

Winning Space Race with Data Science

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- Methodology
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- Conclusion
- Appendix

Executive Summary

- Summary of methodologies
 - Data Collection through API
 - Data Collection with Web Scraping
 - Data Wrangling
 - Exploratory Data Analysis with SQL
 - Exploratory Data Analysis with Data Visualization
 - Interactive Visual Analytics with Folium
 - Machine Learning Prediction
- Summary of all results
 - Exploratory Data Analysis result
 - Interactive analytics in screenshots
 - Predictive Analytics result

Introduction

Project background and context

The Falcon 9 is a reusable two-stage rocket designed and manufactured by SpaceX to reliably and safely transport people and payloads to Earth orbit and beyond. The Falcon 9 is the world's first reusable orbital-class rocket. Reusability allows SpaceX to remanufacture the most expensive parts of the rocket, which in turn reduces the cost of access to space. Therefore, if we can determine if the first stage will land, we can determine the cost of a launch. This goal of the project is to create a machine learning pipeline to predict if the first stage will land successfully.

Problems you want to find answers

The effect each relationshop with certain rocket variables will impact determining the success rate of a successful landing, and what conditions does SpaceX have to achieve to get the best results



Methodology

Executive Summary

- Data collection methodology:
 - SpaceXAPI(https://api.spacexdata.com/v4/rockets/)
 - WebScraping
- Perform data wrangling
 - Identified missing values
 - Replace missing values with mean
 - Identifies Null Values
 - Created a training label class.

Methodology

Executive Summary

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
- Data that was collected until this step were normalized, divided in training and test data sets and evaluated by four different classification models, being the accuracy of each model evaluated using different combinations of parameters.

Data Collection

- Datasets collected from:
 - SpaceX API (https://api.spacexdata.com/v4/rockets/)
 - Wikipedia
 (https://en.wikipedia.org/wiki/List_of_Falcon_9_and_Falcon_Heavy_launches), using web scraping.

Data Collection – SpaceX API

- Using a SpaceX public API, like flowchart:
- Source code:

https://github.com/heliospjunior/IBM-Data-Science-Capstone-SpaceX/blob/main/spacex-data-collection-api.ipynb



Data Collection - Scraping

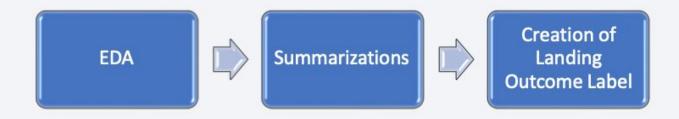
- Data from SpaceX launches obtained from Wikipedia, used like flowchart:.
- Source code:

https://github.com/heliospjunior/IBM-Data-Science-Capst one-SpaceX/blob/main/webscraping.ipynb



Data Wrangling

 Collect data, analysis and check missing data and data types, clean and format data types, change data types if necessary, summary launches per site, and outcome label created from outcome product.



• Source code:

https://github.com/heliospjunior/IBM-Data-Science-Capstone-SpaceX/blob/main/spacex-Data%20wrangling.ipynb

EDA with Data Visualization

Scatterplots and bar plots were used to visualize the relationship between pair of features:

- Payload Mass X Flight Number,
- Launch Site X Flight Number
- Launch Site X Payload Mass
- Orbit and Flight Number
- Payload and Orbit

Bar plots were used to visualize the relationship between pair of features:

Mean X Orbit

Line Graphs were used to visualize the relationship between pair of features:

Mean X Orbit

EDA with SQL

- SQL queries performed:
 - Display the names of the unique launch sites in the space mission;
 - Display the top 5 launch sites whose name begins with the string 'CCA';
 - Display total pay load mass carried by boosters launched by NASA (CRS);
 - Average payload mass carried by booster version F9 v1.1;
 - Date of the first successful landing outcome in ground pad;
 - Listing names of the boosters which have success in drone ship with payload mass between 4000-6000 kg;
 - Total number of successful and failure mission outcomes;
 - Listing names of the booster versions which have carried the maximum payload mass;
 - Failed landing out comes in droneship, their booster versions, and launch site names for in year 2015; and
 - Rank of the count of landing outcomes (such as Failure (droneship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20.
- Source code: https://github.com/heliospjunior/IBM-Data-Science-Capstone-SpaceX/blob/main/eda-sql.ipynb

Build an Interactive Map with Folium

- Launch success rate may depend on the location and proximity of a launch site
- Markers, circles, lines and marker clusters were used with Folium Interative Maps
- Using Folium library, identified all SpaceX launch sites on a map
- Marker clusters indicates groups of events in each coordinate, like launches in a launch site; and lines are used to indicate distances between two coordinates.

