



IBM Developer  
SKILLS NETWORK

# Winning Space Race with Data Science

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# Outline

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

# Executive Summary

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## Methodologies

The aim of this study is to identify the determinants of a successful rocket landing. To achieve this objective, the investigation employed the methodologies including:

- Data Collection
- Data Wrangling
- Exploratory Data Analysis (EDA) using SQL, Pandas, Matplotlib
- Interactive Visual Analysis using Analytics using Folium, Plotly, Dash
- Predictive Analysis using Machine Learning

## Summary

The results were being extracted using the following essential component as shown in subsequent slides.

- **Exploratory Data Analysis (EDA)**
- **Interactive Visual Analysis**
- **Predictive Analysis**

# Introduction

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## Project background

**SpaceX**, founded by Elon Musk in 2002, is a California-based aerospace company focused on **reducing space transportation costs and advancing space exploration**. Notable achievements include operating Falcon 9 and Falcon Heavy rockets, Dragon and Starship spacecraft, and creating the largest-ever satellite constellation called Starlink.

SpaceX has pioneered various milestones, such as the first private liquid-propellant rocket to reach orbit, launching, orbiting, and recovering spacecraft, sending a spacecraft to the International Space Station, and achieving vertical propulsive landings and reusability of orbital rocket boosters.

## Problems statement

- In this capstone project, our objective is to **forecast the successful landing of the Falcon 9 first stage**.
- SpaceX promotes its Falcon 9 rocket launches on its official platform, pricing them at 62 million dollars, significantly lower than other providers whose costs can go up to 165 million dollars per launch.
- The **substantial savings SpaceX achieves are largely attributed to the ability to reuse the first stage**.
- By accurately predicting the first stage landing outcome, we can **effectively estimate the overall launch cost**. This insight becomes crucial for other companies considering bidding against SpaceX for a rocket launch contract.



Section 1

# Methodology

# Methodology

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## Executive Summary

- Data collection methodology:
  - Using SpaceX REST API and web scraping techniques
- Perform data wrangling
  - By filtering the data, handling missing values and applying hot encoding
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - To find best model and parameters

# Data Collection

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This information may prove valuable if an alternate company intends to bid against SpaceX for a rocket launch. The data will be collected and **ensured to be in the correct format from an API.**

Web scraping will be performed to collect **Falcon 9 historical launch records from Wikipedia** page titled "List of Falcon 9 and Falcon Heavy launches."



# Data Collection - SpaceX API

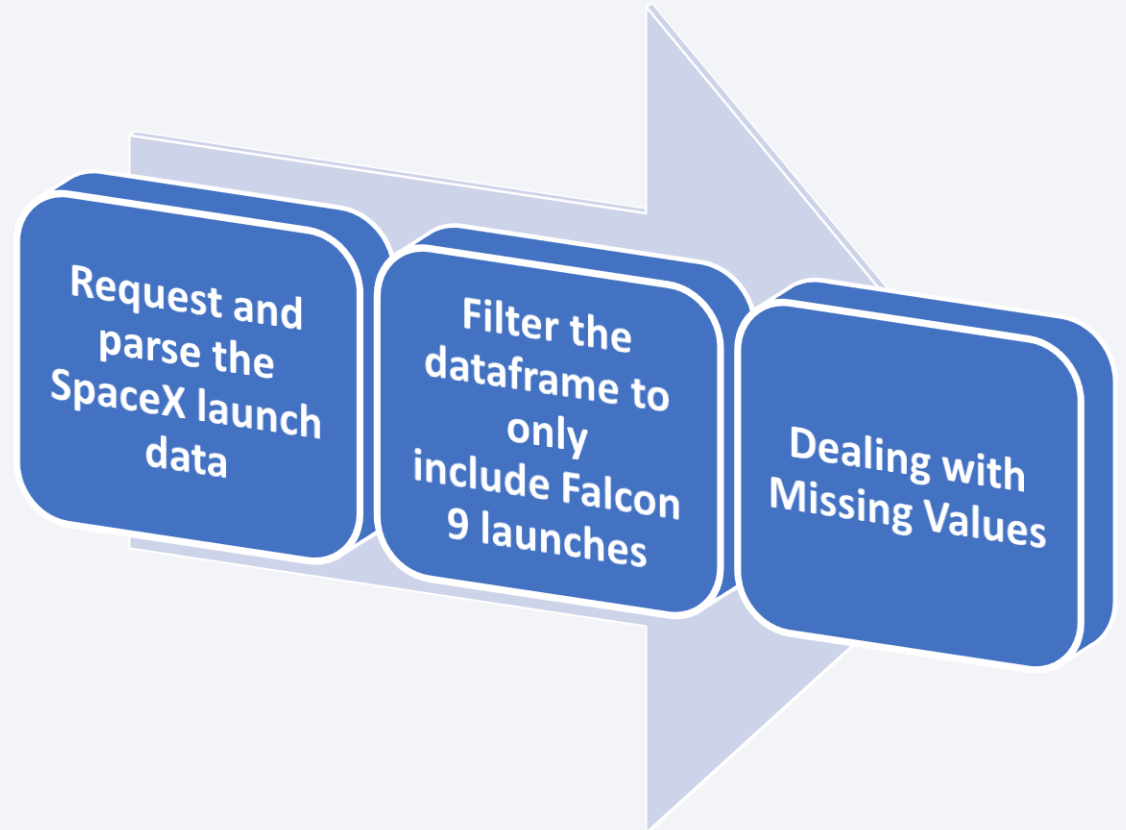
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## GitHub Reference:

