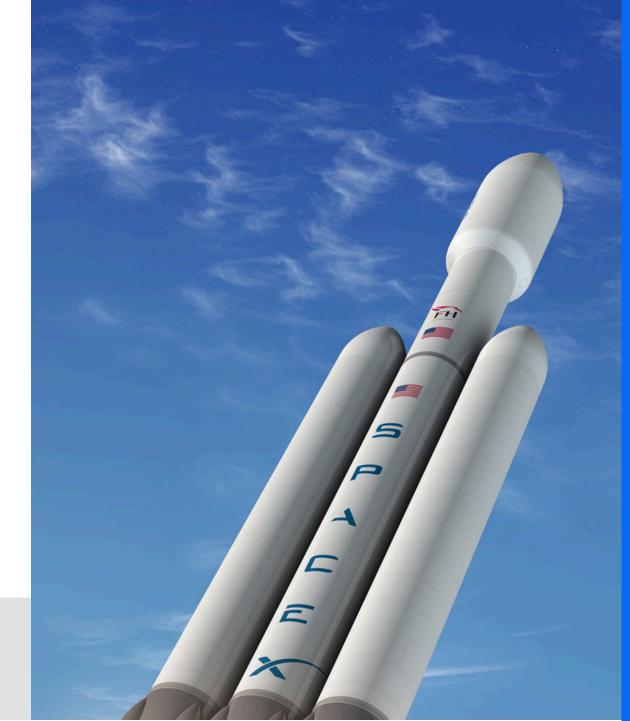




**Space Exploration Technologies** 

## DATA SCIENCE CAPSTONE PROJECT



By Truc Ho 15.5.2025



**TOPIC HIGHLIGHTS** 



- 1. Executive Summary
- 2. Introduction
- 3. Methodology
- 4. Results
- 5. Conclusion

### 1. EXECUTIVE SUMMARY

#### SUMMARY OF METHODDOLOGIES

- Data collection
- Data wrangling
- Exploratory Data Analysis with SQL
- Exploratory Data Analysis with Data Visualization
- Building an interactive map with Folium
- Building a Dashboard with Plotly Dash
- Predictive analysis (Classification)



#### **SUMMARY OF RESULTS**

- Exploratory Data Analysis results
- Interactive analytics
- Predictive analysis results



## 2. INTRODUCTION

#### BACKGROUND AND PROBLEM STATEMENT

The rise of commercial space companies has made space travel more accessible. **SpaceX** leads this effort by significantly cutting launch costs, thanks to its ability to **reuse the first stage** of the Falcon 9 rocket. While a Falcon 9 launch costs about **62 million USD**, competitors may charge over **165 million USD** due to non-reusability.

Predicting whether the **first stage will land successfully** is key to estimating launch costs. In this project, acting as data scientists for a new company, **SpaceY**, we will use **public data and machine learning** to forecast first stage recovery outcomes.







### 2. INTRODUCTION

#### ? KEY QUESTIONS TO EXPLORE

- How do factors like payload, launch site, orbit, and flight number impact landing success?
- Has the success rate improved over time?
- What is the best algorithm for binary classification in this scenario?



- Data collection: using SpaceX Rest API and Web Scrapping from Wikipedia
- Data Wrangling: filtering data, handle missing values, using One Hot Encoding
- Exploratory Data Analysis: using visualization and SQL
- Interactive visual analytics: using Folium, Plotly Dash
  - Built and optimized classification models for predictive analysis.

Data collection process involved a combination of API requests from SpaceX REST API and Web Scraping data from table in SpaceX's Wikipedia entry.

#### **SPACEX REST API**

#### **Data Columns are obtained:**

FlightNumber, Date, BoosterVersion, PayloadMass, Orbit, LaunchSite, Outcome, Flights, GridFins, Reused, Legs, LandingPad, Block, ReusedCount, Serial, Longitude, Latitude

#### **WIKIPEDIA WEB SCRAPING**

#### **Data Columns are obtained:**

Flight No., Launch site, Payload, PayloadMass, Orbit, Customer, Launch outcome, Version Booster, Booster landing, Date, Time



#### **SPACEX REST API**

#### **DATA COLLECTION**

#### **Send API Request**

Retrieve rocket
launch data from
SpaceX API

## Convert JSON to DataFrame

Use .json() and json\_normalize().

## Extract Required Information

Apply custom functions to get specific data.

#### **Build a Dictionary**

Organize the extracted data into a dictionary.

#### Create a DataFrame

Convert the dictionary into a DataFrame.

## Filter for Falcon 9 Launches

Keep only Falcon 9 launch records.

## Handle Missing Values

Replace missing values with the column mean

#### **Export to CSV**

Save the cleaned data to a CSV file.

#### **DATA COLLECTION**

Send Request
Retrieve rocket
launch data from

WIKIPEDIA

**->** 

Creating a
BeautifulSoup
object

Extract Required
Information
from the HTML table
header

Collecting data by parsing HTML tables

WIKIPEDIA WEB SCRAPING **Export to CSV** 

Save the cleaned data to a CSV file.

Cor dictio

**Create a DataFrame** 

Convert the dictionary into a DataFrame.

**Build a Dictionary** 

Organize the extracted data into a dictionary.

#### **Understand Landing Outcomes**

Outcomes like True/False Ocean, True/False RTLS, and True/False ASDS indicate whether the booster landed successfully or failed (in ocean, on land, or on a drone ship).

