

Winning Space Race with Data Science

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Outline

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

Executive Summary

- Summary of methodologies:
 - Problem: creating predictions and visualisations for SPACE Y future lanches
 - Collecting data from SPACE X public API
 - Data wrangling with python (Exploratory Data Analysis)
 - Model development (models as: logistic regression, support vector machine,

decision tree classifier, K nearest neighbours)

Summary of all results:
 Best accuracy method is sigmoid

Introduction

- SpaceX, founded by Elon Musk in 2002, is a private aerospace manufacturer and space transportation company
- Dataset of Space X's lunch missions
- Analyzing the success rate of Space X missions
- Predicting the outcome of future Space Y launches
- Understanding relationships between importanat variables
- Evaluating the performance

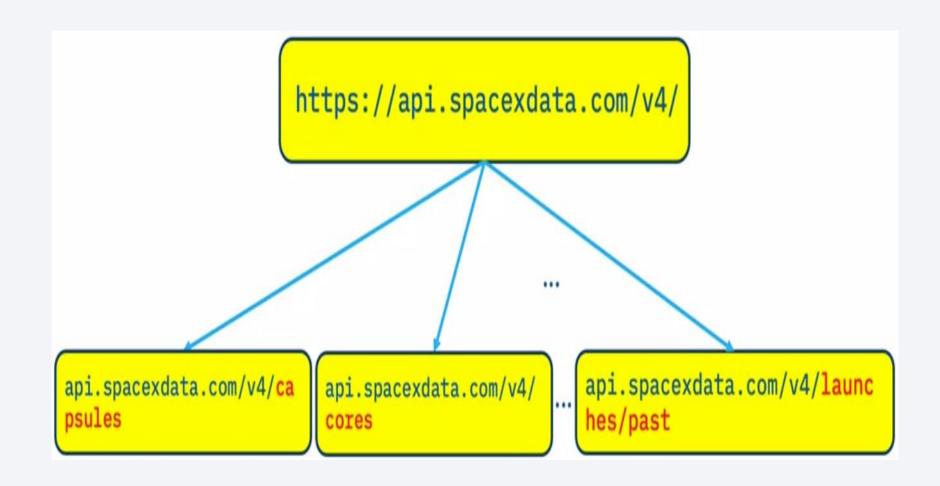


Methodology

Executive Summary

- Data collection methodology:
 - API Space X REST API
- Perform data wrangling
 - Json_normalize function generating a Pandas Dataframe
- 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, Decision trees, Support Vector Machines (SVM),
 K-Nearest Neighbors (KNN)

Data Collection – SpaceX API



Data Collection - Scraping



Data Wrangling

Data Preprocessing:

- Split the data into training and testing sets to evaluate model performance.
- Perform feature selection to identify the most relevant features for modeling.
- Address class imbalance issues if present by oversampling, undersampling, or using techniques like SMOTE.

EDA with Data Visualization

- Scatter Plot (Flight Number vs. Payload Mass): Used to visualize the relationship between the flight number and payload mass, aiming to understand any trends or patterns over time.
- Scatter Plot with Hue (Flight Number vs. Launch Site with Outcome as Hue): Utilized to
 explore the relationship between the flight number and launch site while incorporating
 the outcome of the launches as a hue, helping to identify any correlation between launch
 sites and mission outcomes.
- Scatter Plot with Hue (Payload Mass vs. Launch Site with Outcome as Hue): Employed to investigate the relationship between payload mass and launch site while considering the outcome of the launches as a hue, aiming to discern any impact of payload mass on mission success across different launch sites.
- Bar Chart (Success Rate of Each Orbit Type): Created to visualize the success rate of each orbit type, facilitating comparison and identification of trends in mission success across different orbit types.