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/1\\_Data%20Collection%20API.ipynb](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/1_Data%20Collection%20API.ipynb)

## CSV file outcome after methodology:

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/dataset\\_part\\_1.csv](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/dataset_part_1.csv)





# Data Collection - Scraping

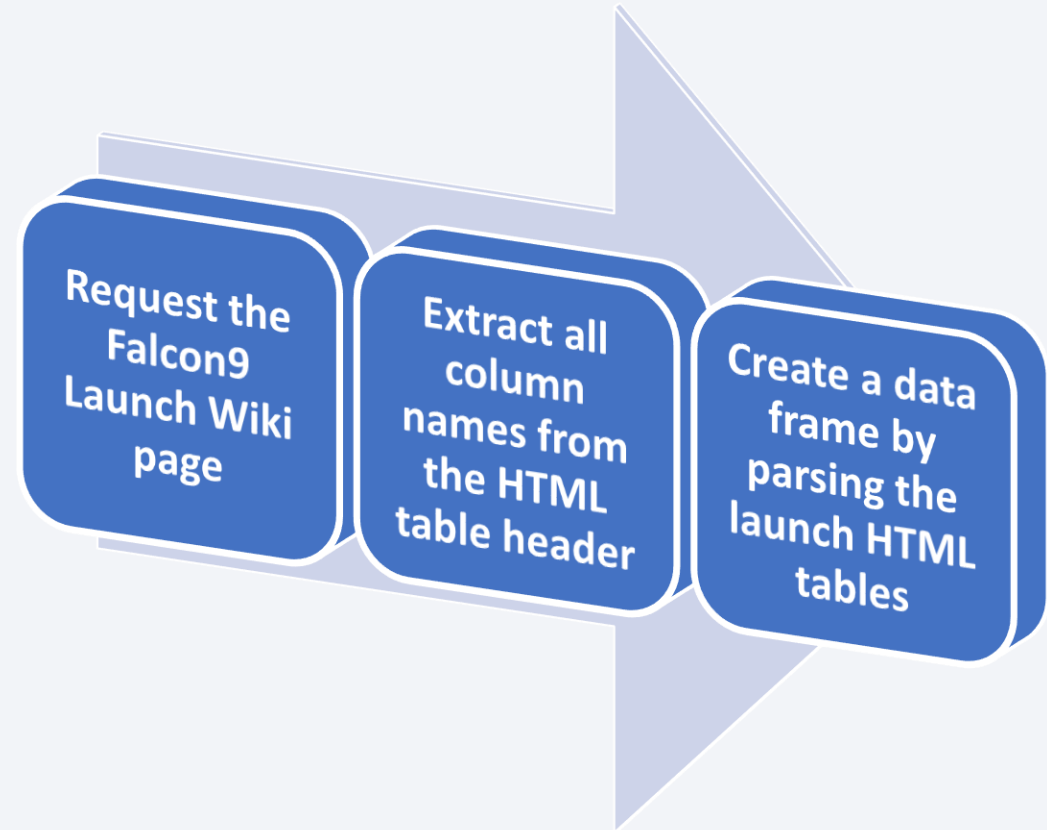
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## GitHub Reference:

