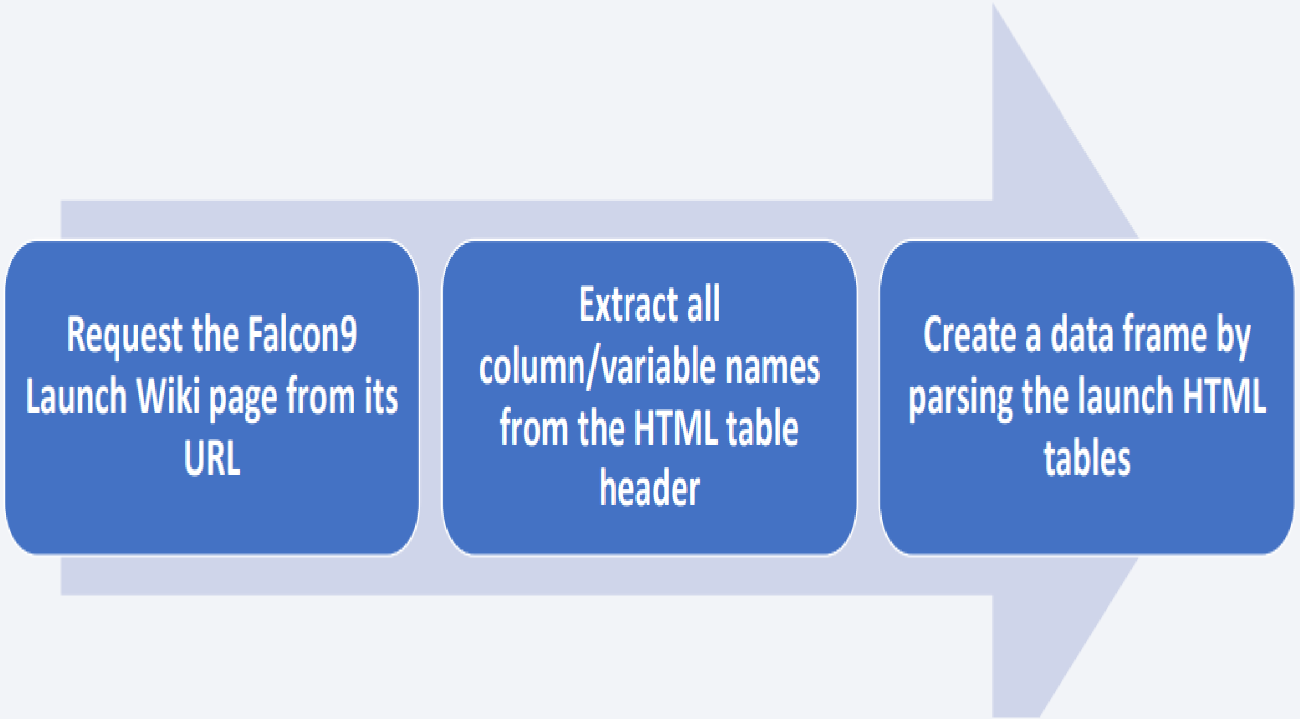
- Used SpaceX API for primary data collection. ○ Helper Functions: Defined functions to extract relevant launch data using unique identifiers. ○ Data Request: Performed GET requests to SpaceX API URL for rocket launch data. ○ Data Processing: Parsed JSON results, then decoded and filtered to include only Falcon 9 launches, converting data into a Pandas DataFrame.
- Web Scraping - Falcon 9 Launch Records (Wikipedia) ○ Access Wikipedia Page ○ Extract HTML Table ○ Parse and Convert data to structured format ○ Export Data



Data Collection - Scraping

- Present your web scraping process using key phrases and flowcharts

- Add the GitHub URL of the completed web scraping notebook, as an external reference and peer-review purpose



Data Wrangling

Pattern Analysis ○ Performed Exploratory Data Analysis (EDA) to identify patterns and assess landing success rates for label assignment in model training. ● Define Landing Outcomes ○ Examined various mission outcomes: ■ True Ocean: Successful landing in the ocean. ■ False Ocean: Attempted landing but unsuccessful in the ocean. ● Label Creation ○ Converted mission outcomes into training labels: ■ 1 for successful landings. ■ 0 for unsuccessful landings. ● Export Labeled Data

EDA with Data Visualization

- Bar chart
- • To compare the success rate of each orbit
- • Scatter plot chart
- • Identify the correlation between:
- Line chart
- • Visualize the launch success yearly trend

EDA with SQL

1. %sql select distinct(LAUNCH_SITE) from SPACEXTBL;
2. %sql select * from SPACEXTBL where LAUNCH_SITE like 'CCA%' limit 5;
3. %sql select sum(PAYLOAD_MASS__KG_) as payloadmass from SPACEXTBL;
4. %sql select avg(PAYLOAD_MASS__KG_) as payloadmass from SPACEXTBL;
5. %sql select min(DATE) from SPACEXTBL;
6. %sql select BOOSTER_VERSION from SPACEXTBL where LANDING_OUTCOME='Success (drone ship)' and
7. %sql select count(MISSION_OUTCOME) as missionoutcomes from SPACEXTBL GROUP BY MISSION_OUTCOME;
8. %sql select BOOSTER_VERSION as boosterversion from SPACEXTBL where PAYLOAD_MASS__KG_=(select max(PAYLOAD_MASS__KG_) from SPACEXTBL);
9. %sql SELECT substr(Date, 6, 2) as month, MISSION_OUTCOME, BOOSTER_VERSION, LAUNCH_SITE FROM SPACEXTBL WHERE [Landing_Outcome] = 'Failure (drone ship)' AND substr(Date, 1, 4) = '2015’;
10. %sql SELECT LANDING_OUTCOME FROM SPACEXTBL WHERE DATE BETWEEN '2010-06-04' AND '2017- 03-20' ORDER BY DATE DESC;