• Source code: https://github.com/heliospjunior/IBM-Data-Science-Capstone-SpaceX/blob/main/launch_site_location.ipynb

Build a Dashboard with Plotly Dash

Build an interactive dashboard with Plotly Dash including:

- Dropdown menu for select sites
- Pie charts displaying success rate.
- Scatter graph with the relationship about Outcome and Payload Mass (Kg) for the different booster version.
- Range Slider for select range of payload mass (Kg)

Predictive Analysis (Classification)

- EDA using numpy and pandas, transforme data and split our data into training and testing.
- Built some machine learning models and tune different hyperparameters using GridSearchCV.
- Used accuracy as the metric for our model, improved the model using feature engineering and algorithm tuning.
- Found the best performing classification model.

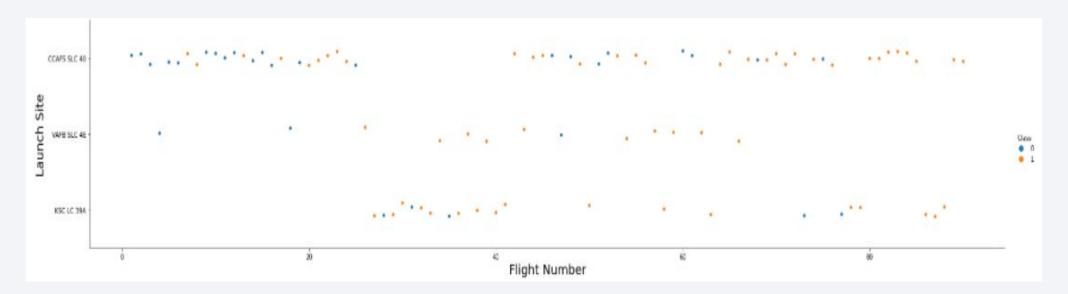
Results

- Exploratory data analysis results:
 - Space X uses 4 different launch sites;
 - The first launches were done to Space X itself and NASA;
 - The average payload of F9 v1.1 booster is 2,928 kg;
 - The first success landing outcome happened in 2015 fiver year after the first launch;
 - Many Falcon 9 booster versions were successful at landing in drone ships having payload above the average;
 - Almost 100% of mission outcomes were successful;
 - The number of landing outcomes became as better as years passed.
 - Using interactive analytics was possible to identify that launch sites use to be in safety places, near sea, for example and have a good logistic infrastructure around.
 - Most launches happens at east cost launch sites.



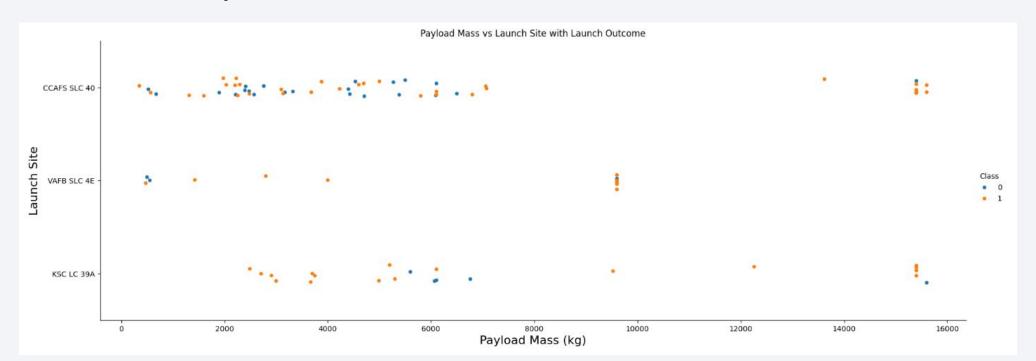
Flight Number vs. Launch Site

- CCAFS-SL 40 is the most used launch site.
- CCAFS-SLC 40 has most of failures in the early stage of Falcon9