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/2 Web scraping%20.ipynb](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/2%20Web scraping%20.ipynb)

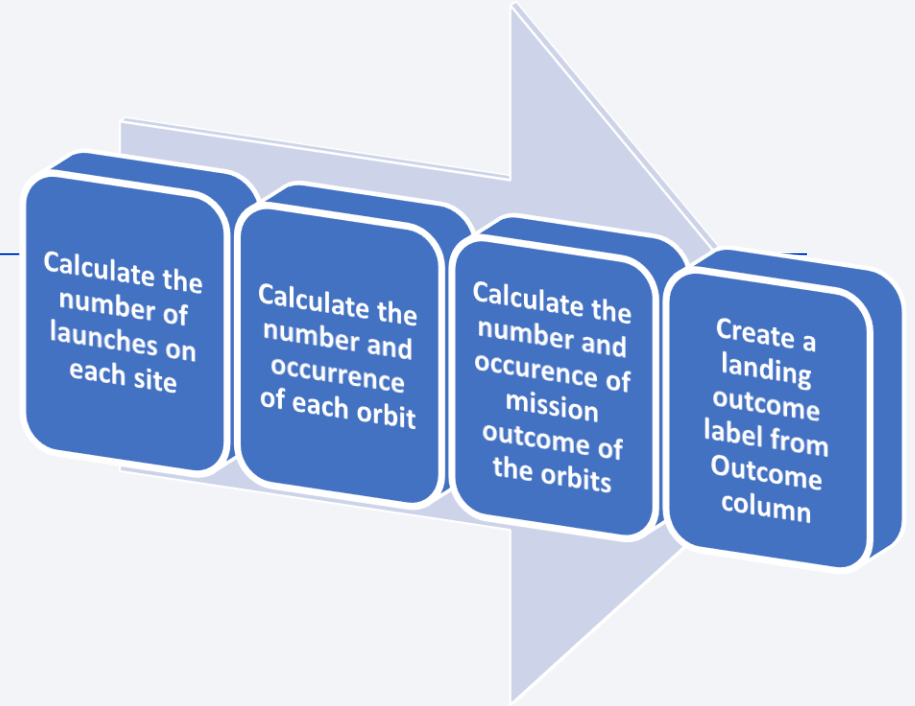
## CSV file outcome after methodology:

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/spacex web scraped.csv](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/spacex%20web%20scraped.csv)



# Data Wrangling

- Exploratory Data Analysis (EDA) will be performed to identify patterns in the data and determine the label for training supervised models.
- Within the dataset, various cases exist where the booster did not land successfully. Instances include attempted landings that failed due to accidents.
- The primary focus of this lab will be to transform these outcomes into training labels, where 1 signifies a successfully landed booster, and 0 signifies an unsuccessful landing.



## GitHub Reference:

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/3\\_Data%20Wrangling.ipynb](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/3_Data%20Wrangling.ipynb)

## CSV file outcome after methodology:

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/dataset\\_part\\_2.csv](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/dataset_part_2.csv)

# EDA with Data Visualization

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The relationship through the chart is visualised

- Scatterplots
- Barplots

Then, create dummy variables to categorical column and cast numeric columns to .float

**GitHub Reference:**

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/5 EDA Data%20Visualisation.ipynb](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/5%20EDA%20Data%20Visualisation.ipynb)

**CSV file outcome after methodology:**

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/dataset\\_part\\_3.csv](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/dataset_part_3.csv)