Build an Interactive Map with Folium

- • folium.Circle

- to add a highlighted circle area with a text label on a specific coordinate

• folium.Marker

- to marker_cluster

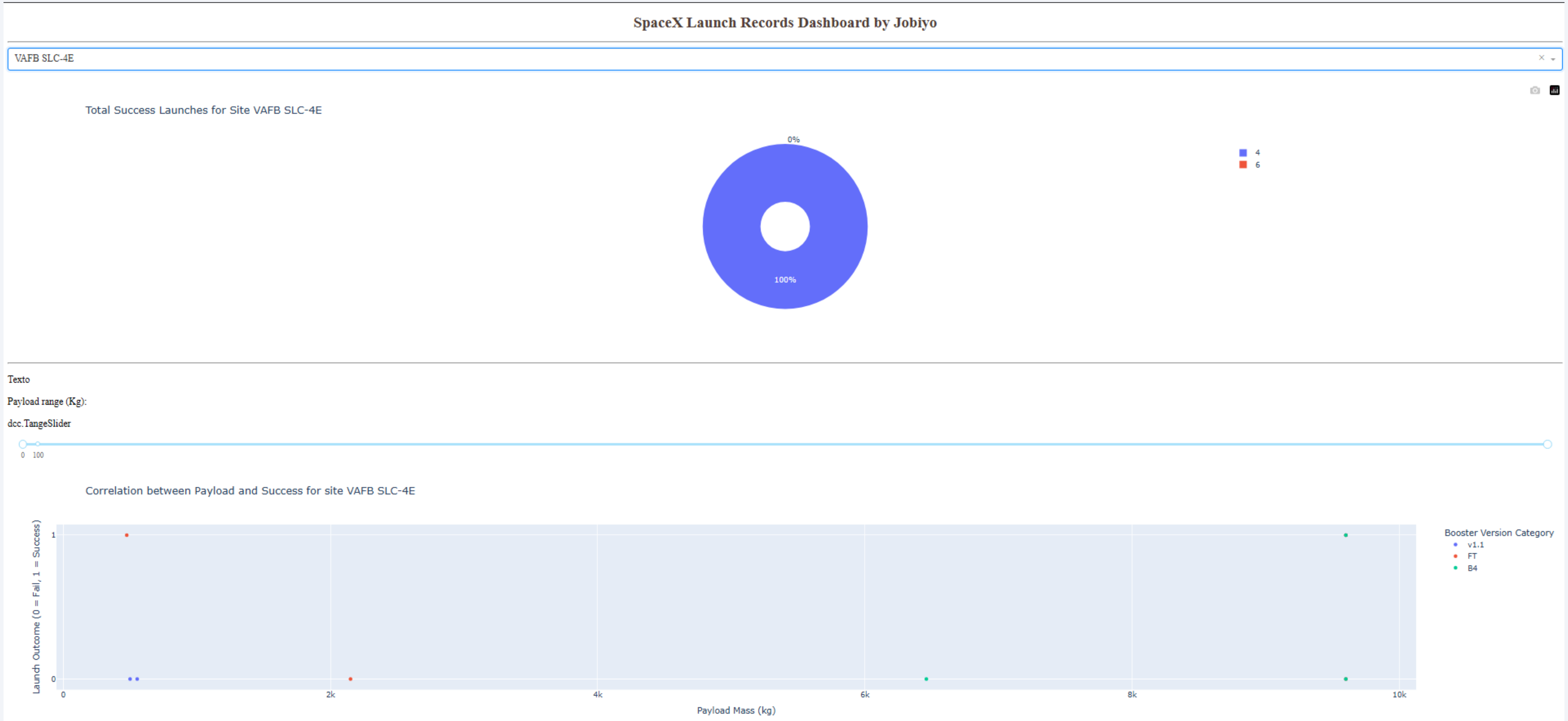
• folium.Popup

- is used to display additional information when a user clicks on a marker on a map

• folium.Map

- It allows you to set the initial location, zoom level, and other map settings.

Build a Dashboard with Plotly Dash



Predictive Analysis (Classification)

- • Perform exploratory Data Analysis and determine Training Labels • create a column for the class • Standardize the data • Split into training data and test data • Find best Hyperparameter for SVM, Classification Trees and Logistic Regression • Find the method performs best using test data

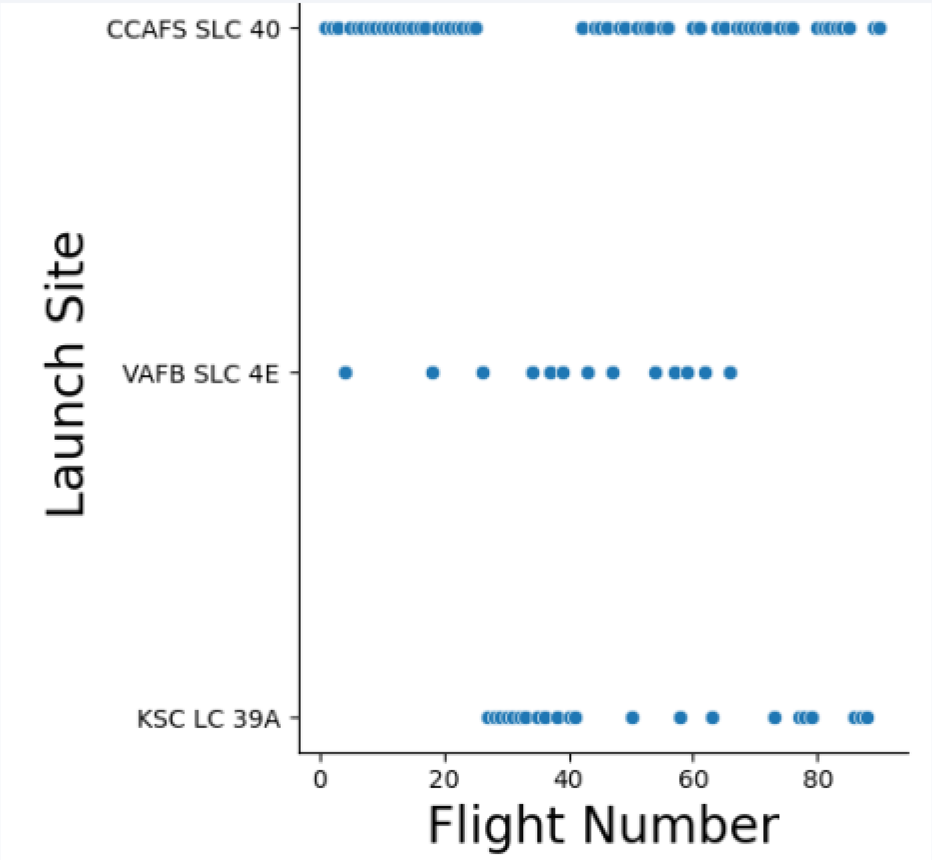
Results

- Exploratory data analysis results
  - Launch success has improved • KSC LC-39A has the highest success rate among landing sites • Orbits ES-L1, GEO, HEO, and SSO have a 100% cusses rate
- Interactive analytics demo in screenshots
  - Launch sites are fare engough to make demage
- Predictive analysis results
  - Decision Tree model is the best predictive model for the dataset

