

# Winning Space Race with Data Science

Rafael Jeronymo August 14, 2025



#### **Outline**

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

#### **Executive Summary**

- This project explores SpaceX Falcon 9 launch data
- The goal is to understand what factors affect successful landings and to build simple models to predict outcome
- The methodology I used was the collection of data, cleaning it, analyzing it with SQL and visualizations, creating dashboards and testing machine learning

#### Introduction

- The case study of SpaceX launches is relevant because it brings some questions that are important for both the company and organizations interested in this data
- The main question would be: Can we predict if a Falcon 9 first stage will land successfully?
- This is important because successful landings lower costs and make space more accessible.



### Methodology

#### **Executive Summary**

- Data collection methodology:
  - Data was collected from SpaceX API and web scraping from Wikipedia
- Perform data wrangling
  - Data was cleaned for missing values and formatted dates
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Predictive analysis through Logistic Regression, Decision Trees and KNN

#### **Data Collection**

- Data was collected using the SpaceX API to pull launch records and web scraping from a Wikipedia page titled "List of Falcon 9 and Falcon Heavy launches"
- Data collection process using API:
  - Request rocket launch data from SpaceX API
  - Parse the data using the GET request
  - Decode the response content as Json and turn it into a Pandas dataframe
- Data collection process using Web Scraping:
  - Request the wiki page from its URL as an HTTP response
  - Create a BeautifulSoup object from the HTML response
  - Extract all column/variables from the HTML
  - Create a data frame by parsing the launch HTML tables

### **Data Wrangling**

- •Data was processed for missing values and replaced with the mean
- Data wrangling process
  - Calculate the mean for the PayloadMass
  - Use the mean to replace the missing values with the mean calculated
  - Create a landing outcome label from Outcome column
  - Create a list where zero is bad outcome and 1 is otherwise

#### **EDA** with Data Visualization

- Below are the charts used and why they were used
  - Flight Number vs Launch Site: showed launch frequency and success by site
  - PayloadMass vs Launch Site: explored effect of payload size on outcomes
  - Success Rate by Orbit Type: bar chart revealed which orbits had higher success
  - Flight Number / Payload vs Orbit: scatter plots to see trends across missions
  - Launch Success Rate by Year: line chart showing improvement over time

#### **EDA** with SQL

- Display the names of the unique launch sites
- Display 5 records where launch site begins with 'CCA'
- Display the total payload mass carried by boosters launched by NASA
- Display average payload mass carried by booster version F9 v1.1
- List the date when the first successful landing outcome in ground pad was achieved
- List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- List the total number of successful and failure mission outcomes
- List all the booster versions that have carried the maximum payload mass
- Display the month names, failure landing outcomes in drone ship ,booster versions, launch site for the months in year 2015
- Rank the count of landing outcomes between the date 2010-06-04 and 2017-03-20, in descending order

### Build an Interactive Map with Folium

- Mark all launch sites on a map
- Use circles for the success/failed launches for each site on the map
- Draw lines to show the distances between a launch site to its proximities

• These objects helped visualize launch locations, outcomes, and geographical, factors that may affect mission success.

### Build a Dashboard with Plotly Dash

- Pie Chart
  - Showed launch success rates for all sites
- Pie Chart per Site
  - Highlighted site with highest success ratio
- Scatter Plot
  - Explored how payload size affects success
- Dropdown Filter
  - Let users select specific launch sites

Why: these interactions made the dashboard dynamic, helping explore success patterns by site and payload

### Predictive Analysis (Classification)

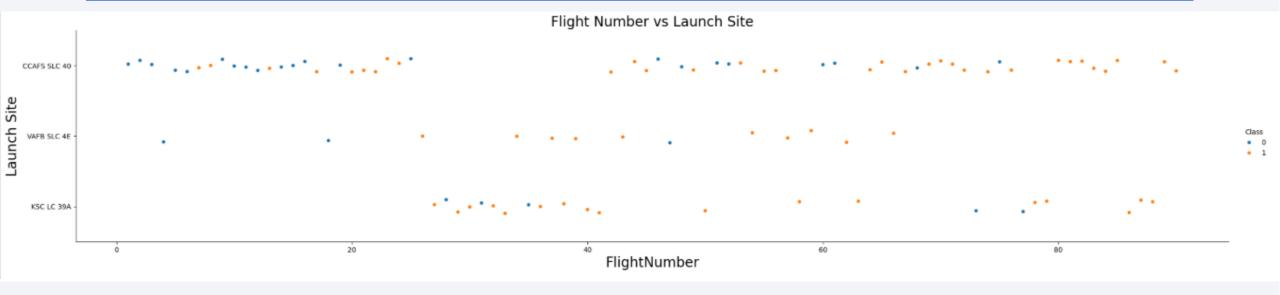
- Models used: Logistic Regression, SVM, Decision Tree, KNN
- Process (Flowchart)
  - Raw Data Feature Selection Train/Test Split Model
     Training Model Evaluation Best Model
- Best model: Decision Tree with training accuracy of 88%