#### **Label the Data**

Convert outcomes into training labels:

- "1" = successful landing
- "0" = unsuccessful landing

#### **Data Processing Steps**

- Perform EDA and define labels
- Count launches per site
- Count and analyze orbits
- Analyze outcomes by orbit type
- Create outcome labels from the original data
- Export the cleaned dataset to CSV

#### **Charts created:**

- Flight Number vs. Payload Mass / Launch Site / Orbit Type
- Payload Mass vs. Launch Site / Orbit Type
- Orbit Type vs. Success Rate
- Success Rate Trend by Year

#### **Chart types & purposes:**

- Scatter Plots → Show relationships between numerical variables (used for model input analysis)
- Bar Charts → Compare categories (e.g., launch sites or orbit types)
- Line Charts → Show trends over time (e.g., yearly success rate)

#### **EDA WITH SQL**

#### **Key SQL Queries Performed:**

#### **Launch Sites:**

- List unique launch sites
- Filter launch sites starting with 'CCA'

#### **Payload Analysis:**

- Total payload mass by NASA launches
- Average payload mass for booster version F9 v1.1
- Boosters on drone ships with payload between 4000–6000
- Booster versions with maximum payload mass

#### **Landing Outcomes:**

- First successful ground pad landing date
- Count of successful vs. failed missions
- Failed drone ship landings in 2015 with booster & site details
- Ranked outcomes (Success/Failure) between 2010-06-04 and 2017-03-20

## BUILD AN INTERACTIVE MAP WITH FOLIUM

#### 1. Launch Site Markers:

Placed circular markers with popup and text labels using latitude/longitude for:

- NASA Johnson Space Center (start location)
- All launch sites (to visualize geographic spread and proximity to coasts/equator)

#### 2. Launch Outcome Markers:

- Used green markers for successful and red markers for failed launches
- Applied Marker Cluster to group markers and identify high-success launch sites

#### 3. Distance Visualization:

Drew colored lines to show distances from a launch site (e.g., KSC LC-39A) to nearby:

Railways, Highways, Coastlines, and Closest Cities

## BUILD A DASHBOARD WITH PLOTLY DASH

#### 1. Launch Sites Dropdown List:

Enables users to filter data by selecting specific launch sites.

#### 2. Pie Chart (Launch Success Visualization):

 Shows total success vs. failure launches for: All launch sites, a selected launch site (if chosen from dropdown)

#### 3. Payload Mass Range Slider:

Allows selection of a payload mass range to filter data dynamically.

#### 4. Scatter Chart (Payload vs. Success Rate by Booster Version):

 Visualizes correlation between payload mass and launch success, broken down by booster version.

#### 1. Prepare the Data

 Extract the Class column into a NumPy array (used as the target variable).

### 2. Preprocess the Features

Standardize the features using StandardScaler to normalize the data.

### 3. Split the Dataset

Use train\_test\_split to divide the data into training and testing sets.

### 4. Model Selection & Hyperparameter Tuning

 Create a GridSearchCV object with cross-validation (cv=10) to find the best model parameters.

#### **PREDICTIVE ANALYSIS**

### 3. METHODOLOGY

#### 5. Model Training

Apply GridSearchCV on:

- Logistic Regression (LogReg)
- Support Vector Machine (SVM)
- Decision Tree
- K-Nearest Neighbors (KNN)

#### 6. Model Evaluation

Evaluate model performance by:

- Computing accuracy on test data using .score()
- Analyzing the confusion matrix
- Comparing Jaccard Score and F1 Score

#### **Understand Landing Outcomes**

Outcomes like True/False Ocean, True/False RTLS, and True/False ASDS indicate whether the booster landed successfully or failed (in ocean, on land, or on a drone ship).

#### **Label the Data**

Convert outcomes into training labels:

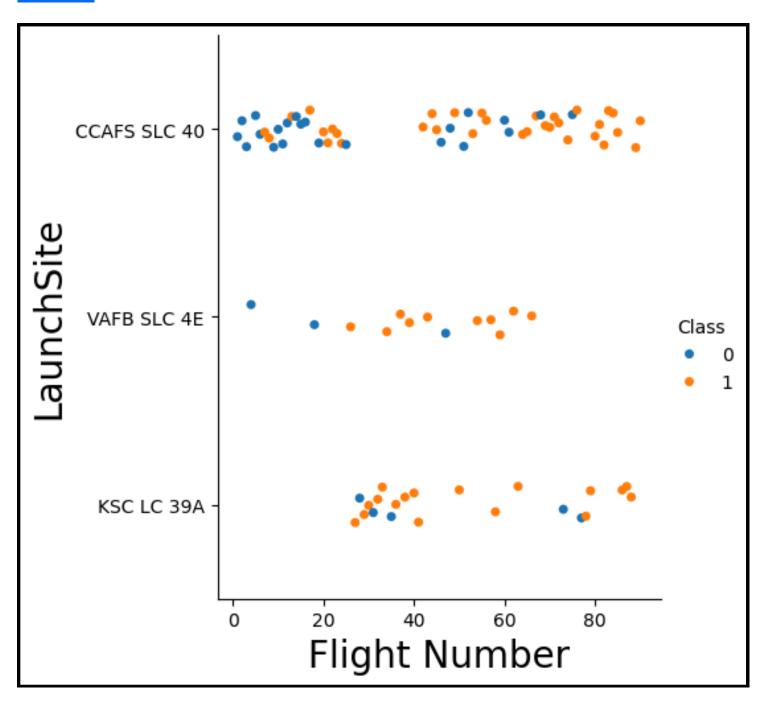
- "1" = successful landing
- "0" = unsuccessful landing