EDA with SQL

- Queries:
- create table SPACEXTABLE as select * from SPACEXTBL where Date is not null
- select launch_site from SPACEXTABLE group by launch_site
- select * from SPACEXTABLE where launch_site like 'CCA%' limit 5
- select sum(PAYLOAD_MASS__KG_) as 'Total payload mass' from SPACEXTABLE where customer = 'NASA (CRS)'
- select avg(PAYLOAD_MASS__KG_) as 'Total payload mass' from SPACEXTABLE where booster_version like 'F9 v1.1%'
- select
- sum(case when Mission_Outcome = 'Success' then 1 else 0 end)
 as success, sum(case when Mission_Outcome = 'Failure' then 1
 else 0 end) as failure from SPACEXTABLE

Build an Interactive Map with Folium

- Markers: Added markers to indicate the location of launch sites on the map. Each marker represents a specific launch site used by SpaceX.
- Circles: Added circles to visualize the range of landing pads associated with each launch site. The circles represent the landing area where the rocket's first stage attempts to land after launch.
- Lines: Added lines to connect launch sites with their corresponding landing pads, providing a visual representation of the launch trajectory and landing area.
- The purpose of adding these objects is to provide a comprehensive view of SpaceX's launch operations and infrastructure. By visualizing the launch sites, landing pads, and launch trajectories on the map, stakeholders can gain insights into the geographic distribution of SpaceX's activities and understand the proximity of launch sites to potential landing areas.

Build a Dashboard with Plotly Dash

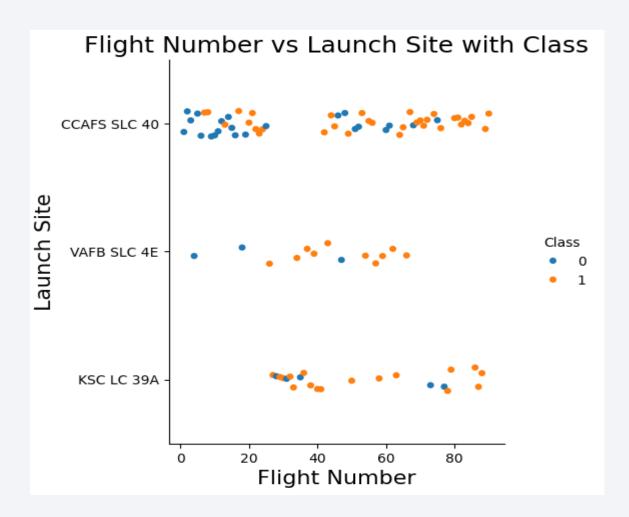
- Line Chart: Showing the trend of SpaceX mission outcomes over time, allowing users to visualize the success rate of missions.
- Bar Chart: Displaying the distribution of mission outcomes (success/failure) across different launch sites, providing insights into the performance of each site.
- Scatter Plot: Illustrating the relationship between payload mass and mission outcome, helping users understand how payload mass affects mission success.
- Pie Chart: Presenting the distribution of mission outcomes for different orbit types, allowing users to compare success rates across orbits.

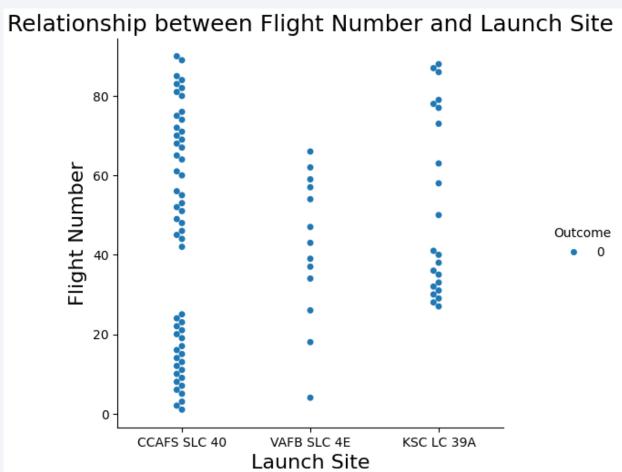
Predictive Analysis (Classification)