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Payload Mass vs Flight Number

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Launch Site vs Flight Number

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Launch Site vs Payload Mass

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Orbit vs Flight Number

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Payload vs Orbit

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Year vs Average Success Rate

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# EDA with SQL

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## Queries

- Names of the unique launch sites in the space mission;
- Top 5 launch sites whose name begin with the string 'CCA';
- Total payload mass carried by boosters launched by NASA (CRS);
- Average payload mass carried by booster version F9 v1.1;
- Date when the first successful landing outcome in ground pad was achieved;
- Names of the boosters which have success in drone ship and have payload mass between 4000 and 6000 kg;
- Total number of successful and failure mission outcomes;
- Names of the booster versions which have carried the maximum payload mass;
- Failed landing outcomes in drone ship, their booster versions, and launch site names for in year 2015;
- Rank of the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20.

## GitHub Reference:

[https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/4\\_EDA\\_SQLite.ipynb](https://github.com/andyngjw/IBM-DataSc-Capstone-SpaceX/blob/main/4_EDA_SQLite.ipynb)

# Build an Interactive Map with Folium

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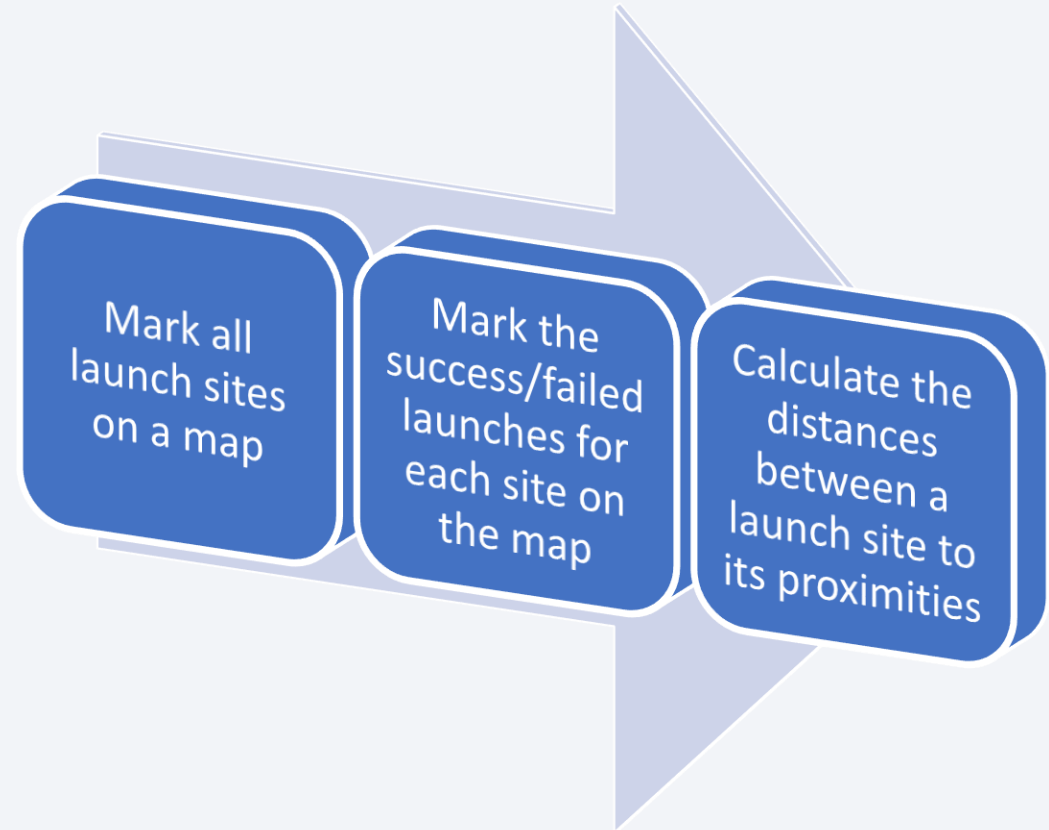
The launch success rate may depend on **location and proximities** of a launch site, i.e., the initial position of rocket trajectories.