#### **Data Processing Steps**

- Perform EDA and define labels
- Count launches per site
- Count and analyze orbits
- Analyze outcomes by orbit type
- Create outcome labels from the original data
- Export the cleaned dataset to CSV

## EDA WITH VISUALIZE

## 4. RESULTS

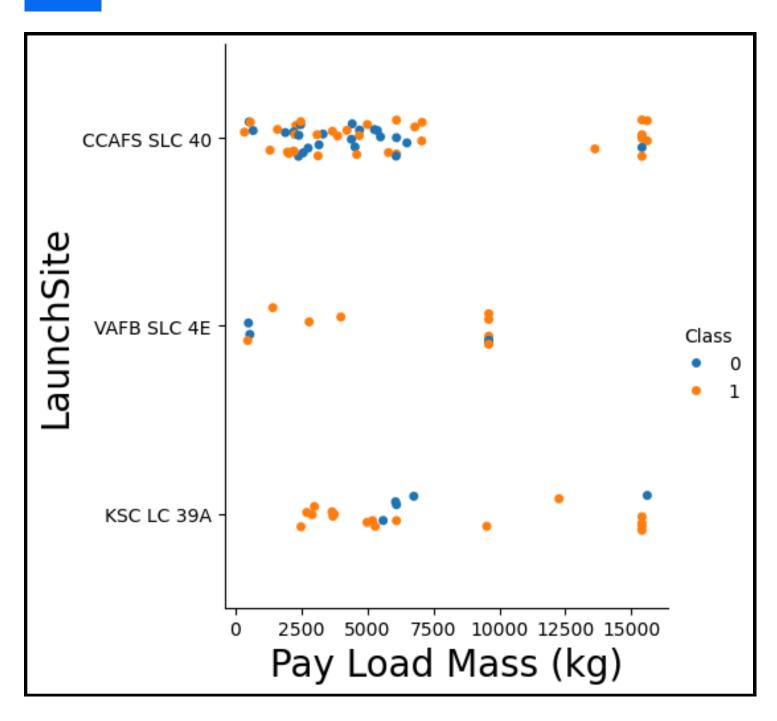


FLIGHT NUMBER AND LAUNCH SITE

From the scatter plot, we observe that:

- Most launches from CCAFS SLC 40 and KSC LC 39A resulted in successful missions (Class 1), especially as the flight number increases, indicating improved reliability over time.
- Launches from VAFB SLC 4E show a mix of successes and failures, with fewer overall flights compared to the other sites.
- There is a noticeable trend where higher flight numbers tend to be associated with more successes, suggesting learning or technological improvements over time.

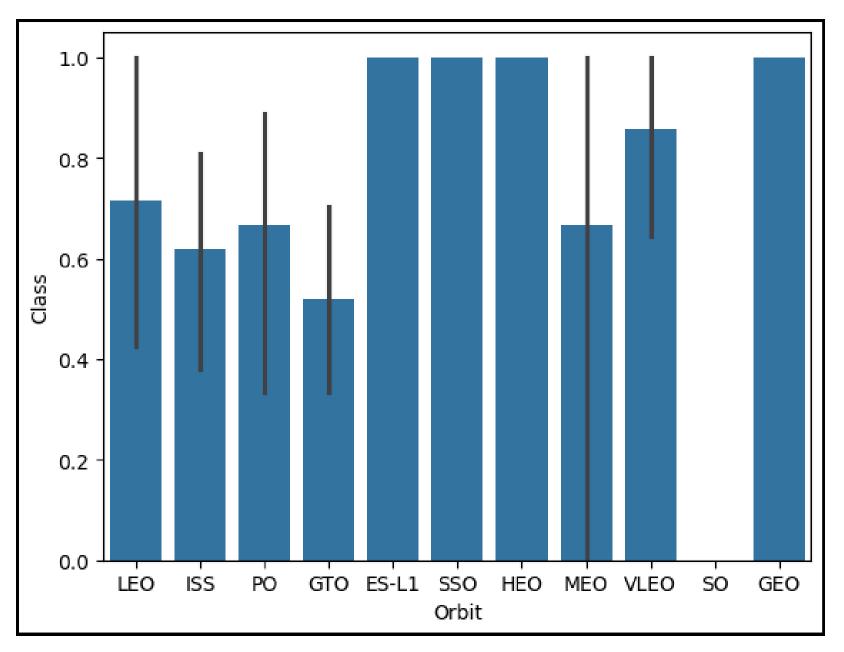
# EDA WITH VISUALIZE



The plot shows that CCAFS SLC 40 had the most launches with high success rates. KSC LC 39A handled the heaviest payloads, mostly with successful outcomes. VAFB SLC 4E had fewer launches with mixed results. Heavier payloads were generally launched successfully.