- Summarize how you built, evaluated, improved, and found the best performing classification model
- You need present your model development process using key phrases and flowchart
- Add the GitHub URL of your completed predictive analysis lab, as an external reference and peer-review purpose

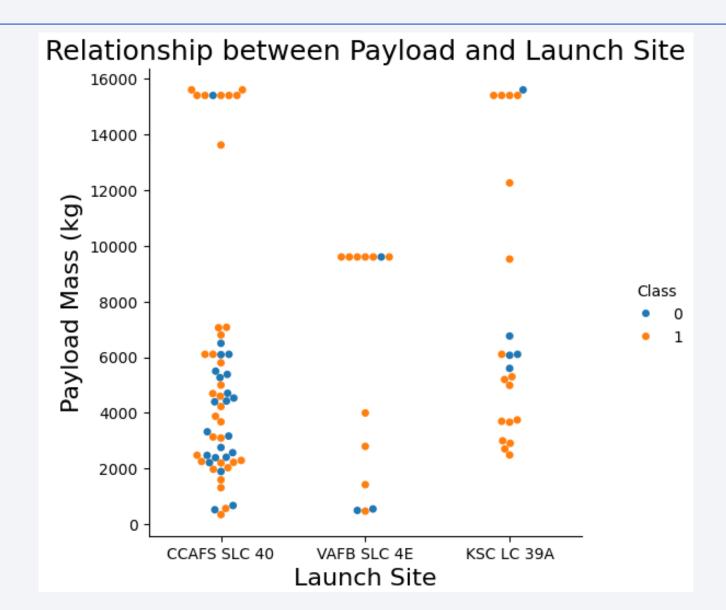


Flight Number vs. Launch Site

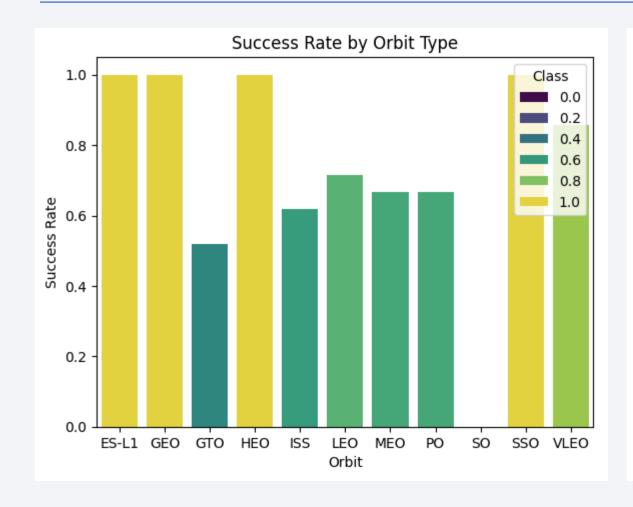


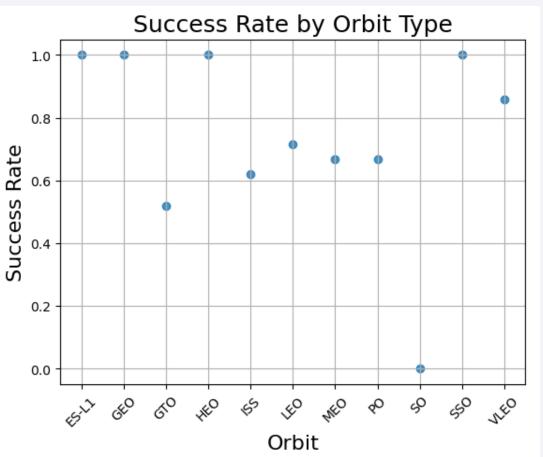


Payload vs. Launch Site