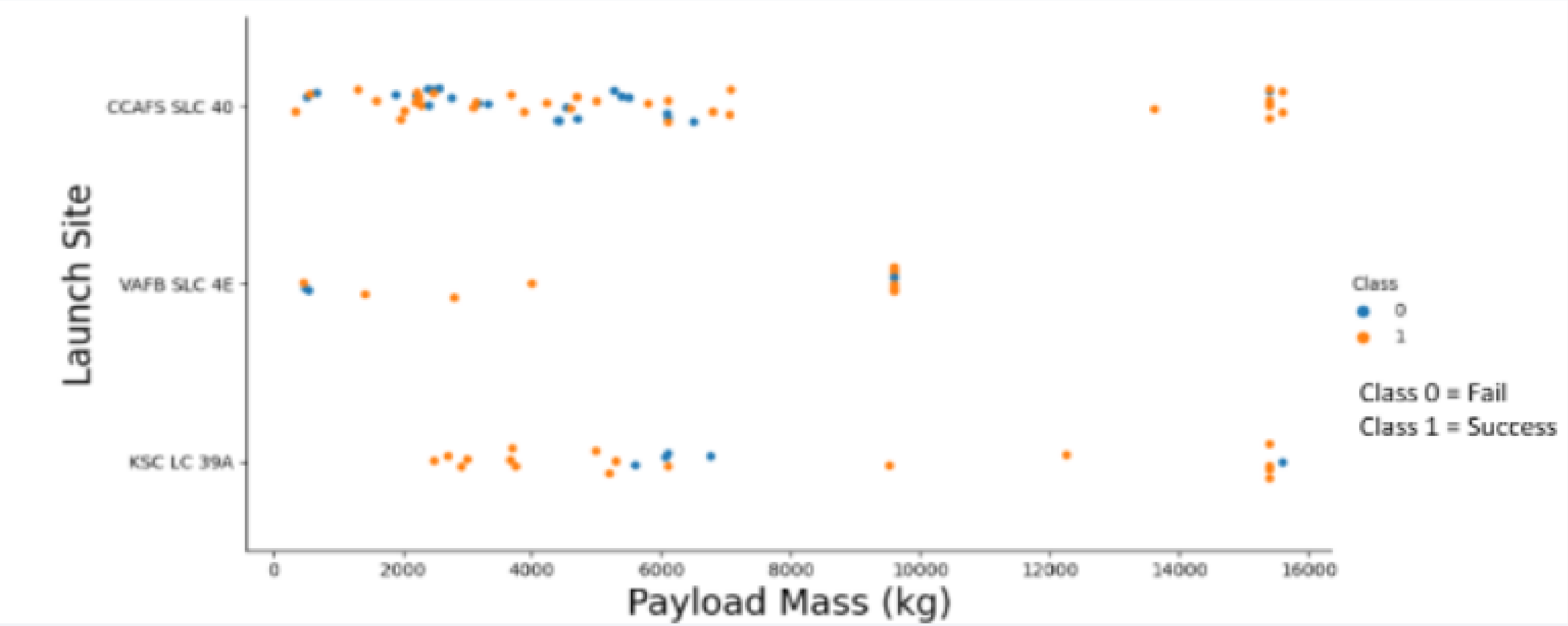
Section 2

Insights drawn from EDA

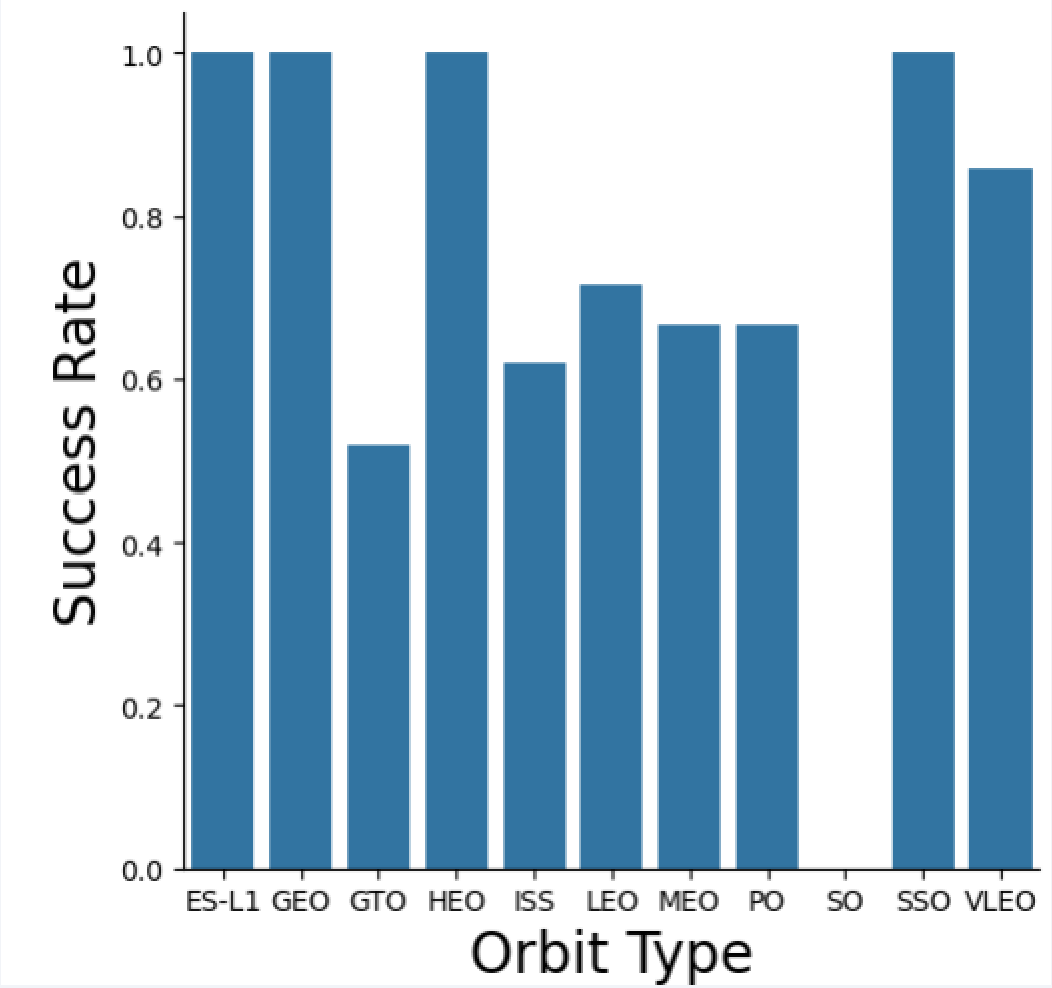
Flight Number vs. Launch Site



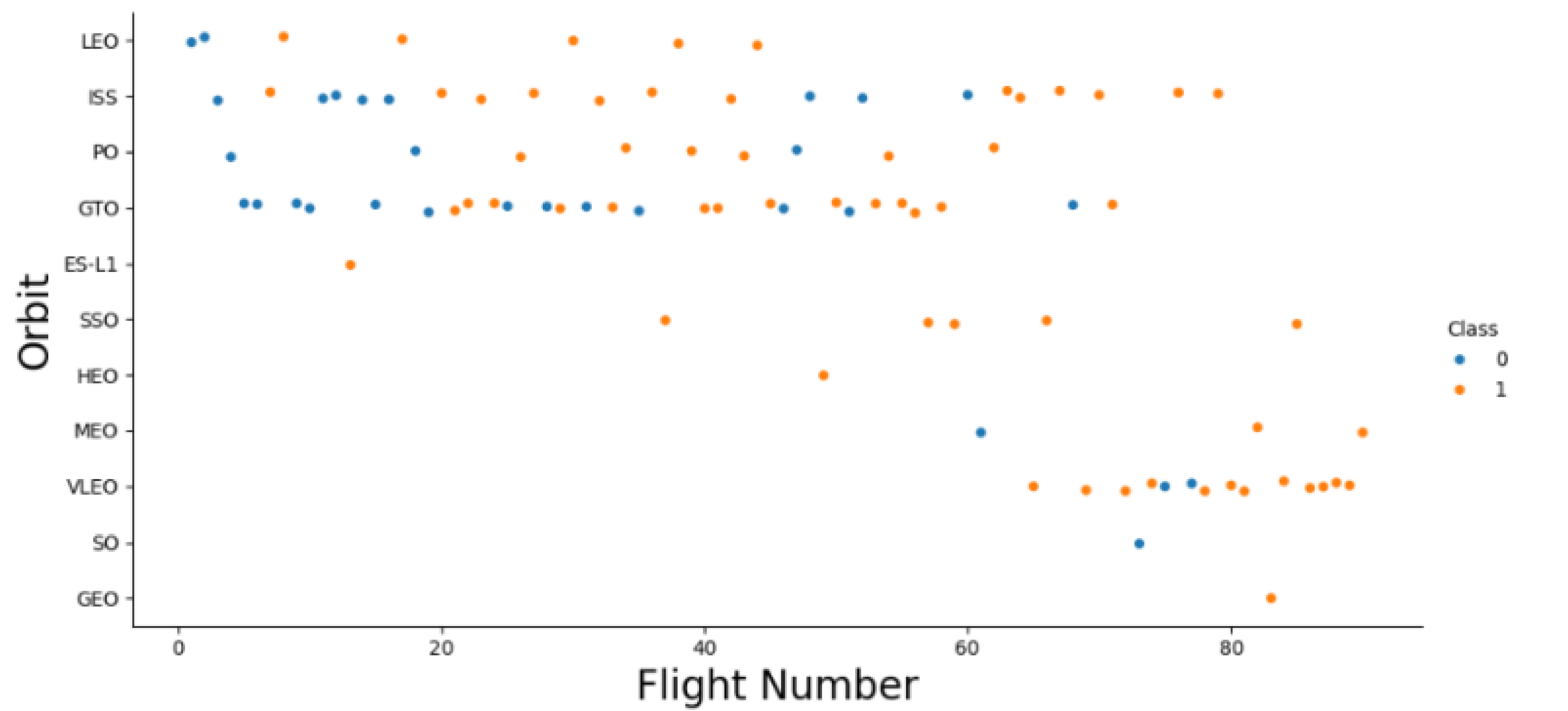
Payload vs. Launch Site



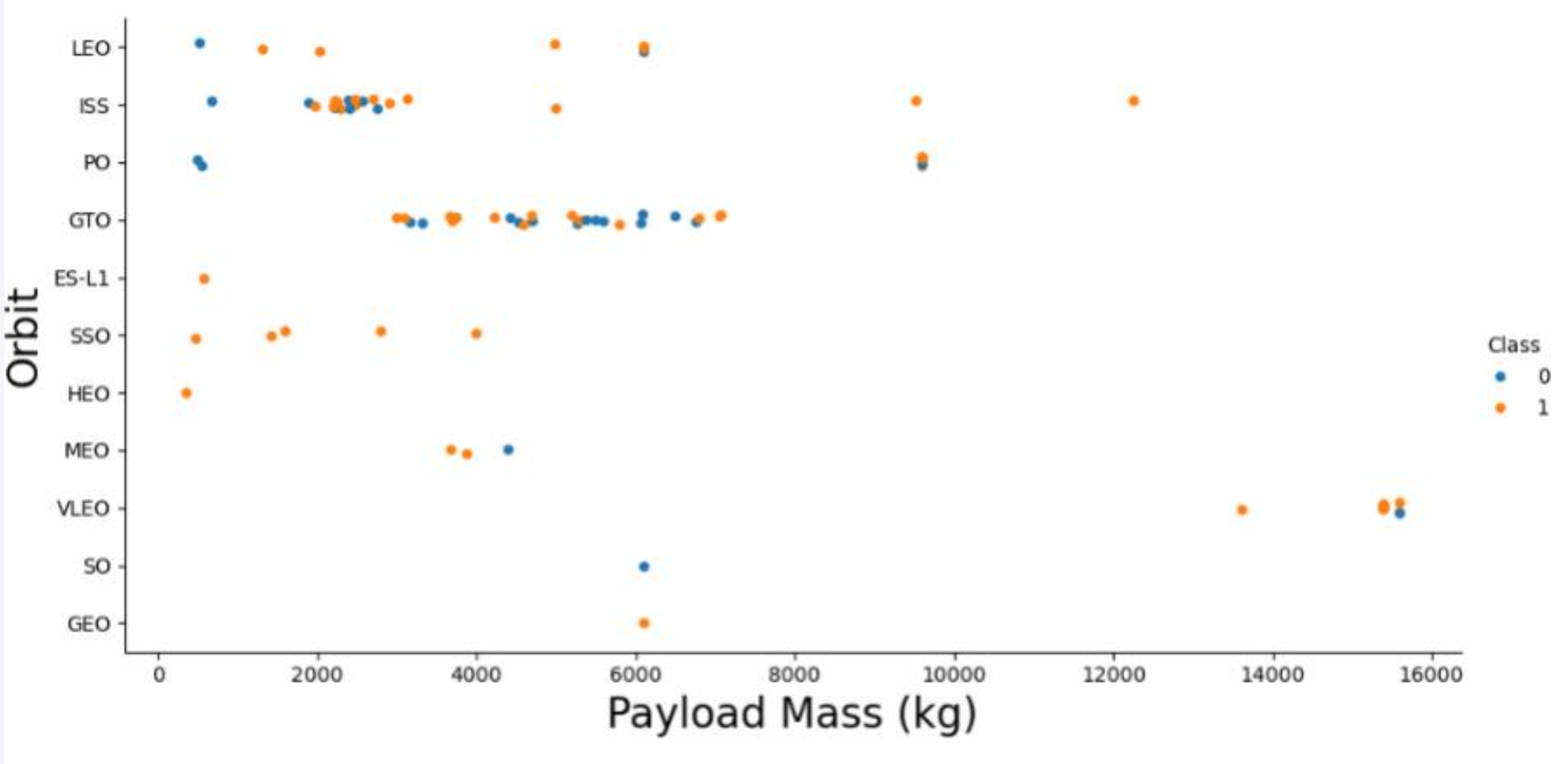
Success Rate vs. Orbit Type



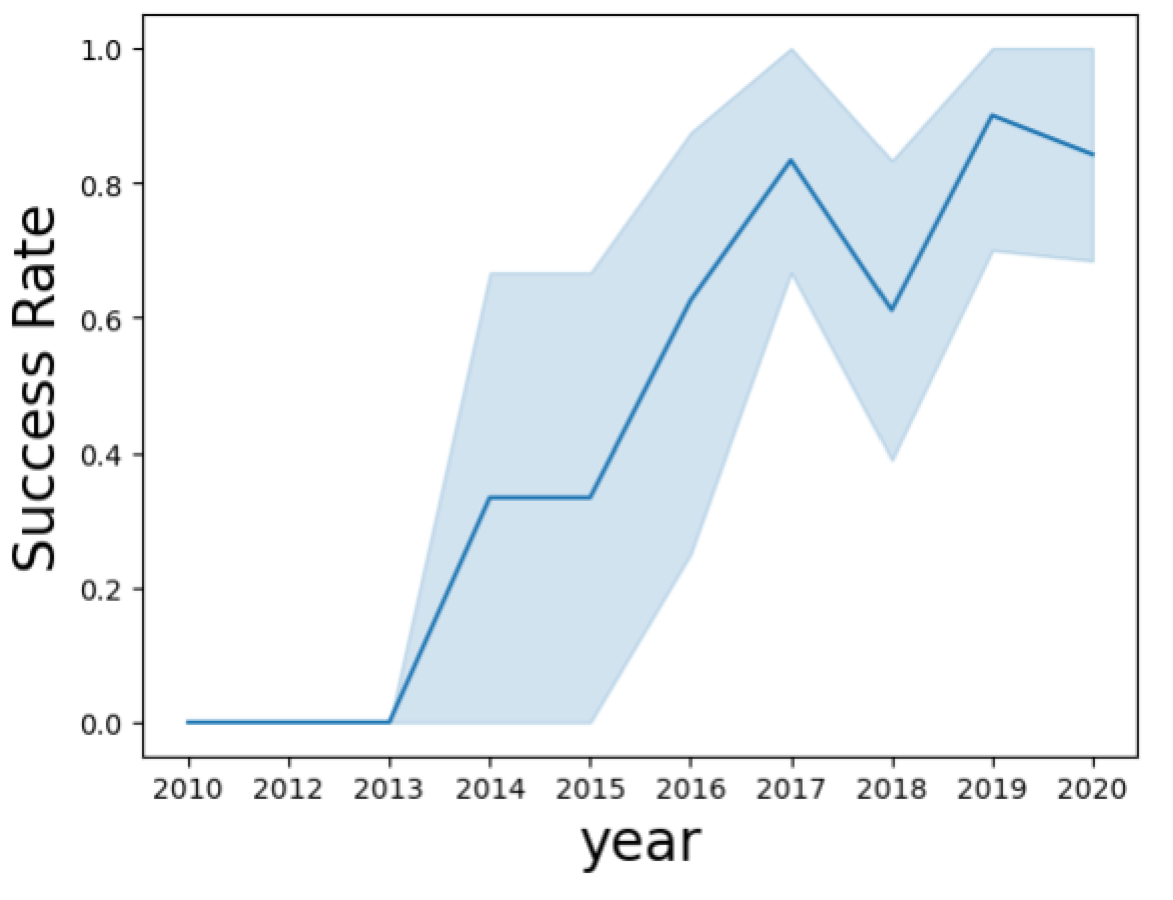
Flight Number vs. Orbit Type



Payload vs. Orbit Type



Launch Success Yearly Trend



All Launch Site Names

```
%sql select distinct(LAUNCH_SITE) from SPACEXTBL;
```

```
 * sqlite:///my_data1.db
Done.
```

Launch_Site
CCAFS LC-40
VAFB SLC-4E
KSC LC-39A
CCAFS SLC-40

Launch Site Names Begin with 'CCA'

Task 2

Display 5 records where launch sites begin with the string 'CCA'

```
%sql select * from SPACEXTBL where LAUNCH_SITE like 'CCA%' limit 5;
```

```
 * sqlite:///my_data1.db
Done.
```

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS__KG_	Orbit	Customer	Mission_Outcome
2010-06-04	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success
2010-12-08	15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0	LEO (ISS)	NASA (COTS) NRO	Success
2012-05-22	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success
2012-10-08	0:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success
2013-03-01	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success

Total Payload Mass