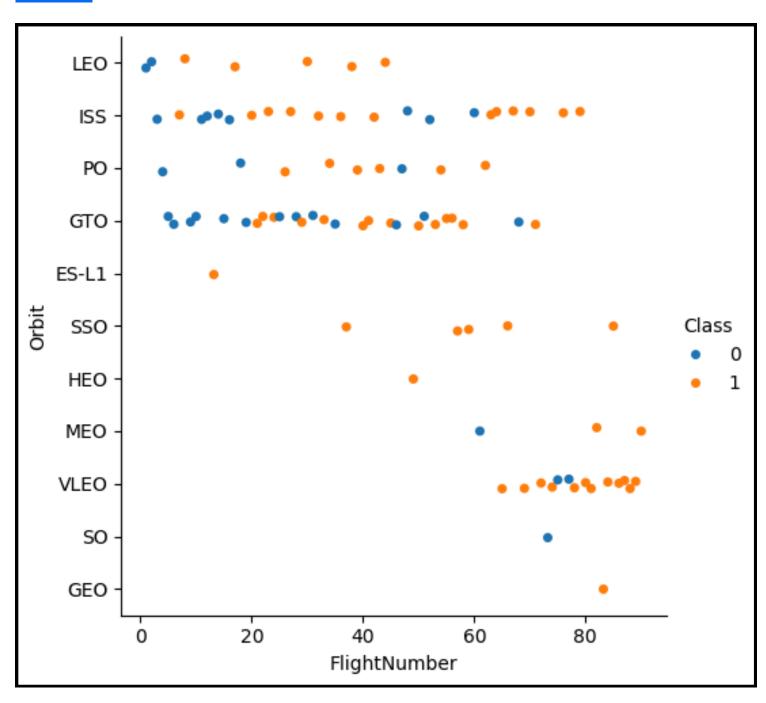
FLIGHT NUMBER AND PAY LOAD MAX

## EDA WITH VISUALIZE



The chart shows that SSO, ES-L1, HEO, and GEO orbits had the highest success rates (100%), while GTO had the lowest. Most orbit types had high success, but several had large error bars, indicating variability in outcomes.

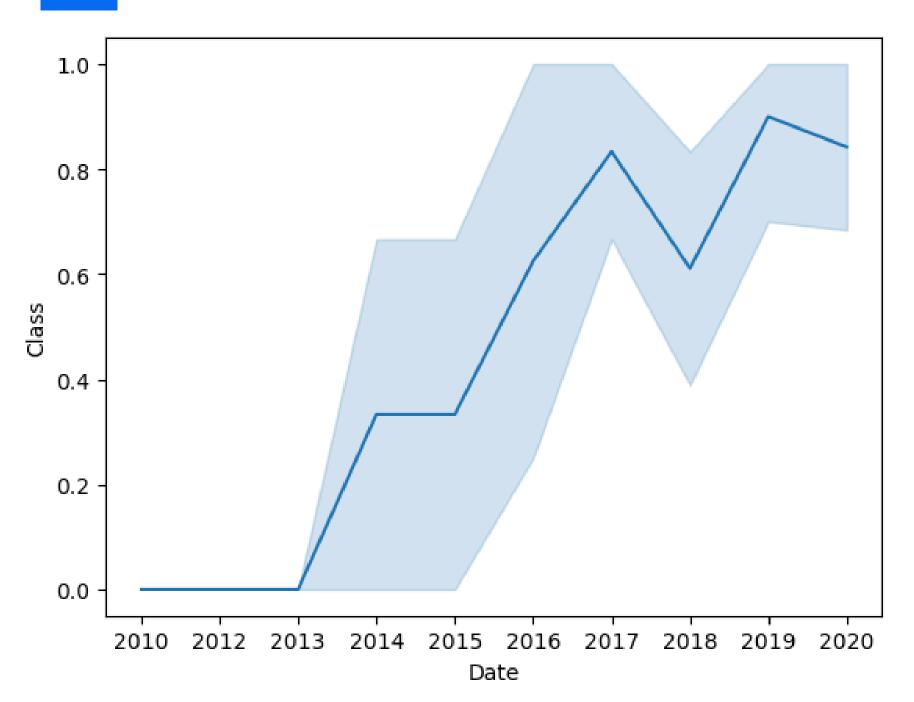
SUCCESS RATE OF EACH ORBIT TYPE



Most orbits were used throughout the flight history. High success (orange) is observed for SSO, ES-L1, and GEO. Failures (blue) are more common in early flights and for orbits like LEO and GTO.

FLIGHTNUMBER AND ORBIT TYPE

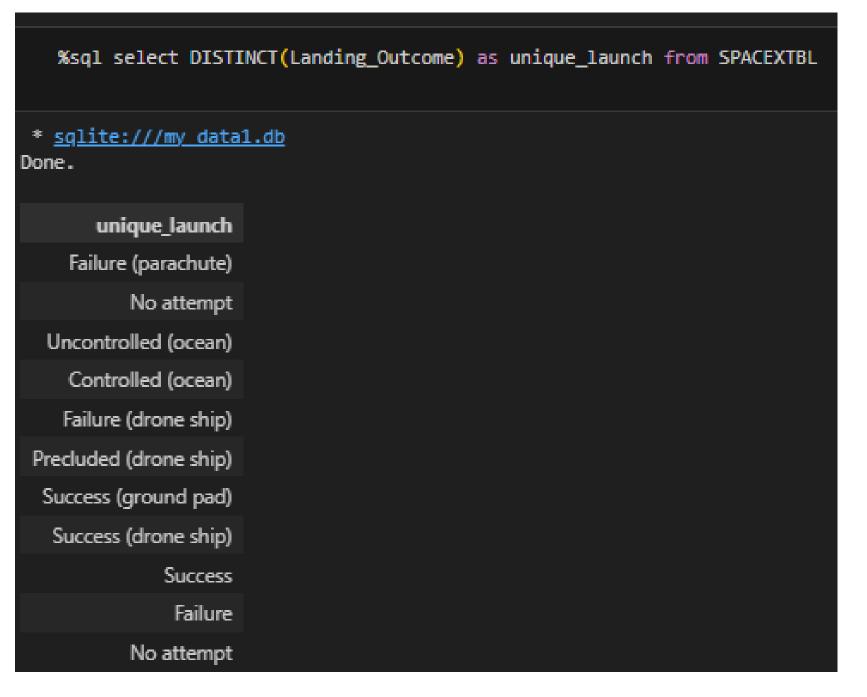
# EDA WITH VISUALIZE



The graph shows a rising success trend from 2010 to 2020, with a sharp increase starting in 2013 and peaking around 2019. Fluctuations indicate variability, but the overall trajectory is upward, suggesting significant progress over the years