#### Results

- Exploratory data analysis results
  - Success rates varied by orbit type and launch site
  - Payload range (2000-6000kg) showed higher success probability
  - Success rates improved steadily over the years
- Interactive analytics
  - Folium maps revealed site proximity to coasts and infrastructure
  - Dash allowed filtering by payload and site for deeper insights
- Predictive analysis results
  - Logistic Regression, KNN, Decision Tree and SVM
  - Best performing model: Decision Tree

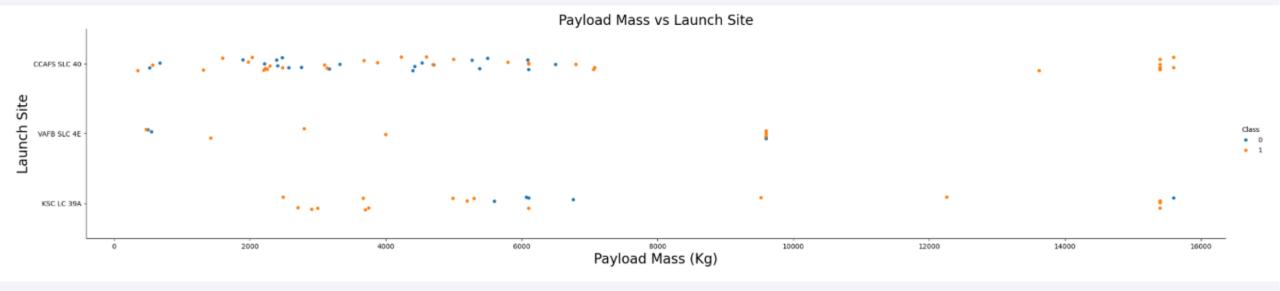


### Flight Number vs. Launch Site



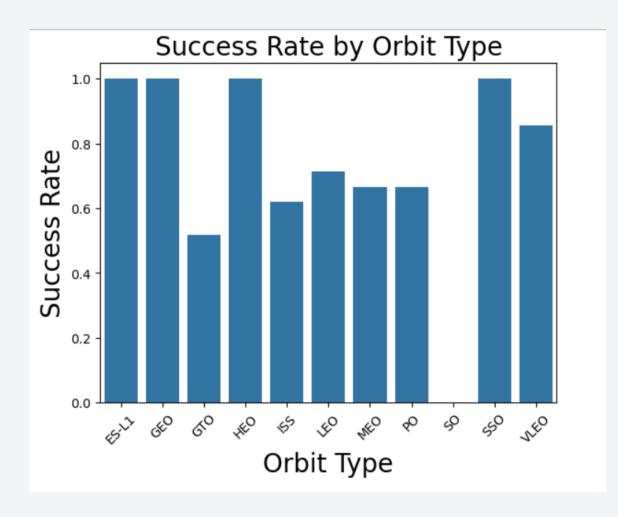
- Scatter plot shows launches grouped by site (CCAFS, VAFB, KSC)
- Early flights had more failures
- Over time, success rate increased at each site
- KSC and CCAFS sites had more frequent launches compared to VAFB