Display the total payload mass carried by boosters launched by NASA (CRS)

```
%sql select sum(PAYLOAD_MASS__KG_) as payloadmass from SPACEXTBL;
```

```
 * sqlite:///my_data1.db
Done.
```

payloadmass
619967

Average Payload Mass by F9 v1.1

Display average payload mass carried by booster version F9 v1.1

```
%sql select avg(PAYLOAD_MASS__KG_) as payloadmass from SPACEXTBL;
```

```
 * sqlite:///my_data1.db
Done.
```

payloadmass
6138.287128712871

First Successful Ground Landing Date

List the date when the first succesful landing outcome in ground pad was acheived.

Hint:Use min function

```
%sql select min(DATE) from SPACEXTBL;
```

```
 * sqlite:///my_data1.db
Done.
```

min(DATE)
2010-06-04

Successful Drone Ship Landing with Payload between 4000 and 6000

List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

```
%sql select BOOSTER_VERSION from SPACEXTBL where LANDING_OUTCOME='Success (drone ship)' and PAYLOAD_MASS__K(
```

```
 * sqlite:///my_data1.db
Done.
```

Booster_Version
F9 FT B1022
F9 FT B1026
F9 FT B1021.2
F9 FT B1031.2

Total Number of Successful and Failure Mission Outcomes

List the total number of successful and failure mission outcomes

```
%sql select count(MISSION_OUTCOME) as missionoutcomes from SPACEXTBL GROUP BY MISSION_OUTCOME;
```

```
 * sqlite:///my_data1.db
Done.
```