#### **SUCCESS YEARLY TREND**

### **EDA WITH SQL**



**ALL LAUNCH SITE NAMES** 

## 4. RESULTS

%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	Landing_Outcome
2010-06-04	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
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	Failure (parachute)
2012-05-22	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
2012-10-08	0:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
2013-03-01	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

DISPLAY 5 RECORDS WHERE LAUNCH SITES BEGIN WITH THE STRING 'CCA'

```
%sql select SUM(PAYLOAD_MASS__KG_) from SPACEXTBL where Customer = 'NASA (CRS)'

* sqlite://my data1.db
Done.

SUM(PAYLOAD_MASS__KG_)

45596
```

DISPLAY THE TOTAL PAYLOAD MASS CARRIED BY BOOSTERS LAUNCHED BY NASA (CRS)

```
%sql select SUM(PAYLOAD_MASS__KG_) from SPACEXTBL where Customer = 'NASA (CRS)'

* sqlite://my data1.db
Done.

SUM(PAYLOAD_MASS__KG_)

45596
```

## DISPLAY THE TOTAL PAYLOAD MASS CARRIED BY BOOSTERS LAUNCHED BY NASA (CRS)

```
% %sql select AVG(PAYLOAD_MASS__KG_) as `average payload mass` from SPACEXTBL where Booster_Version LIKE 'F9 v1.1%'

* sqlite://my data1.db
Done.

average payload mass
    2534.666666666665
```

DISPLAY AVERAGE PAYLOAD MASS CARRIED BY BOOSTER VERSION F9 V1.1

## 4. RESULTS

```
%sql select MIN(Date) as `first successful landing outcome in ground pad was acheived` from SPACEXTBL where Landing_Outcome == 'Success (ground pad)'

* sqlite:///my_datal.db
Done.

first successful landing outcome in ground pad was acheived

2015-12-22
```

THE FIRST SUCCESFUL LANDING OUTCOME IN GROUND PAD WAS ACHEIVED.

## 4. RESULTS

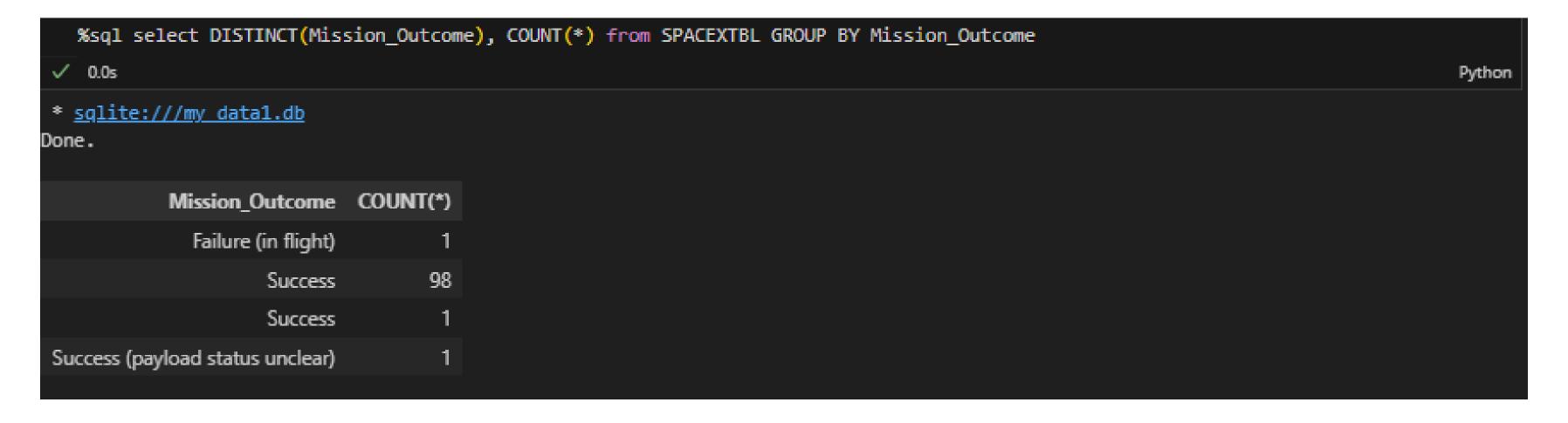
```
%sql select Booster_Version from SPACEXTBL where Landing_Outcome = 'Precluded (drone ship)' and 4000
Python

* sqlite://my data1.db
Done.

Booster_Version
F9 v1.1 B1018
```