### Payload vs. Launch Site



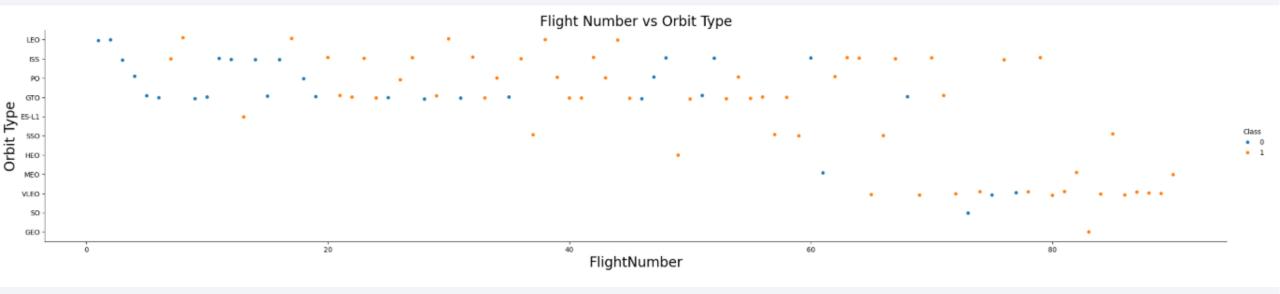
- Scatter plot compares payload size with launch site outcomes
- Most payloads between 2000-6000kg had higher success rates
- Very heavy payloads (> 10,000kg) showed some success but less frequent
- CCAFS and KSC handled the widest range of payloads, while VAFB had fewer launches

### Success Rate vs. Orbit Type



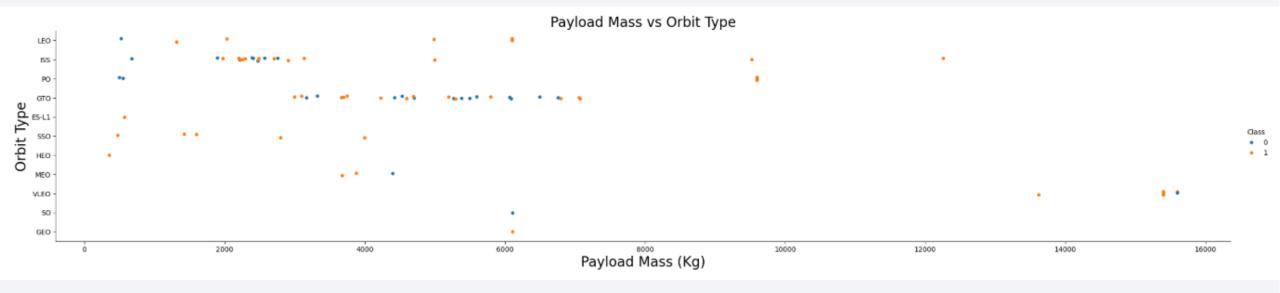
- Some orbits like ES-L1, GEO, HEO and SSO had almost 100% success rates
- Others like GTO and ISS showed lower reliability
- It shows how orbit type influences landing success probability

### Flight Number vs. Orbit Type



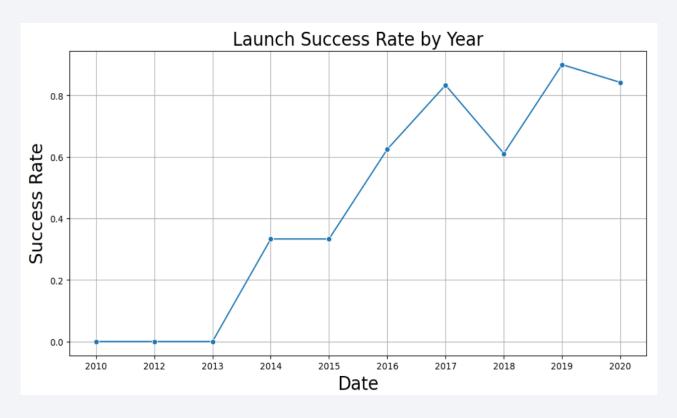
- Scatter plot shows missions grouped by orbit type across flight history
- Early missions had more failures across all orbits
- Over time, success rates improved across most orbit types
- GTO and ISS missions stood out with mixed results, while LEO and SSO showed more consistent successes later

### Payload vs. Orbit Type



- Scatter plot shows how payload mass relates to orbit type and landing outcome
- Medium payloads (2000-6000 kg) had higher success rates
- LEO and SSO orbits supported a wide range of payloads and showed more consistent successes

### Launch Success Yearly Trend



- Early launches (2010-2013) had no successful landings
- Success rates improved significantly starting in 2014
- By 2017-2019, rates reached above 80-90%, showing strong reliability improvements
- This trend shows how fast SpaceX progressed in landing technology