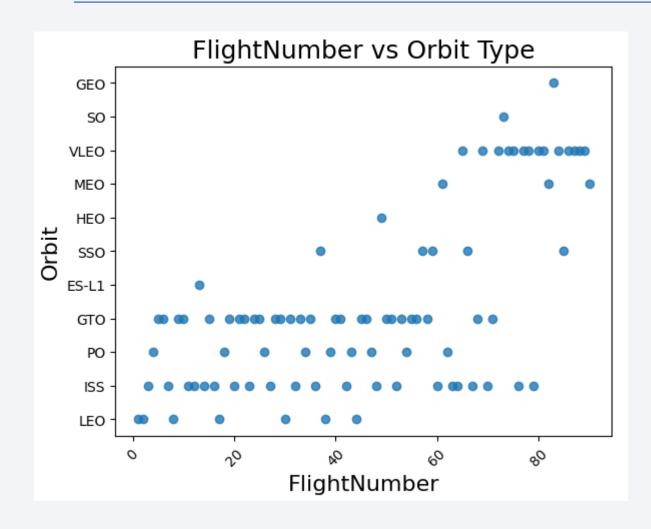


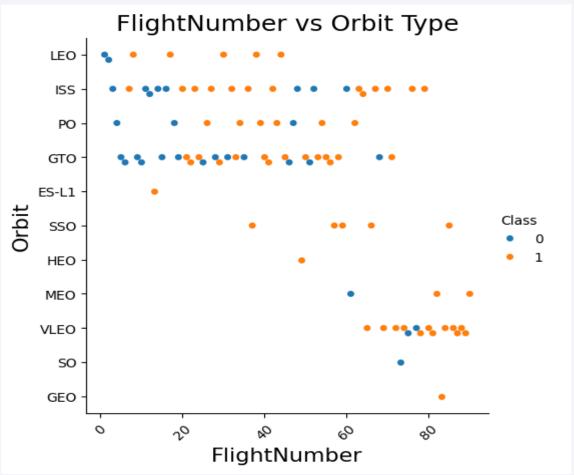
Success Rate vs. Orbit Type



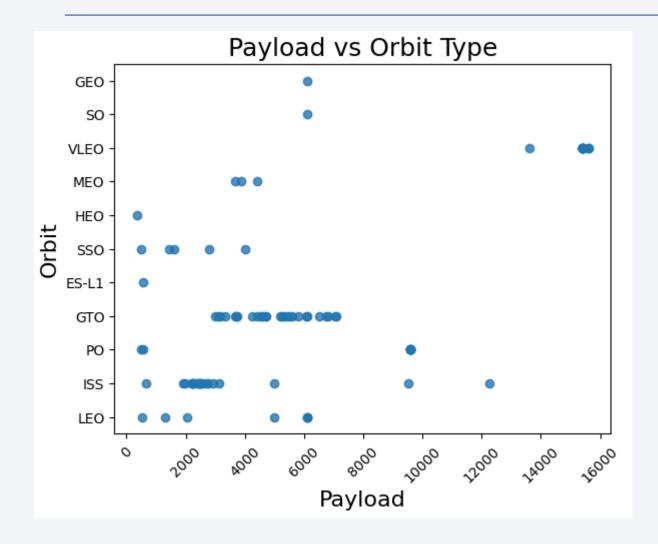


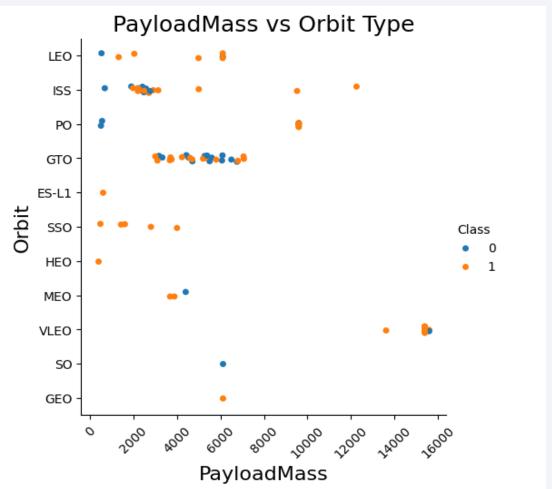
Flight Number vs. Orbit Type





Payload vs. Orbit Type





Launch Success Yearly Trend



All Launch Site Names

```
%sql select launch_site from SPACEXTABLE group by launch_site
* sqlite:///my_data1.db
Done.
 Launch_Site
 CCAFS LC-40
CCAFS SLC-40
  KSC LC-39A
 VAFB SLC-4E
```