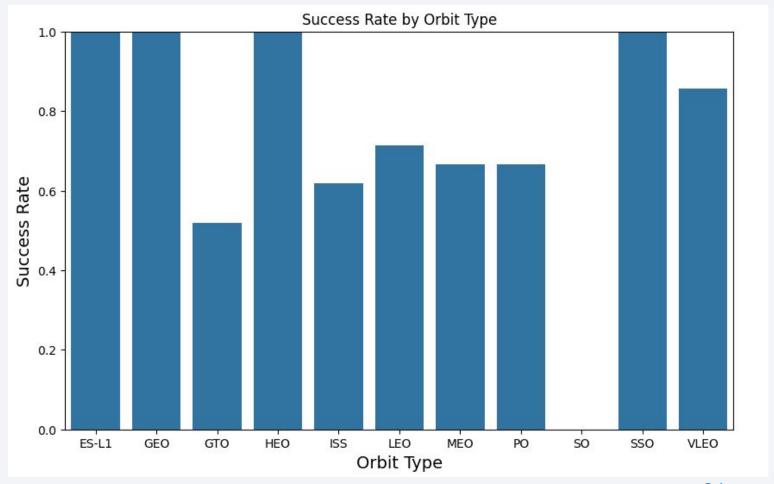
Payload vs. Launch Site

- Falcon9 specifications, heavy payloads sent to low/medium orbits.
- The percentage of failures is lower for heavy payloads.
- Success Rate X Payload Orbit needed some more information.



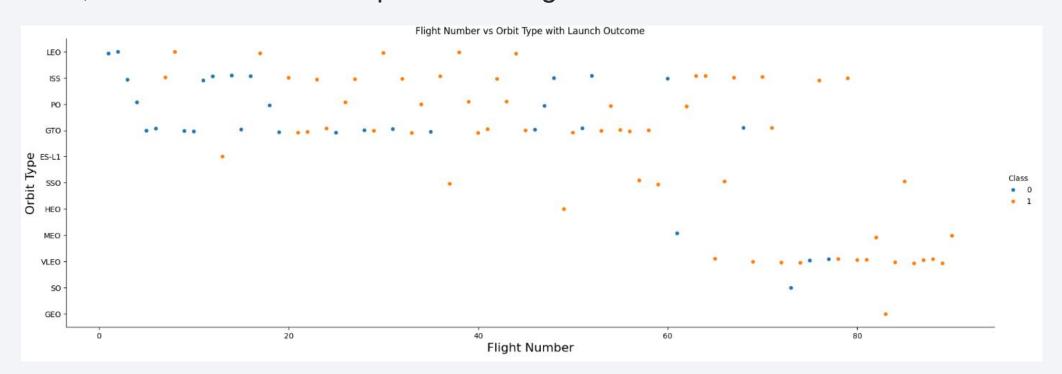
Success Rate vs. Orbit Type

- From the plot, we can see that ES-L1, GEO, HEO, SSO, VLEO had the most success rate.
- GTO sees the lowesr success rate