Markers, circles, lines and marker clusters were used with Folium Maps

- Markers: points
- Circles: highlighted areas around specific coordinates
- Lines: distances between two coordinates
- Marker clusters: groups of events in each coordinate

**GitHub Reference:**

[https://github.com/andyngiw/IBM-DataSc-Capstone-SpaceX/blob/main/6 Data%20Visulisation Folium.ipynb](https://github.com/andyngiw/IBM-DataSc-Capstone-SpaceX/blob/main/6%20Data%20Visualisation%20Folium.ipynb)





# Build a Dashboard with Plotly Dash

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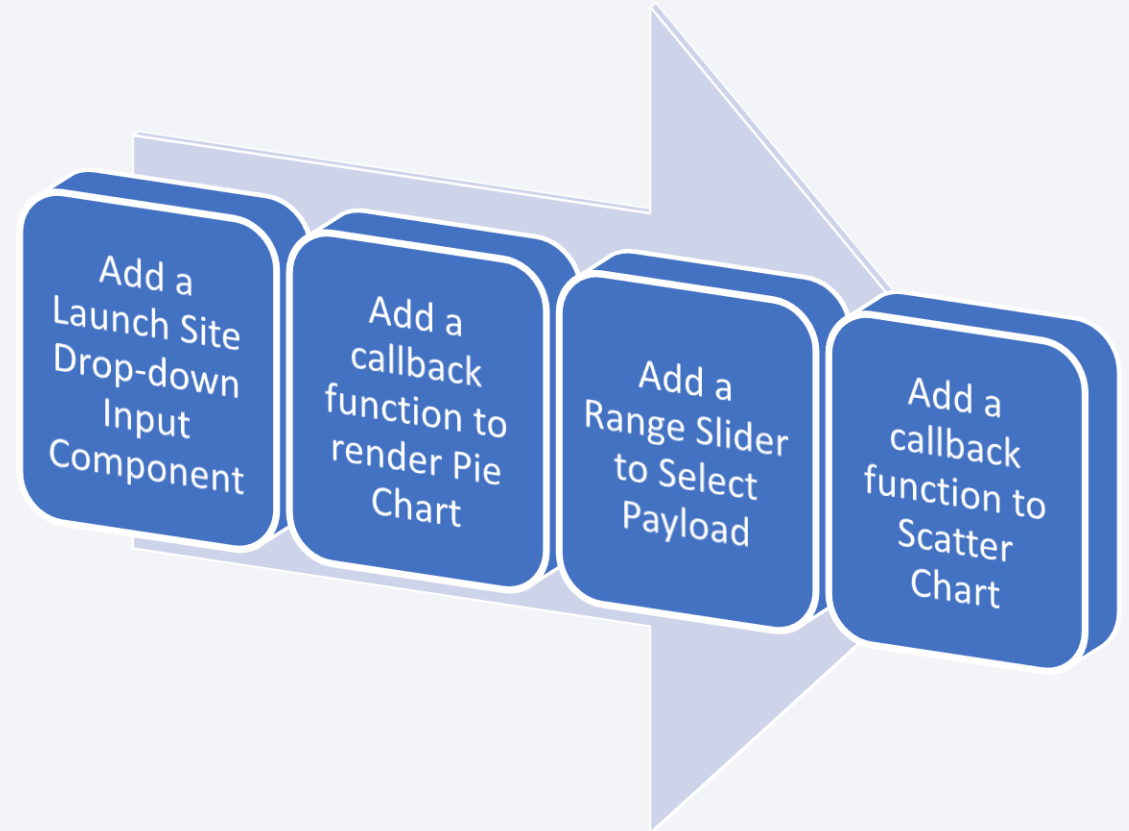
A Plotly Dash application will be built for users to engage in interactive visual analytics on SpaceX launch data in real-time.