Launch Site Names Begin with 'CCA'

%sql select * from SPACEXTABLE where launch_site like 'CCA%' limit 5 * sqlite:///my data1.db Done. Date Time (UTC) Booster_Version Launch_Site Payload PAYLOAD_MASS_KG_ Customer Mission Outcome Landing Outcome Orbit Dragon Spacecraft Qualification Unit Success Failure (parachute) 2010-06-04 F9 v1.0 B0003 CCAFS LC-40 LEO 18:45:00 SpaceX Success Failure (parachute) 2010-12-08 F9 v1.0 B0004 CCAFS LC-40 Dragon demo flight C1, two CubeSats, barrel of Brouere cheese 0 LEO (ISS) NASA (COTS) NRO 15:43:00 Dragon demo flight C2 2012-05-22 7:44:00 F9 v1.0 B0005 CCAFS LC-40 525 LEO (ISS) NASA (COTS) No attempt Success F9 v1.0 B0006 CCAFS LC-40 2012-10-08 No attempt 0:35:00 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) No attempt Success

Total Payload Mass

```
%sql select sum(PAYLOAD_MASS__KG_) as 'Total payload mass' from SPACEXTABLE where customer = 'NASA (CRS)'
  * sqlite://my_data1.db
Done.
  Total payload mass
  45596
```

Average Payload Mass by F9 v1.1

```
%sql select avg(PAYLOAD_MASS__KG_) as 'Total payload mass' from SPACEXTABLE where booster_version like 'F9 v1.1%'

* sqlite://my_data1.db
Done.

Total payload mass

2534.666666666666665
```

First Successful Ground Landing Date

```
%sql select min(date) from SPACEXTABLE where Landing_Outcome like 'Success (ground%'
  * sqlite://my_data1.db
Done.
  min(date)
2015-12-22
```

Successful Drone Ship Landing with Payload between 4000 and 6000

%sql select Booster_Version from SPACEXTABLE where Landing_Outcome like 'Success (drone%' and PAYLOAD_MASS__KG_ between 4000 and 6000 group by Booster_Version
 * sqlite://my_data1.db
Done.

Booster_Version

F9 FT B1021.2

F9 FT B1022

F9 FT B1026

Total Number of Successful and Failure Mission Outcomes

```
%%sql
select
sum(case when Mission_Outcome = 'Success' then 1 else 0 end) as success, sum(case when Mission_Outcome = 'Failure' then 1 else 0 end) as failure from SPACEXTABLE

* sqlite:///my_data1.db
Done.
success failure

98 0
```

Boosters Carried Maximum Payload

%%sql select booster version, PAYLOAD MASS KG from SPACEXTABLE where PAYLOAD MASS KG = (select max(PAYLOAD MASS KG) from SPACEXTABLE) group by booster version, PAYLOAD MASS KG * sqlite:///my data1.db Done. Booster Version PAYLOAD MASS KG F9 B5 B1048.4 15600 F9 B5 B1048.5 15600 F9 B5 B1049.4 15600 F9 B5 B1049.5 15600 F9 B5 B1049.7 15600 F9 B5 B1051.3 15600 F9 B5 B1051.4 15600 F9 B5 B1051.6 15600 F9 B5 B1056.4 15600 F9 B5 B1058.3 15600 F9 B5 B1060.2 15600 F9 B5 B1060.3 15600