## THE BOOSTERS WHICH HAVE SUCCESS IN DRONE SHIP AND HAVE PAYLOAD MASS GREATER THAN 4000 BUT LESS THAN 6000

## 4. RESULTS



TOTAL NUMBER OF SUCCESSFUL AND FAILURE MISSION OUTCOMES

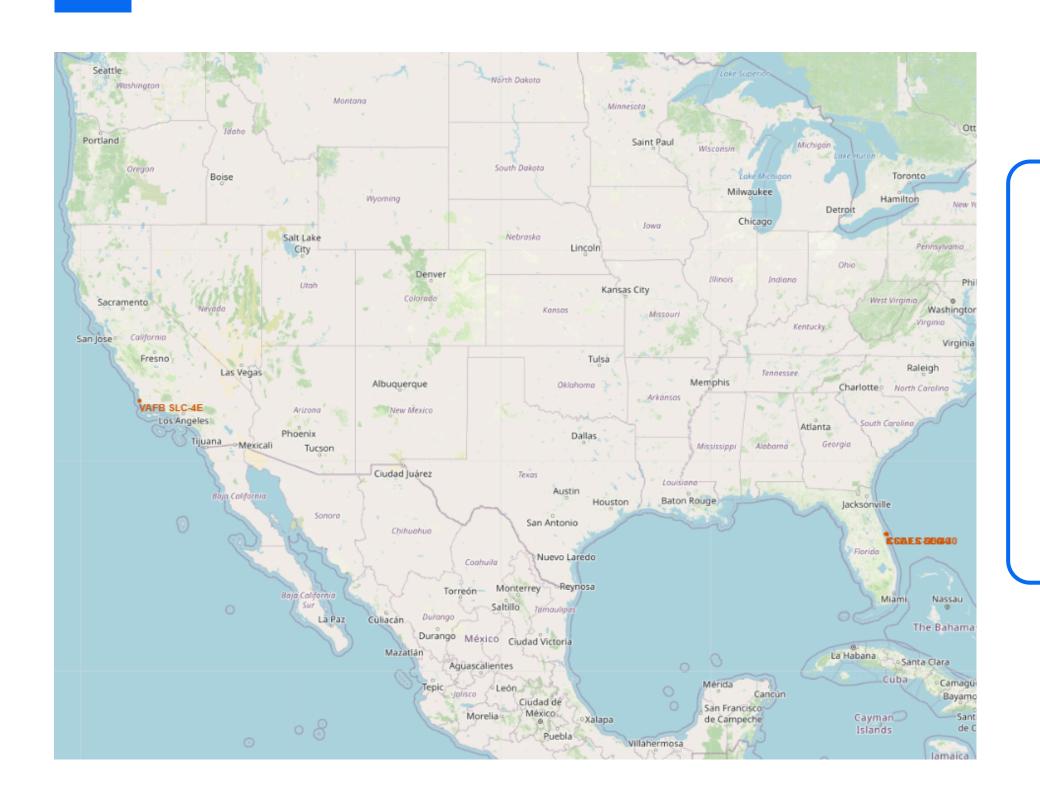
#### **EDA WITH SQL**

```
%sql SELECT Booster_Version FROM SPACEXTBL WHERE PAYLOAD_MASS__KG_ = (SELECT MAX(PAYLOAD_MASS__KG_) FROM SPACEXTBL);
 ✓ 0.0s
* sqlite:///my data1.db
Done.
 Booster_Version
   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
```

ALL THE BOOSTER\_VERSIONS THAT HAVE CARRIED THE MAXIMUM PAYLOAD MASS. USE A SUBQUERY.

THE RECORDS WHICH WILL DISPLAY THE MONTH NAMES, FAILURE LANDING\_OUTCOMES IN DRONE SHIP ,BOOSTER VERSIONS, LAUNCH\_SITE FOR THE MONTHS IN YEAR 2015.

## INTERACTIVE MAP WITH FOLIUM



Most launch sites are near the equator because the Earth's rotation gives rockets an extra speed boost of 1670 km/h due to inertia, helping them achieve orbit more efficiently. Additionally, launch sites are usually close to the coast to reduce the risk of debris falling in populated areas.



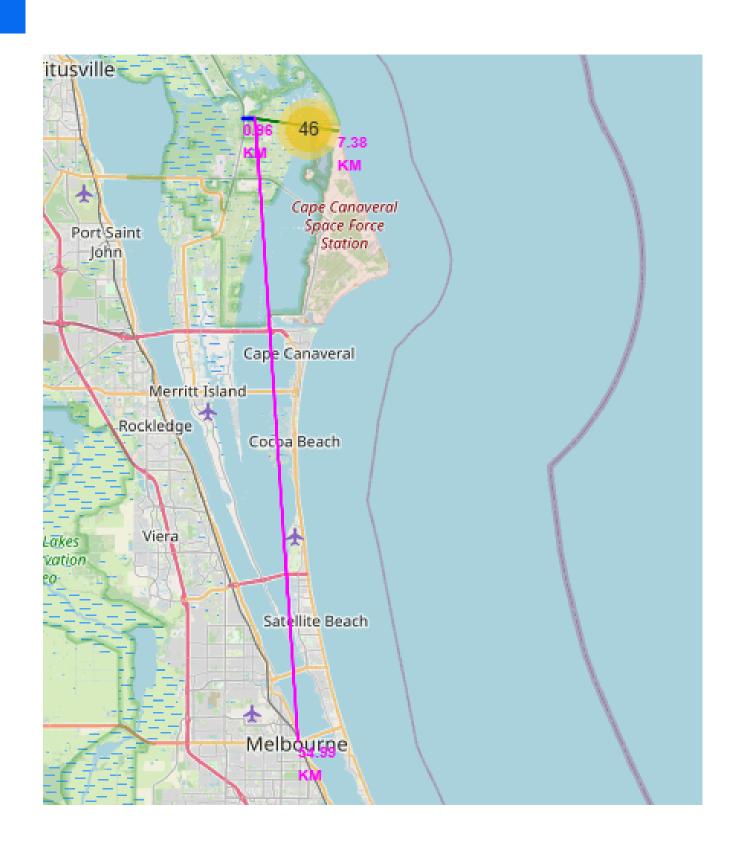
## INTERACTIVE MAP WITH FOLIUM



From the colour-labeled markers we should be able to easily identify which launch sites have relatively high success rates.