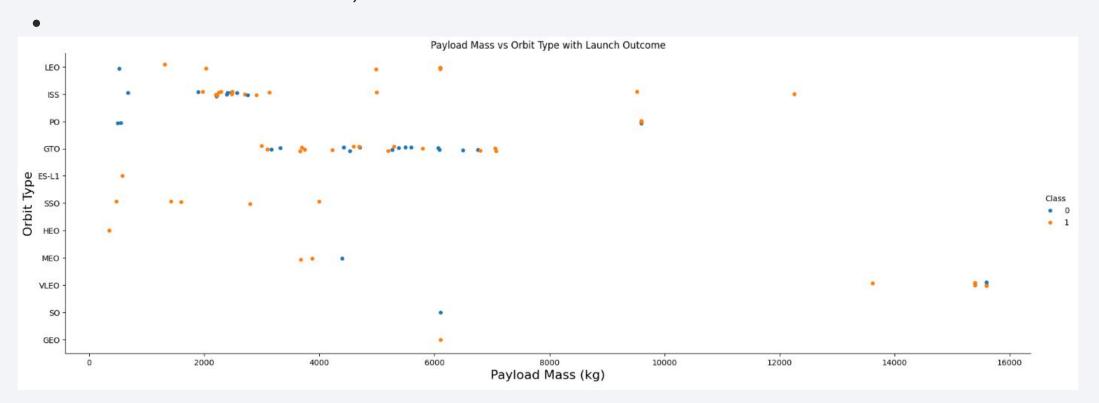
Flight Number vs. Orbit Type

• In the LEO orbit, success is related to the number of flights whereas in the GTO orbit, there is no relationship between flight number and the orbit.



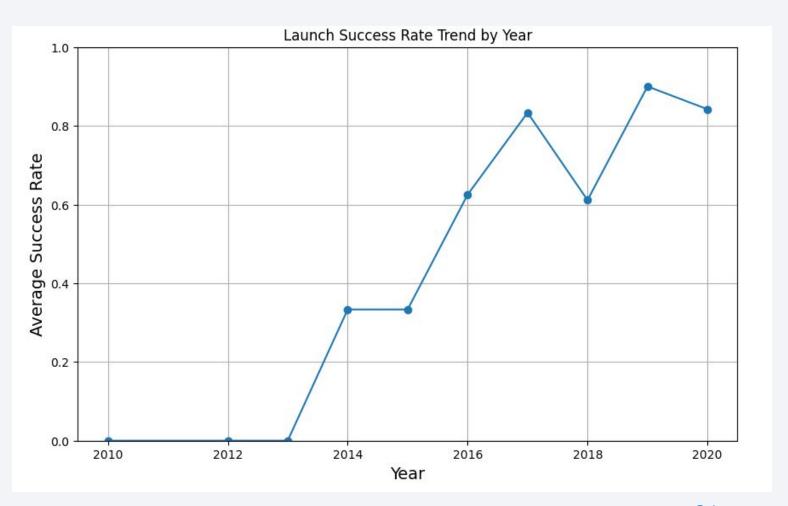
Payload vs. Orbit Type

- Max Success rate with low orbit, except ISS and low payload mass
- Between 2000-7500 ks, success rate seems to be distributed for GTO



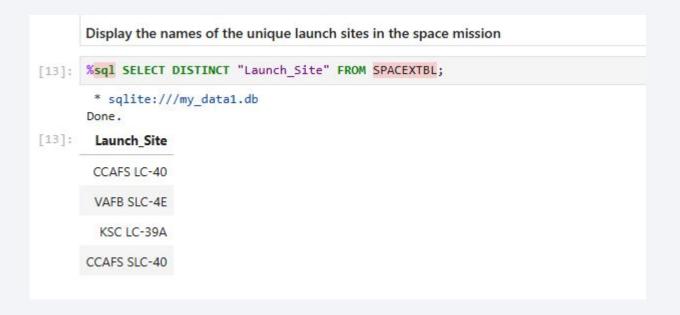
Launch Success Yearly Trend

- The success rate since 2013 kept on increasing till 2020.
- Falcon9 average booster recovery success rate 66%

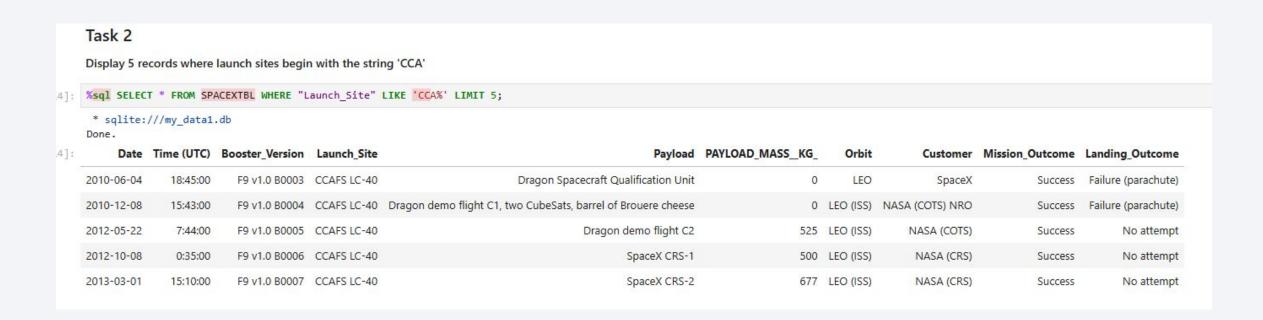


All Launch Site Names

• With **DISTINCT** we show only unique launch sites from the SpaceX data.