2015 Launch Records

```
%%sql
select
substr(Date, 6,2) month,
Landing Outcome,
booster version,
launch site
from SPACEXTABLE
where Landing_Outcome = 'Failure (drone ship)' and date like '2015%'
 * sqlite:///my_data1.db
Done.
month Landing_Outcome Booster_Version Launch_Site
   01 Failure (drone ship) F9 v1.1 B1012 CCAFS LC-40
   04 Failure (drone ship) F9 v1.1 B1015 CCAFS LC-40
```

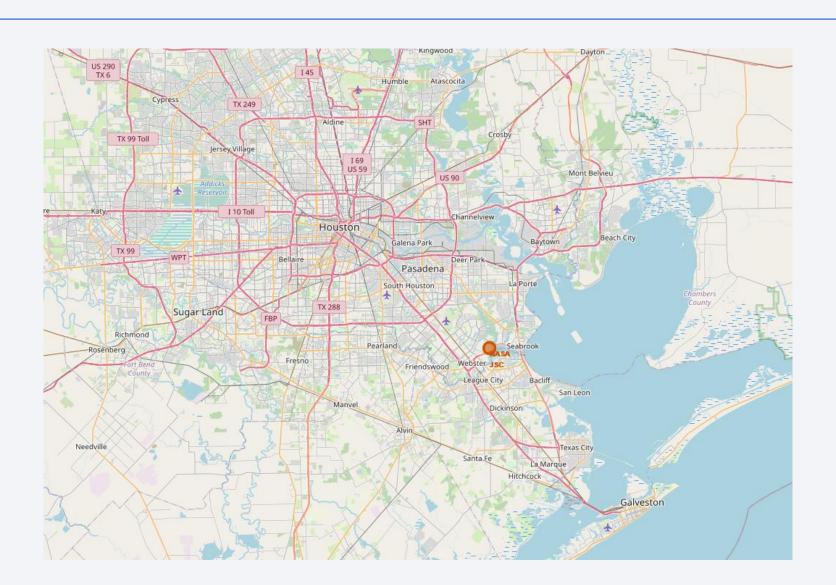
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

```
%sql select Landing Outcome, count(*) from SPACEXTABLE where date between '2010-06-04' and '2017-03-20' group by Landing Outcome order by 2 desc
 * sqlite:///my data1.db
Done.
   Landing_Outcome count(*)
         No attempt
                           10
  Success (drone ship)
   Failure (drone ship)
                            5
 Success (ground pad)
                            3
   Controlled (ocean)
                            3
 Uncontrolled (ocean)
   Failure (parachute)
                            2
Precluded (drone ship)
```