- Green Marker = Successful Launch
- Red Marker = Failed Launch
- Launch Site KSC LC-39A has a very high Success Rate.

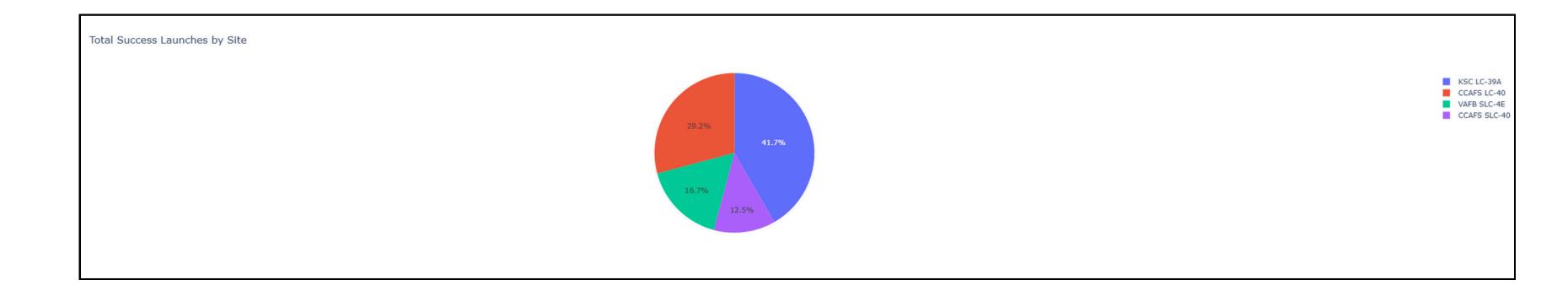
## INTERACTIVE MAP WITH FOLIUM



The launch site, marked in orange on the map, is located at Cape Canaveral Space Force Station. It is positioned northeast of Melbourne, approximately 46 km away. The site is also near the coastline, with a distance of about 7.38 km offshore indicated on the map. This strategic location allows launches to occur over the ocean, minimizing risk to populated areas.

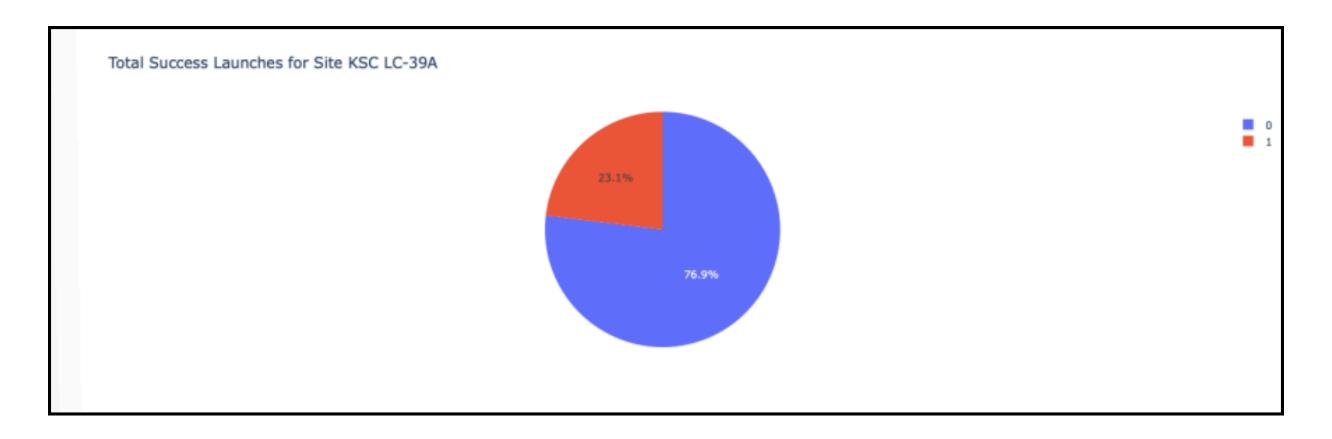


## DASHBOARD WITH PLOTY DASH



The chart clearly shows that from all the sites, KSC LC-39A has the most successful launches.