#### All Launch Site Names

#### Launch\_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

 This is the result of the query showing the four distinct SpaceX launch sites

### Launch Site Names Begin with 'CCA'

| Date           | Time<br>(UTC) | Booster_Version | Launch_Site     | Payload   |
|----------------|---------------|-----------------|-----------------|---|
| 2010-<br>06-04 | 18:45:00      | F9 v1.0 B0003   | CCAFS LC-<br>40 | Dragon Spacecraft Qualification<br>Unit                             |
| 2010-<br>12-08 | 15:43:00      | F9 v1.0 B0004   | CCAFS LC-<br>40 | Dragon demo flight C1, two<br>CubeSats, barrel of Brouere<br>cheese |
| 2012-<br>05-22 | 7:44:00       | F9 v1.0 B0005   | CCAFS LC-<br>40 | Dragon demo flight C2   |
| 2012-<br>10-08 | 0:35:00       | F9 v1.0 B0006   | CCAFS LC-<br>40 | SpaceX CRS-1  |
| 2013-<br>03-01 | 15:10:00      | F9 v1.0 B0007   | CCAFS LC-<br>40 | SpaceX CRS-2  |

• This is the result of the query showing 5 records where launch sites begin with 'CCA'

### **Total Payload Mass**

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

```
%%sql
SELECT SUM("Payload_Mass_Kg_") AS Total_Payload_Mass
FROM SPACEXTABLE
WHERE "Customer" = 'NASA (CRS)'

* sqlite:///my_data1.db
Done.

Total_Payload_Mass

0.0
```

• In this dataset, no payload mass values were recorded for NASA CRS launches.

### 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 Average_Payload

FROM SPACEXTABLE

WHERE "Booster_Version" = 'F9 v1.1'

* sqlite://my_data1.db

Done.

Average_Payload

0.0
```

The dataset returned no payload mass for this booster version.

### 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) AS First_Successful_Ground_Landing
FROM SPACEXTABLE

WHERE "Landing_Outcome" = 'Success (ground pad)'

* sqlite:///my_data1.db
Done.

First_Successful_Ground_Landing

2015-12-22
```

• The first successful ground landing date was 2015-12-22.

#### 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 SPACEXTABLE
WHERE "Landing_Outcome" = 'Success (drone ship)'
    AND "Payload_Mass_Kg_" > 4000
    AND "Payload_Mass_Kg_" < 6000