missionoutcomes
1
98
1
1

Boosters Carried Maximum Payload

List the names of the booster_versions which have carried the maximum payload mass. Use a subquery

```
%sql select BOOSTER_VERSION as boosterversion from SPACEXTBL where PAYLOAD_MASS__KG_=(select max(PAYLOAD_MA:
```

```
 * sqlite:///my_data1.db
Done.
```

boosterversion
F9 B5 B1048.4
F9 B5 B1049.4
F9 B5 B1051.3
F9 B5 B1056.4
F9 B5 B1048.5
F9 B5 B1051.4
F9 B5 B1049.5
F9 B5 B1060.2
F9 B5 B1058.3
F9 B5 B1051.6
F9 B5 B1060.3
F9 B5 B1049.7

2015 Launch Records

List the records which will display the month names, failure landing_outcomes in drone ship ,booster versions, launch_site for the months in year 2015.

Note: SQLLite does not support monthnames. So you need to use substr(Date, 6,2) as month to get the months and substr(Date,0,5)='2015' for year.

```
%sql SELECT substr(Date, 6, 2) as month, MISSION_OUTCOME, BOOSTER_VERSION, LAUNCH_SITE FROM SPACEXTBL WHERE
```

```
 * sqlite:///my_data1.db
Done.
```

month	Mission_Outcome	Booster_Version	Launch_Site
01	Success	F9 v1.1 B1012	CCAFS LC-40
04	Success	F9 v1.1 B1015	CCAFS LC-40

Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order.

```
%sql SELECT LANDING_OUTCOME FROM SPACEXTBL WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20' ORDER BY DATE DESC;
```

```
 * sqlite:///my_data1.db
Done.
```