Launch Sites Locations Analysis with Folium

Launch sites are marked with airplanes



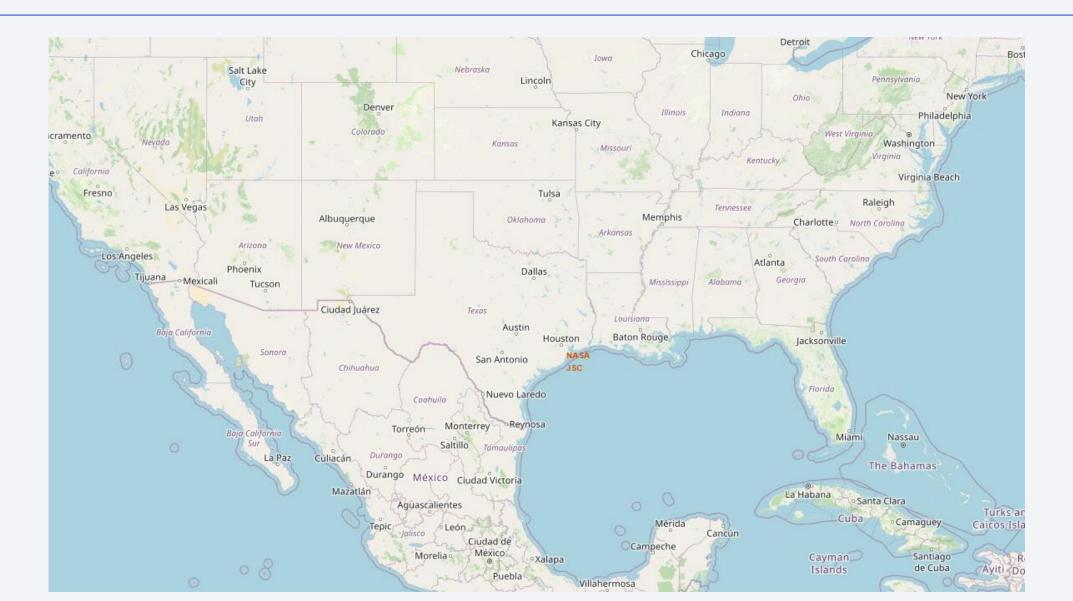
<Folium Map Screenshot 2>

Replace <Folium map screenshot 2> title with an appropriate title

 Explore the folium map and make a proper screenshot to show the colorlabeled launch outcomes on the map

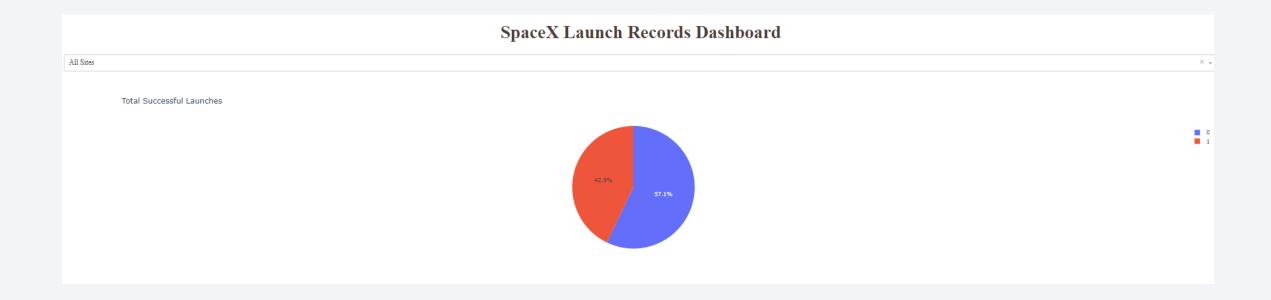
Explain the important elements and findings on the screenshot