* sqlite:///my_data1.db
Done.
Booster_Version</pre>
```

• The dataset did not contain any launches for this criteria

#### Total Number of Successful and Failure Mission Outcomes

| Mission_Outcome                  | Outcome_Count |
|----------------------------------|---------------|
| Failure (in flight)              | 1             |
| Success                          | 98            |
| Success                          | 1             |
| Success (payload status unclear) | 1             |

• The dataset shows SpaceX achieved the majority of missions successfully, with only 1 failure (in flight).

### **Boosters Carried Maximum Payload**

## F9 v1.0 B0003 F9 v1.0 B0004

• The list of boosters is quite large so the screenshot shows some of them.

F9 v1.0 B0005
F9 v1.0 B0006
F9 v1.0 B0007
F9 v1.1 B1003
F9 v1.1
F9 v1.1
F9 v1.1

F9 v1.1

F9 v1.1

#### 2015 Launch Records

| 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 |

• In 2015, two boosters failed to land successfully on drone ships, and both were launched from the site CCAFS LC-40

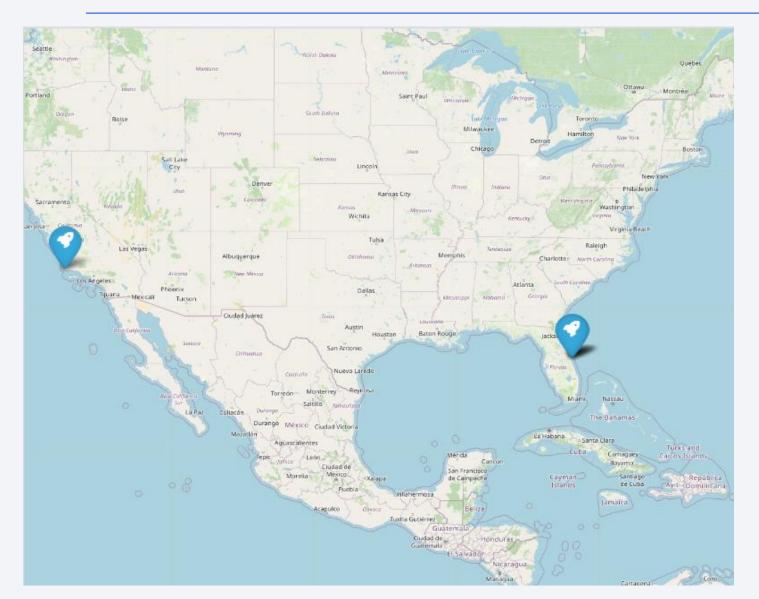
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

| Landing_Outcome      | Outcome_Count |  |
|----------------------|---------------|--|
| Failure (drone ship) | 5             |  |
| Success (ground pad) | 3             |  |

• Between 2010 and 2017, drone ships landings had more failures, while ground pad landings achieved a higher success ratio.

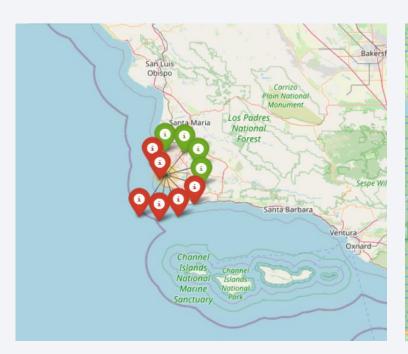


#### **All Launch Sites**

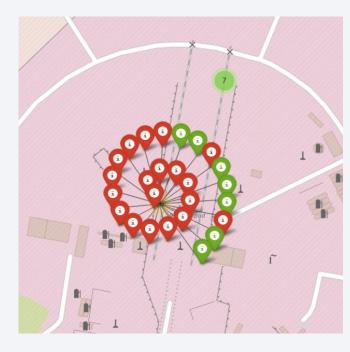


 The map helps visualize SpaceX's geographic strategy in choosing coastal launch sites.