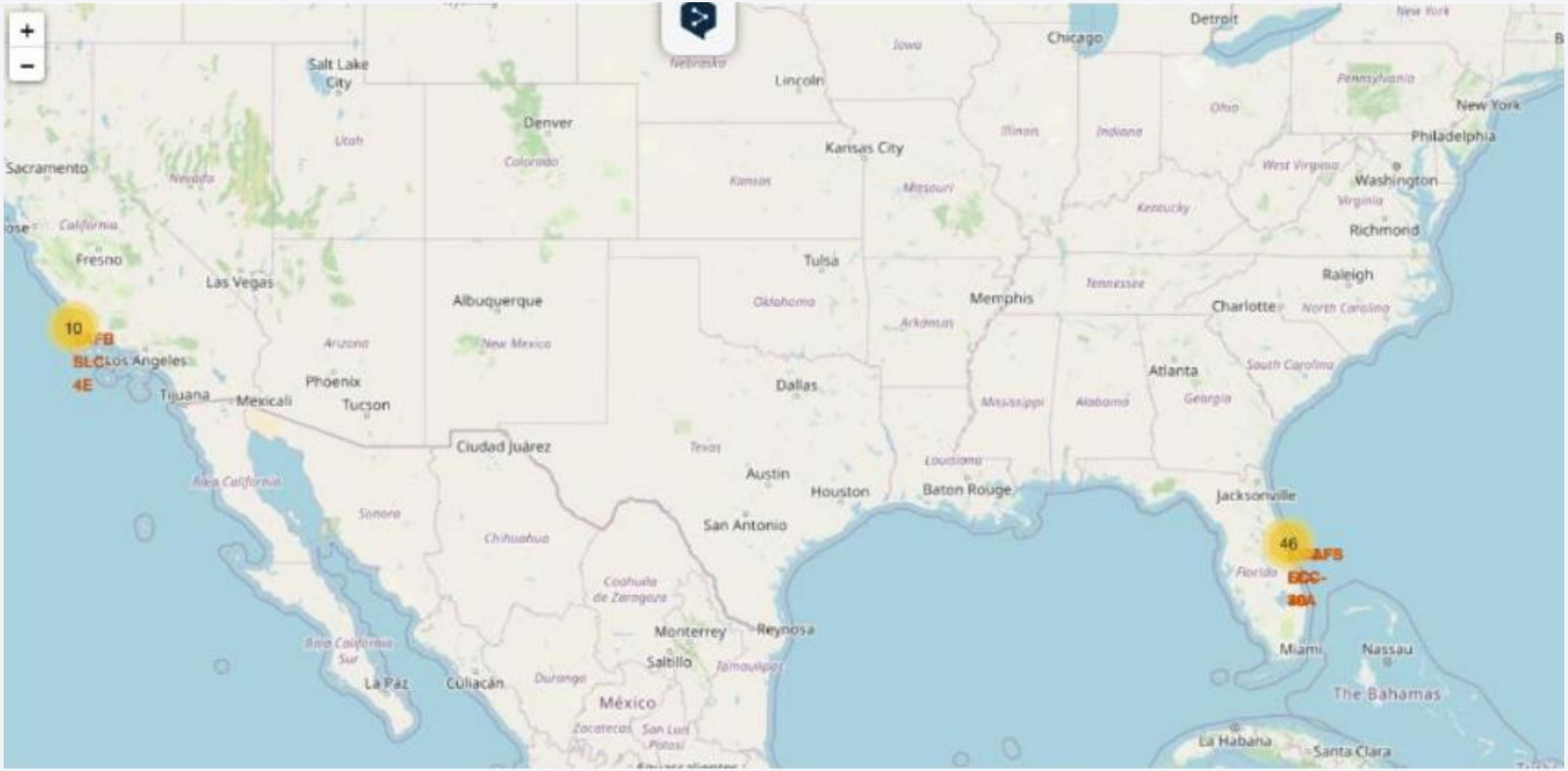
Landing_Outcome
No attempt
Success (ground pad)
Success (drone ship)
Success (drone ship)
Success (ground pad)
Failure (drone ship)
Success (drone ship)
Success (drone ship)
Success (drone ship)
Failure (drone ship)
Failure (drone ship)
Success (ground pad)
Precluded (drone ship)
No attempt
Failure (drone ship)
No attempt
Controlled (ocean)
Failure (drone ship)
Uncontrolled (ocean)