Launch Site Names Begin with 'CCA'



Query to display 5 records where launch sites begin with `CCA`

Total Payload Mass

 Total payload carried by boosters from NASA as 45596 using the query below

```
Task 3

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

5]: %sql SELECT SUM("Payload_Mass_kg_") AS Total_Payload_Mass FROM SPACEXTBL WHERE "Customer" = 'NASA (CRS)';

* sqlite://my_data1.db
Done.

5]: Total_Payload_Mass

45596
```

Average Payload Mass by F9 v1.1

Calculated the average payload mass carried by booster version F9 v1.1 as 2928.4

```
Task 4

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

[16]: %sql SELECT AVG("Payload_Mass_kg_") AS Average_Payload_Mass FROM SPACEXTBL WHERE "Booster_Version" = 'F9 v1.1';

* sqlite://my_data1.db
Done.

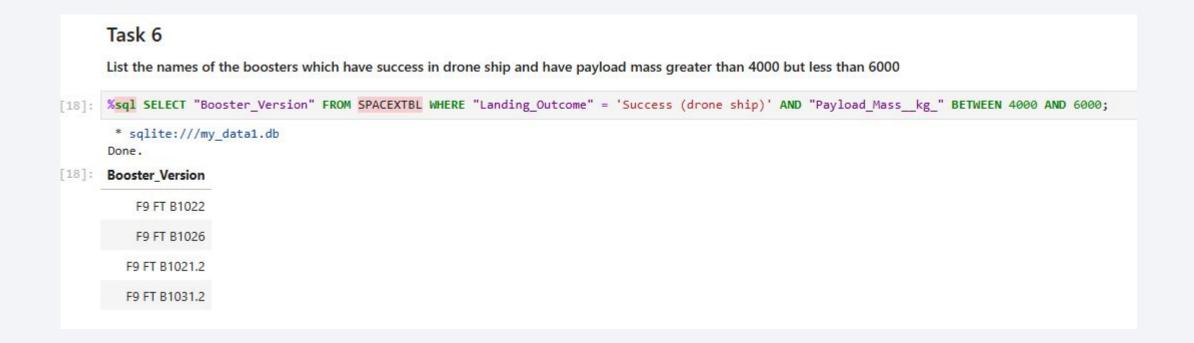
[16]: Average_Payload_Mass

2928.4
```

First Successful Ground Landing Date

The dates of the first successful landing outcome on ground pad was 22nd December 2015

Successful Drone Ship Landing with Payload between 4000 and 6000



Total Number of Successful and Failure Mission Outcomes

 Wildcard like '%' to filter for WHERE MissionOutcome was a success or a failure.

```
List the total number of successful and failure mission outcomes
In [16]:
          task 7a = '''
                  SELECT COUNT(MissionOutcome) AS SuccessOutcome
                  FROM SpaceX
                  WHERE MissionOutcome LIKE 'Success%'
          task_7b = '''
                  SELECT COUNT(MissionOutcome) AS FailureOutcome
                  FROM SpaceX
                  WHERE MissionOutcome LIKE 'Failure%'
          print('The total number of successful mission outcome is:')
          display(create pandas df(task 7a, database=conn))
          print()
          print('The total number of failed mission outcome is:')
          create pandas df(task 7b, database=conn)
         The total number of successful mission outcome is:
            successoutcome
         0
                      100
         The total number of failed mission outcome is:
Out[16]:
            failureoutcome
         0
```