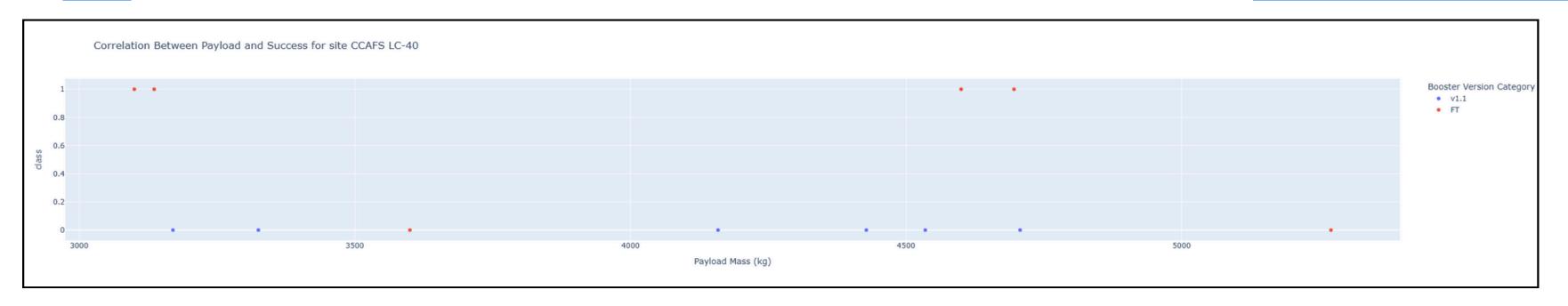
## DASHBOARD WITH PLOTY DASH

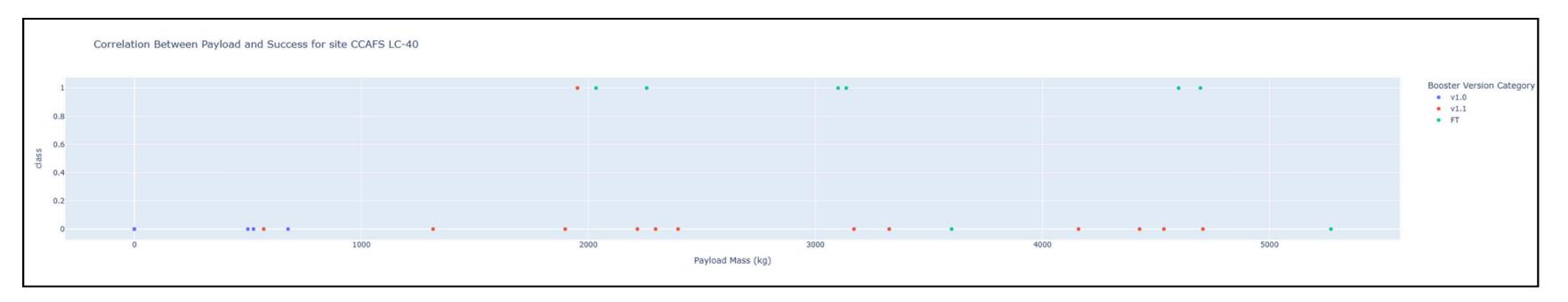


KSC LC-39A has the highest launch success rate (76.9%) with 10 successful and only 3 failed landings



# DASHBOARD WITH PLOTY DASH







## PREDICTIVE ANALYSIS

	LogReg	SVM	Tree	KNN
Jaccard_Score	0.800000	0.800000	0.800000	0.800000
F1_Score	0.888889	0.888889	0.888889	0.888889
Accuracy	0.833333	0.833333	0.833333	0.833333

#### SCORES AND ACCURACY OF THE TEST SET

	LogReg	SVM	Tree	KNN
Jaccard_Score	0.833333	0.845070	0.882353	0.819444
F1_Score	0.909091	0.916031	0.937500	0.900763
Accuracy	0.866667	0.877778	0.911111	0.855556

- The tables compare four machine learning models—Logistic Regression, SVM, Decision Tree, and KNN—based on Jaccard Score, F1 Score, and Accuracy.
- In test results, all models achieve 83.33% accuracy. However, Decision Tree performs best on the full dataset with 91.11% accuracy, followed by SVM (87.78%), Logistic Regression (86.67%), and KNN (85.56%).
- Decision Tree shows the highest overall performance, but consistency across different datasets should be considered before selecting the best model.

SCORES OF PREDICTION ON WHOLE DATASET

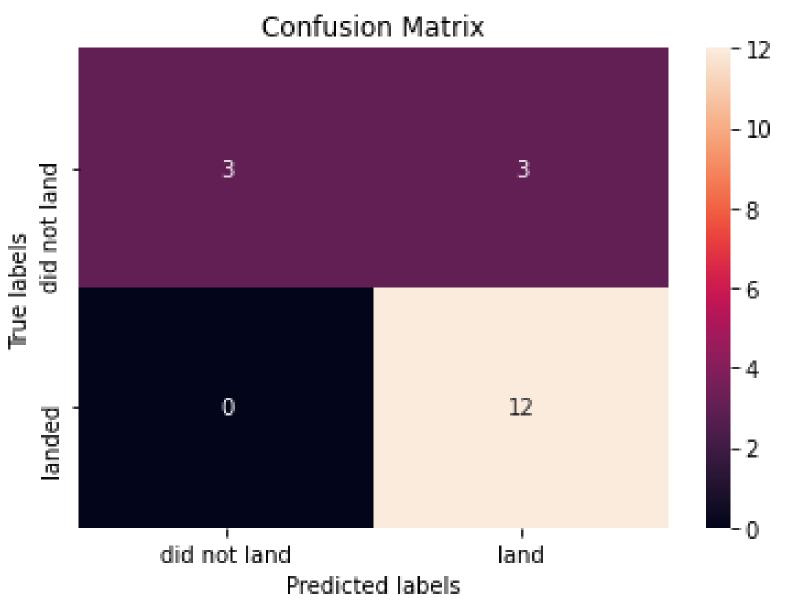


#### **CONFUTION MATRIX**

#### **Confusion Matrix**

			Ground Truth Label			
	Total Oh	servations	has disease	no disease		
	Total Ot	(n)	Condition Positive (CP)	Condition Negative (CN)		
d Label	test positive	Test Outcome Positive (TOP)	True Positive (TP)	False Positive (FP)		
Predicted	test negative	Test Outcome Negative (TON)	False Negative (FN)	True Negative (TN)		

Figure 1: Basic colour coded confusion matrix with marginal sums



**Confution matrix of decision tree** 



### 5. CONCLUSION

- The Decision Tree model demonstrates the highest performance for this dataset.
- Launches with lighter payloads tend to yield better results compared to those with heavier payloads. Most launch sites are situated near the Equator and are positioned very close to the coastline, likely optimizing launch conditions.
- Over time, the overall success rate of launches has improved. Among all sites, KSC LC-39A has the highest success rate. Additionally, certain orbital destinations—ES-L1, GEO, HEO, and SSO—achieve a perfect 100% success rate.