<Folium Map Screenshot 3>



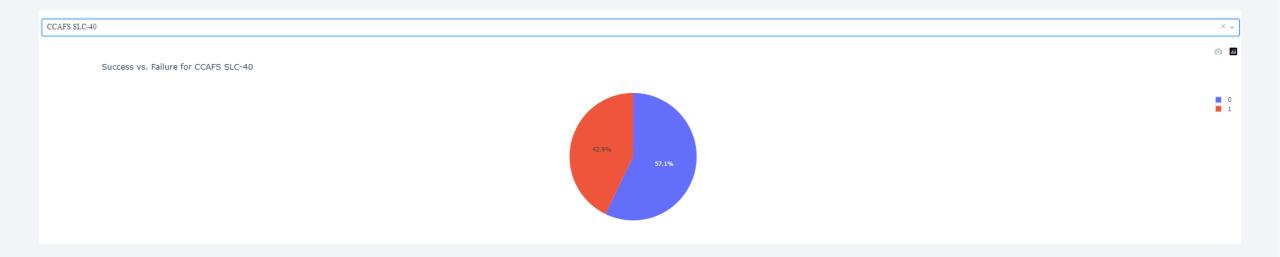


Dashboard Application with Plotly Dash



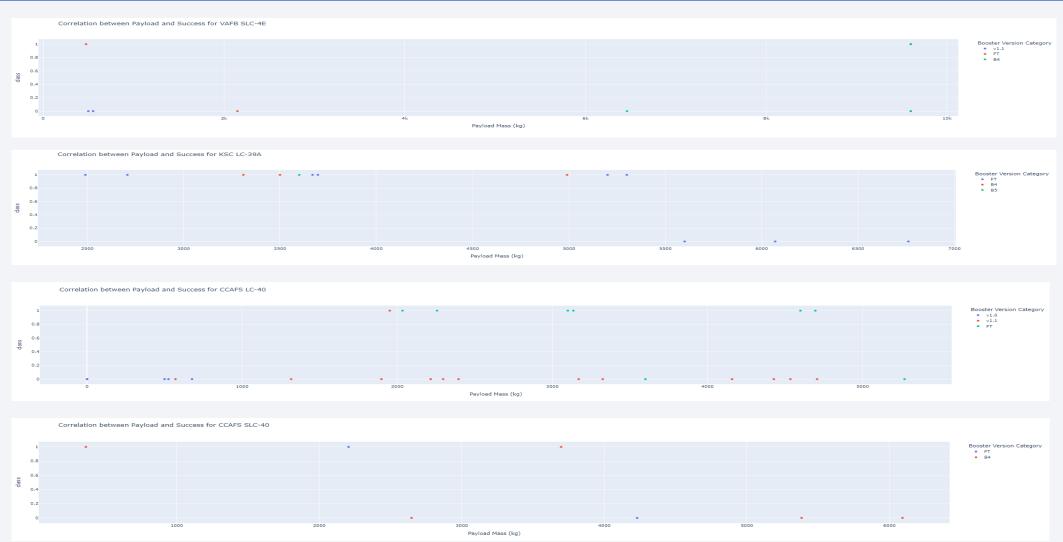
In 57.1% launches finished with a bad outcome like the booster did not land 😊

Piechart of the most successful ratio



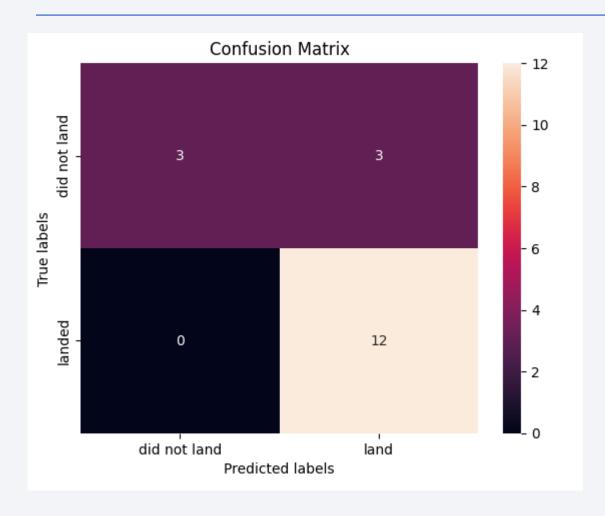
• The most successful site was CCAFS SLC-40

< Dashboard Screenshot 3>





Confusion Matrix



The best performing method is Sigmoid, because:

```
tree accuracy = tree cv.score(X test, Y test)
svm_accuracy = svm_cv.score(X_test, Y_test)
knn_accuracy = knn_cv.score(X_test, Y_test)
# Accuracies
print("Accuracy for Decision Tree:", tree accuracy)
print("Accuracy for Support Vector Machine:", svm accuracy)
print("Accuracy for K-Nearest Neighbors:", knn accuracy)
# Find the method with the highest accuracy
best method = max(tree accuracy, svm accuracy, knn accuracy)
if best_method == tree_accuracy:
    print("Decision Tree performs best.")
elif best_method == svm_accuracy:
    print("Support Vector Machine performs best.")
else:
   print("K-Nearest Neighbors performs best.")
best_params_svm = svm_cv.best_params_
# Extract the best kernel
best_kernel = best_params_svm['kernel']
print("Best kernel for Support Vector Machines:", best_kernel)
Accuracy for Support Vector Machine: 0.8333333333333333
Accuracy for K-Nearest Neighbors: 0.83333333333333334
Decision Tree performs best.
Best kernel for Support Vector Machines: sigmoid
```