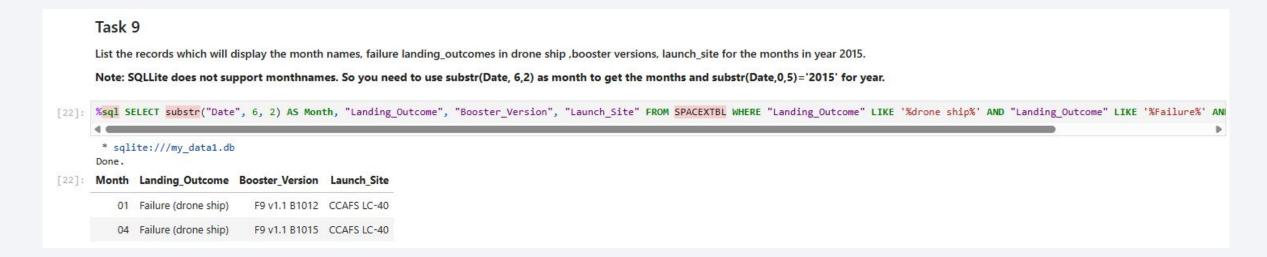
Boosters Carried Maximum Payload

• The booster that have carried the maximum payload using a subquery in the **WHERE** clause and the **MAX()** function.



2015 Launch Records

 Used WHERE clause, LIKE, AND, and BETWEEN conditions to filter for failed landing outcomes in drone ship, their booster versions, and launch site names for year 2015



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

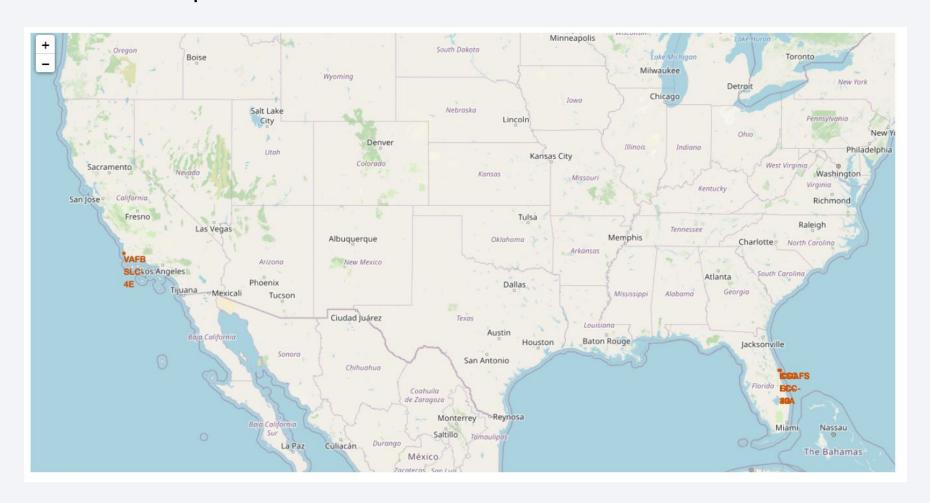
 Applied the GROUP BY clause to group the landing outcomes and the ORDER BY clause to order the grouped landing outcome in descending order.



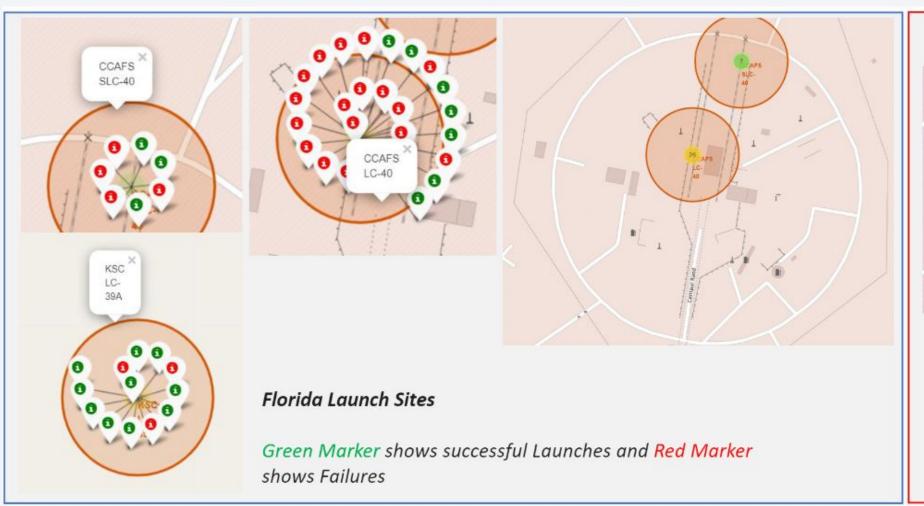


Launch sites global map

• We can see the Space X launch sites in the USA coasts.