#### **Launch Outcomes**

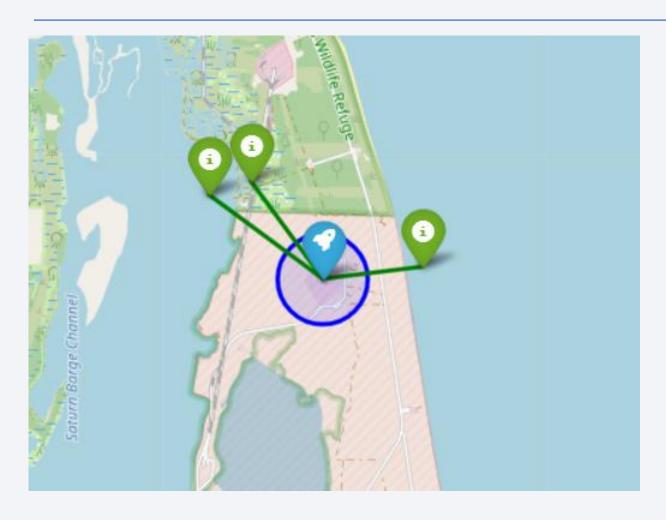






- It shows performance differences across sites
- Green for success and red for failure

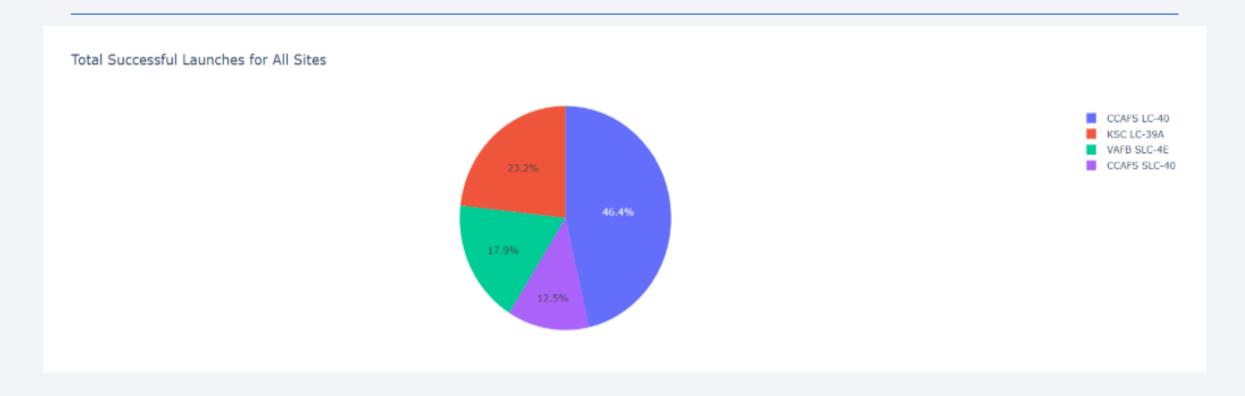
### Site Proximity Analysis



- Distance to nearest Railroad: 1.58Km
- Distance to nearest Coastline: 1.12Km
- Distance to nearest Highway: 1.36Km

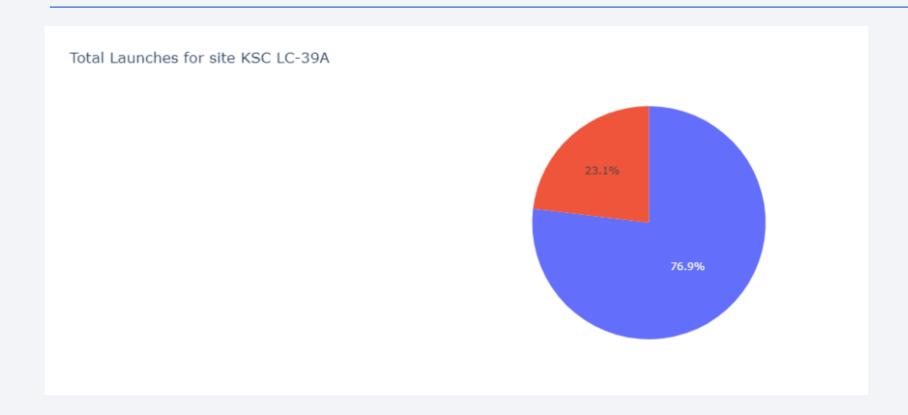


### Launch Success by Site



• The pie charts shows the overall reliability across all sites

### **Highest Performing Site**



• The site KSC LC-39A had the highest rate of success

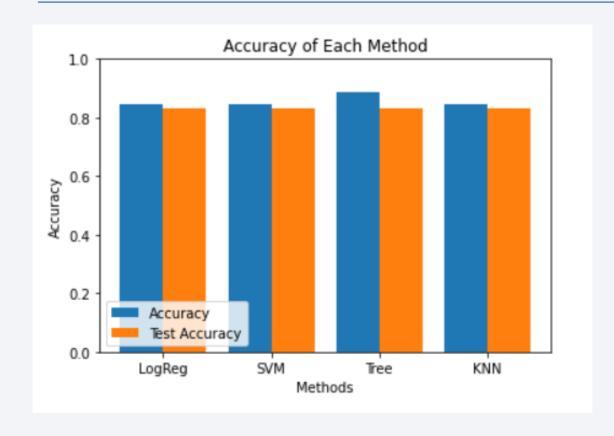
### Payload vs Launch Outcome



 It shows how payload range and site relate to success; reveals ranges with higher success probability

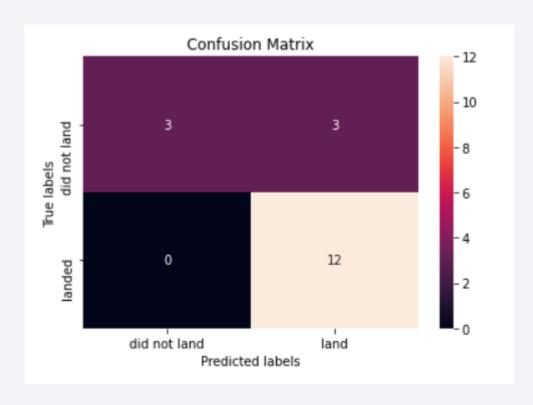


### Classification Accuracy



• The Decision Tree model gave the best prediction performance.

#### **Confusion Matrix**



- True positives: 12 correctly predicted landings
- True negatives: 3 correctly predicted failures
- False positives: 3 predicted landing but failed
- False negatives: O all landings were captured correctly

#### Conclusions

- SpaceX launch success rates have steadily improved over the years
- Success varies by orbit type and payload mass (best results between 2000-6000kg)
- Interactive tools such as Folium and Dash helped visualize launch location, outcomes, and patterns
- Machine learning models predicted landing success, with Decision Tree giving the best accuracy
- Overall, data analysis shows how SpaceX increased reliability and efficiency in booster landings