Section 3

Launch Sites Proximities Analysis

Launch site map

Launch site map

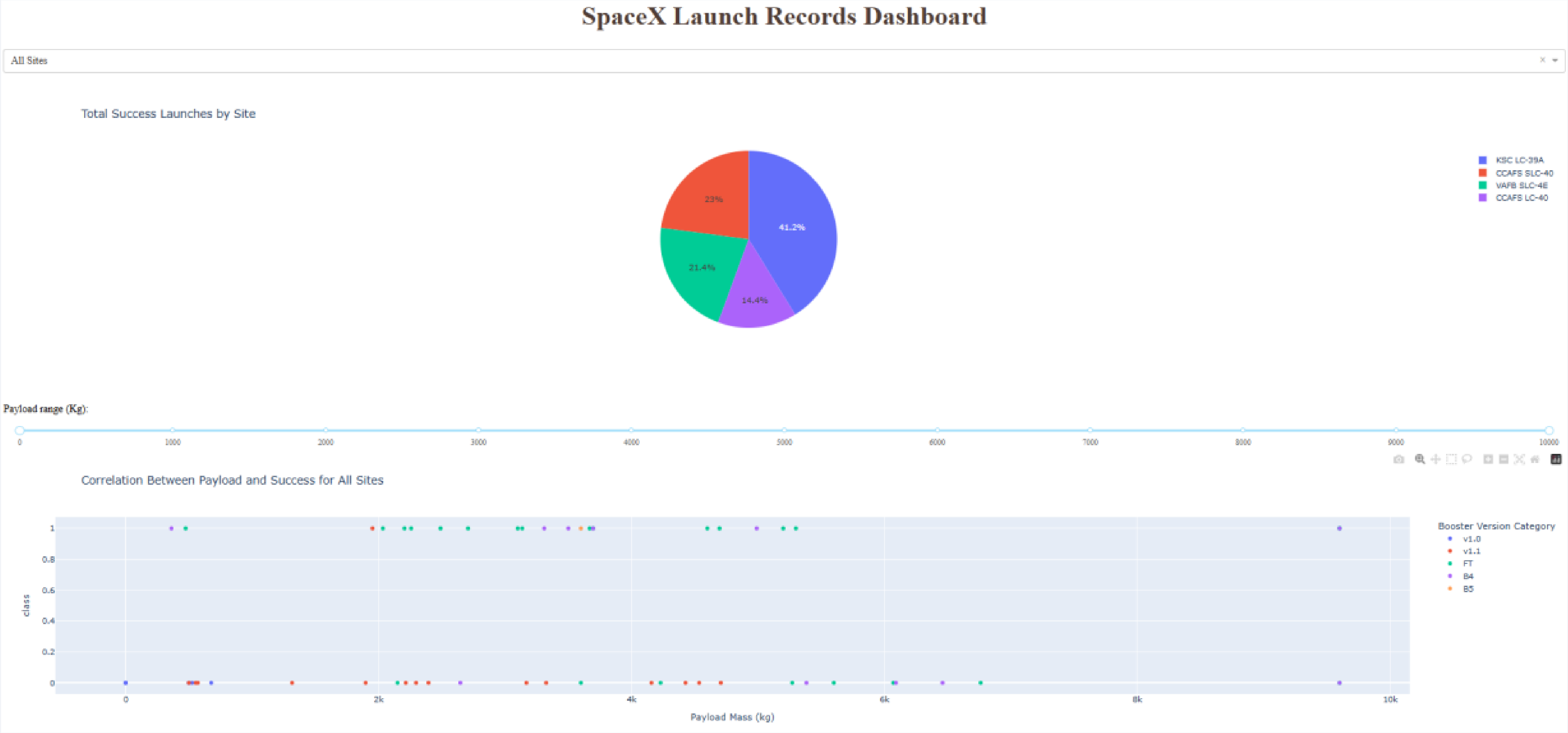


Launch site map with markers

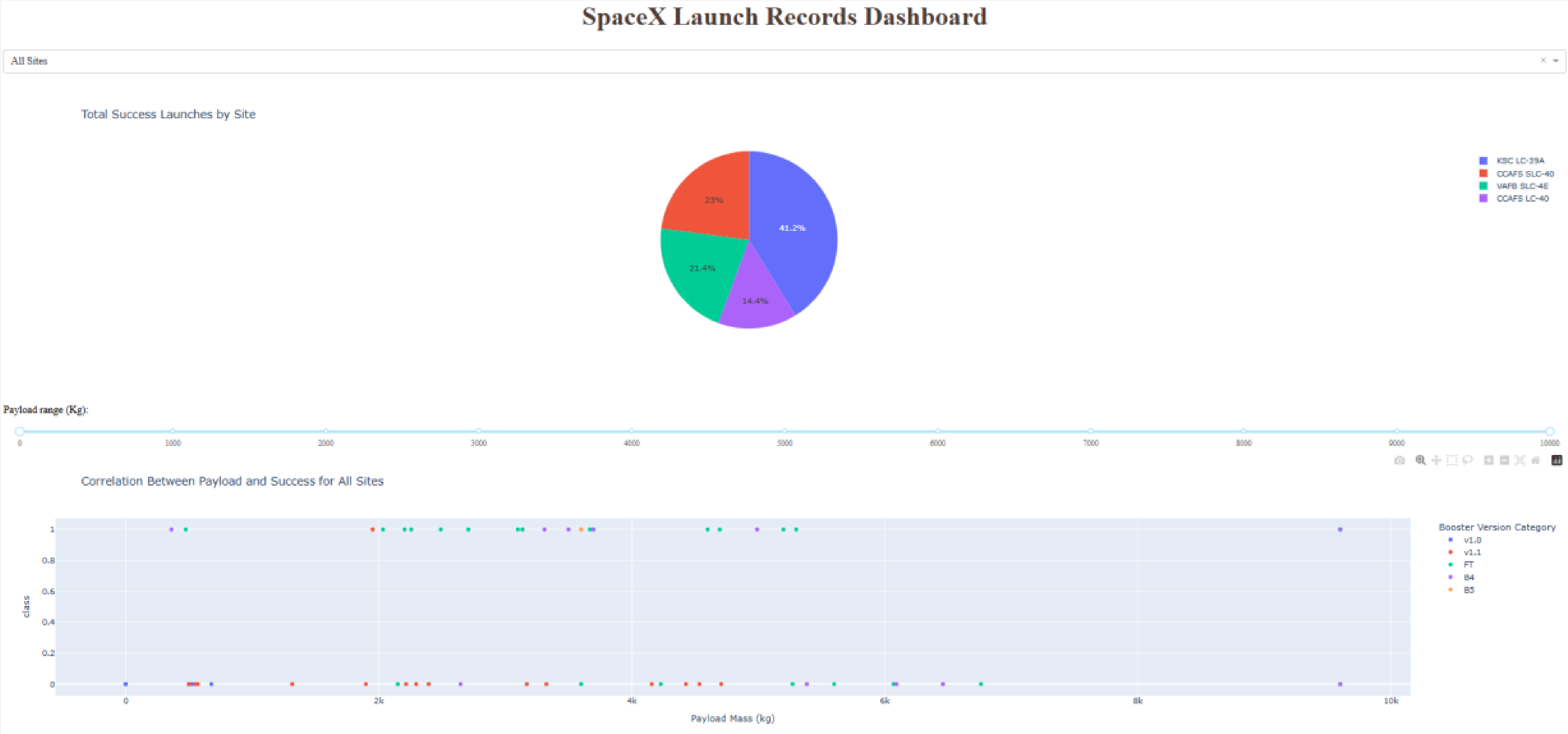
Section 4

Build a Dashboard with Plotly Dash

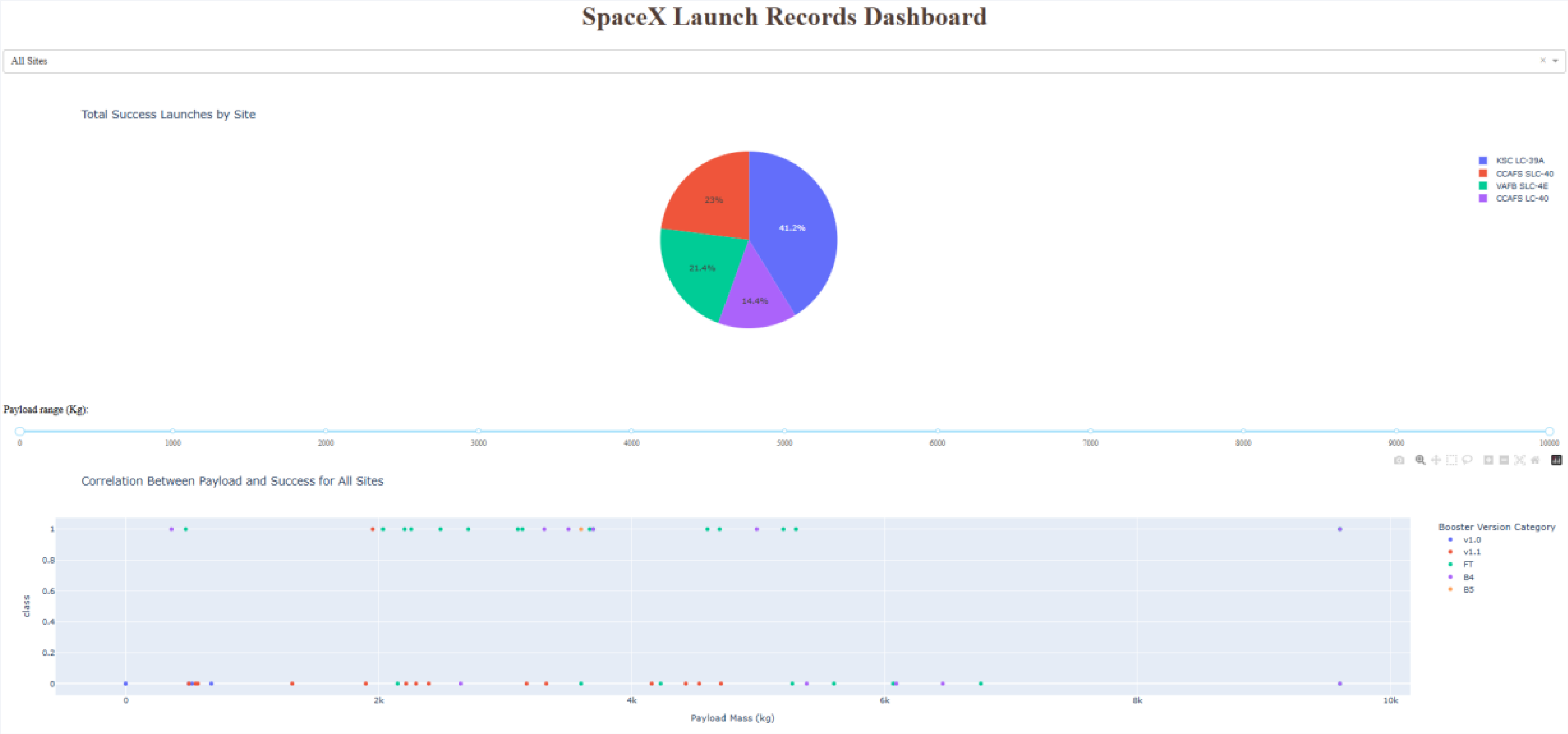
<Dashboard Screenshot 1>



<Dashboard Screenshot 2>



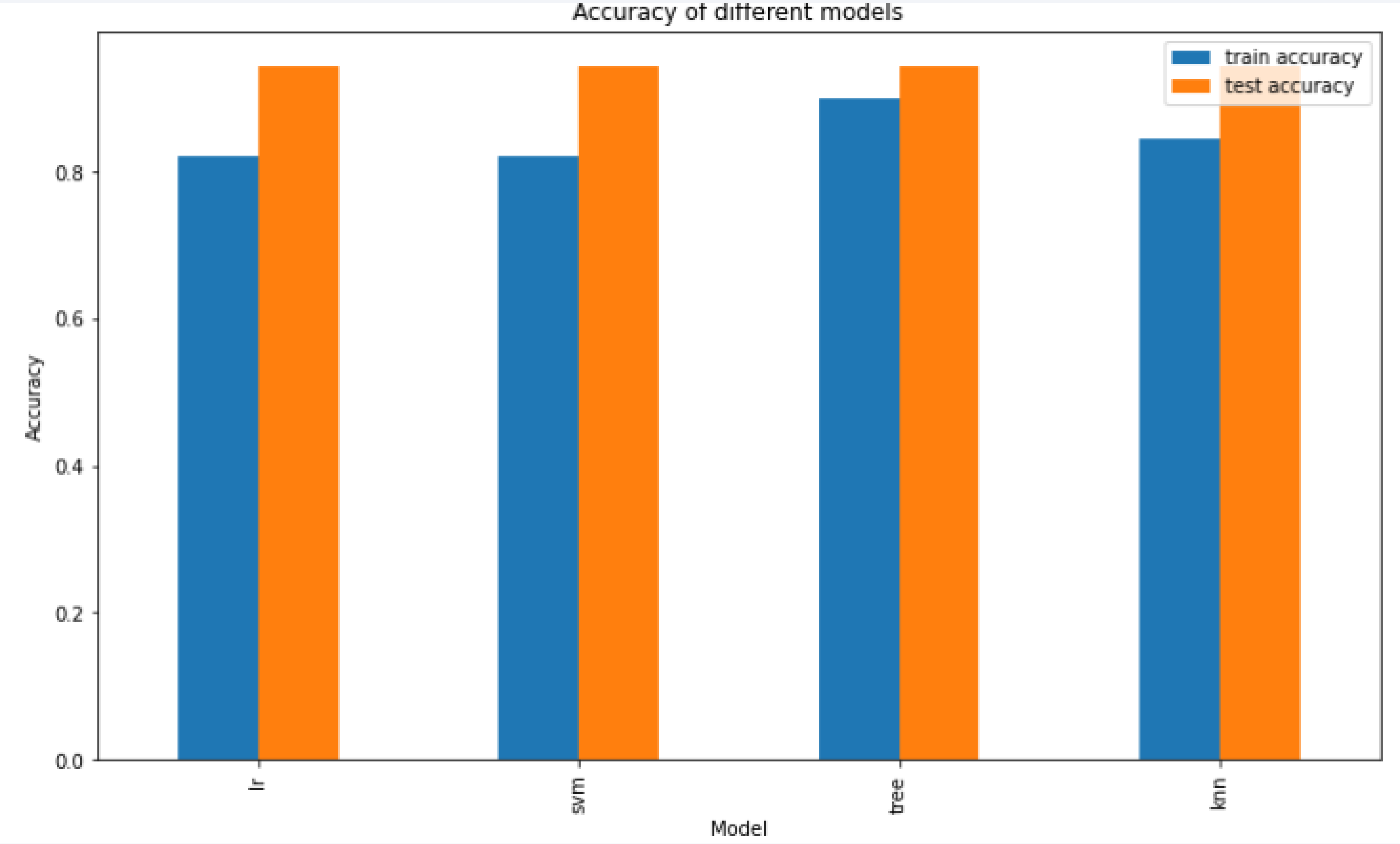
<Dashboard Screenshot 3>



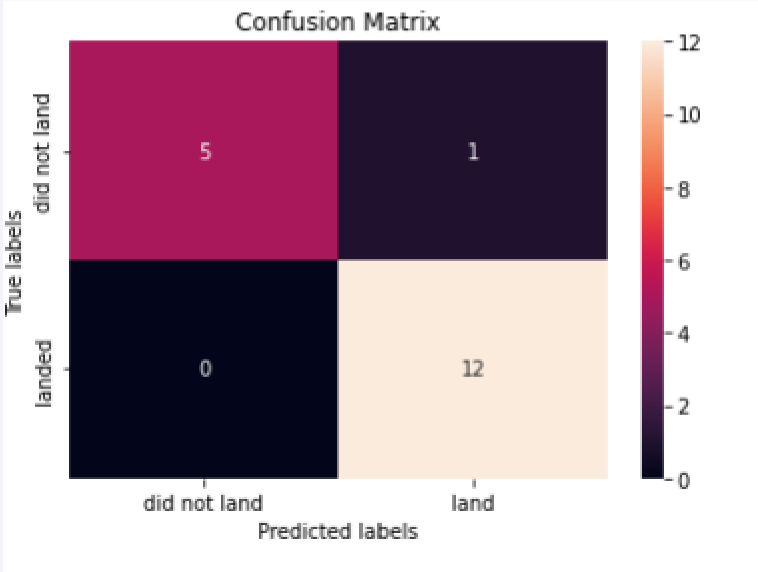
Section 5

Predictive Analysis (Classification)

Classification Accuracy



Confusion Matrix



TASK 11

Calculate the accuracy of tree_cv on the test data using the method `score` :

```
knn_score = knn_cv.score(X_test, Y_test)
knn_score
```

```
0.9444444444444444
```

We can plot the confusion matrix

```
yhat = knn_cv.predict(X_test)
plot_confusion_matrix(Y_test,yhat)
```

Conclusions

- LR, SVM, KNN are top-performing models for forecasting outcomes in this data

- Lighter paylods have a higher performance

- Launch sites are fare engough to make demage

Appendix

- Python code snippets
- SQL queries, charts
- Notebook outputs

Thank you!