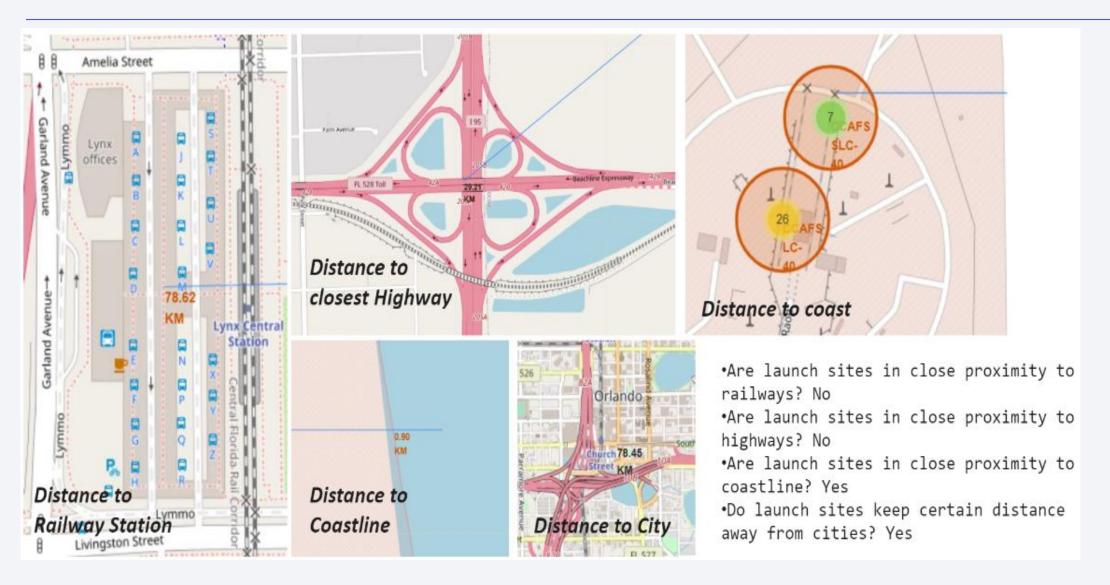


Markers showing launch sites with color labels





Launch Site distance to landmarks

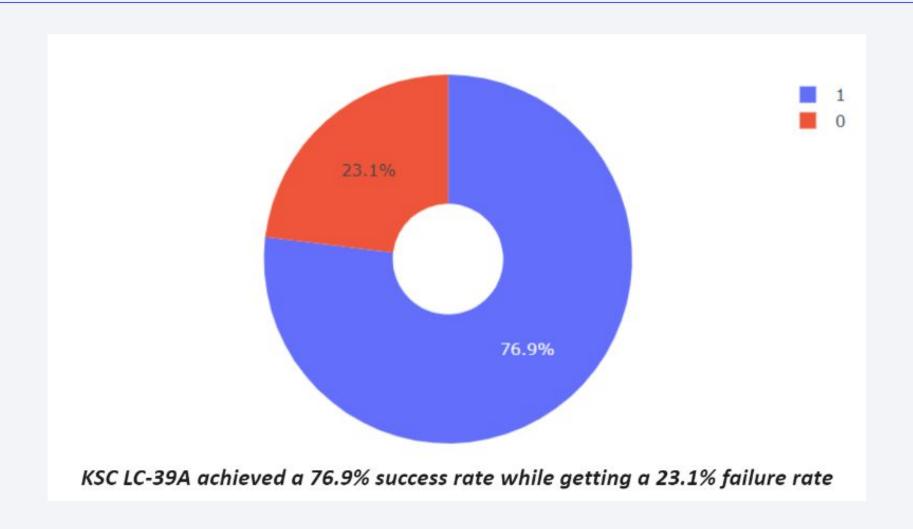




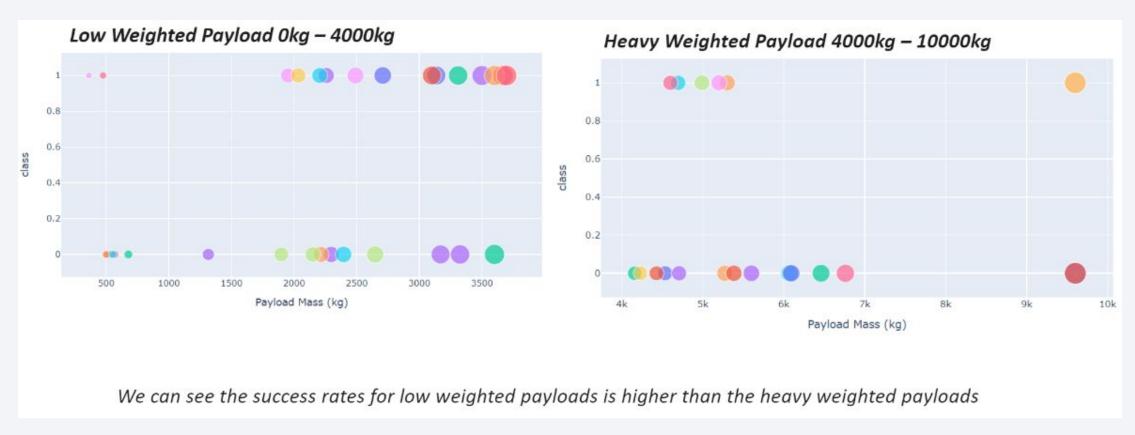
Pie chart showing the success percentage achieved by each launch site



Pie chart showing the Launch site with the highest launch success ratio



Scatter plot of Payload vs Launch Outcome for all sites, with different payload selected in the range slider





Classification Accuracy

KNN Accuracy: 0.8333

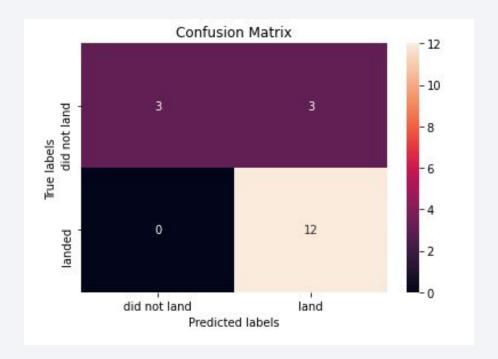
Best performing model is: Logistic Regression with accuracy 0.8333

 The Logistic Regression classifier is the model with the highest classification accuracy

```
Find the method performs best:
41]: accuracy lr = logreg cv.score(X test, Y test)
     accuracy_svm = svm_cv.score(X_test, Y_test)
     accuracy tree = tree cv.score(X test, Y test)
     accuracy knn = knn cv.score(X test, Y test)
     print(f"Logistic Regression Accuracy: {accuracy lr:.4f}")
     print(f"SVM Accuracy: {accuracy svm:.4f}")
     print(f"Decision Tree Accuracy: {accuracy_tree:.4f}")
     print(f"KNN Accuracy: {accuracy_knn:.4f}")
     best accuracy = max(accuracy 1r, accuracy svm, accuracy tree, accuracy knn)
     if best_accuracy == accuracy_lr:
         best model = "Logistic Regression"
     elif best accuracy == accuracy svm:
         best model = "SVM"
     elif best_accuracy == accuracy_tree:
         best model = "Decision Tree"
         best model = "KNN"
     print(f"\nBest performing model is: {best model} with accuracy {best accuracy:.4f}")
     Logistic Regression Accuracy: 0.8333
     SVM Accuracy: 0.8333
     Decision Tree Accuracy: 0.7778
```

Confusion Matrix

- The confusion matrix for the decision tree classifier shows distinguish between different classes.
- A problem is the false positives, unsuccessful landing marked as successful landing by the classifier.



Conclusions

- The larger the flight amount at a launch site, greater is the success rate
- Orbits ES-L1, GEO, HEO, SSO, VLEO had the most success rate.
- KSC LC-39A had the most successful launches of any sites.
- Success rate started to increase in 2013 till 2020.
- The Decision tree classifier is the best tool for this task.