The dashboard application contains input components, including:

- A dropdown list to interact with a pie chart and a scatter point chart
- A range slider to interact with a scatter point chart

## GitHub Reference:

[https://github.com/andyngiw/IBM-DataSc-Capstone-SpaceX/blob/main/7 Data%20Visulisation Plotly Dash.py](https://github.com/andyngiw/IBM-DataSc-Capstone-SpaceX/blob/main/7%20Data%20Visulisation%20Plotly%20Dash.py)



# Predictive Analysis (Classification)

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Exploratory Data Analysis will be performed

- Create a column for the class
- Standardize the data
- Split into training data and test data

Then, the method which performs best using test data is determined

*by finding the best Hyperparameter for SVM, Classification Trees and Logistic Regression*

**GitHub Reference:**

[https://github.com/andyngiw/IBM-DataSc-Capstone-SpaceX/blob/main/8\\_Machine%20Learning.ipynb](https://github.com/andyngiw/IBM-DataSc-Capstone-SpaceX/blob/main/8_Machine%20Learning.ipynb)

Logistics  
regression

Vector  
Support  
Machine

Decision Tree

k Nearest  
Neighbours

# Results

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## **Exploratory data analysis results through graphs**

- CCAFS SLC 40 has the highest success rate, compared to KSC LC 39A and VAFB SLC 4E
- The higher the payload mass, the higher the success rate
- ES-L1, GEO, HEO, and SSO orbit has 100% success rate
- Some orbits has higher success rate, but the total flight number in those orbits is less
- The payload does not show significant difference compared to orbit type
- The success rate shows improvement throughout the year starting from 2013 to 2020

## **Interactive analytics results**

- All launch sites were built in seashore area, and far away from the city, highway, and railway
- This is to ensure the safety of the citizens, and avoid causing further damage when the landing is unsuccessful
- KSC LC-39A has the highest success rate (41.2%), CCAFS LC-40 has the lowest success rate (14.4%)

## **Predictive analysis results**

- The Decision Tree Classifier shows the highest accuracy as it shows the highest value of true positive and true negative values



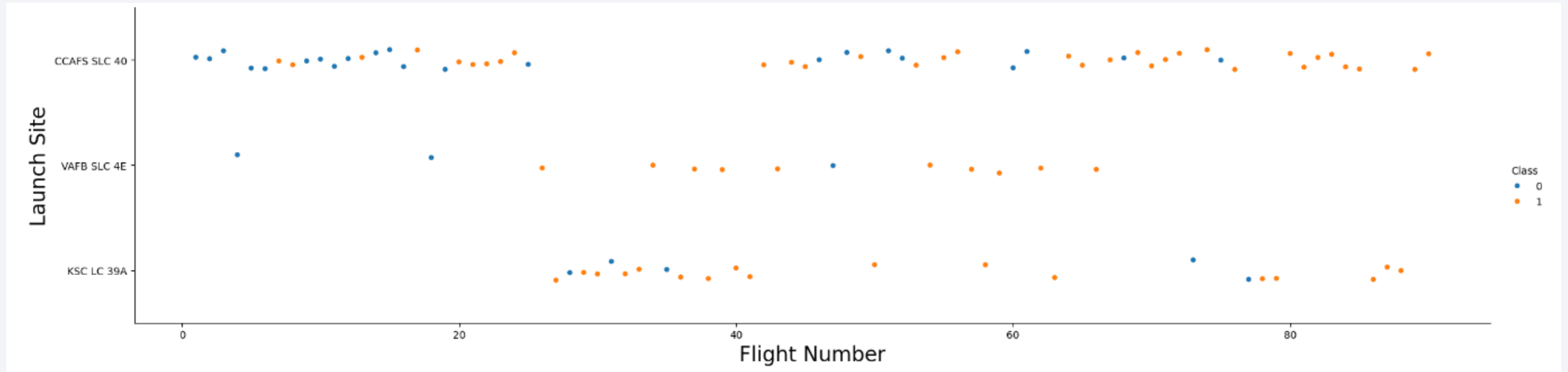
The background of the slide is an abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks in shades of red and cyan. A faint, light blue grid pattern is also visible, particularly in the lower-left quadrant. The overall effect is dynamic and technological.

Section 2

# Insights drawn from EDA



# Flight Number vs. Launch Site

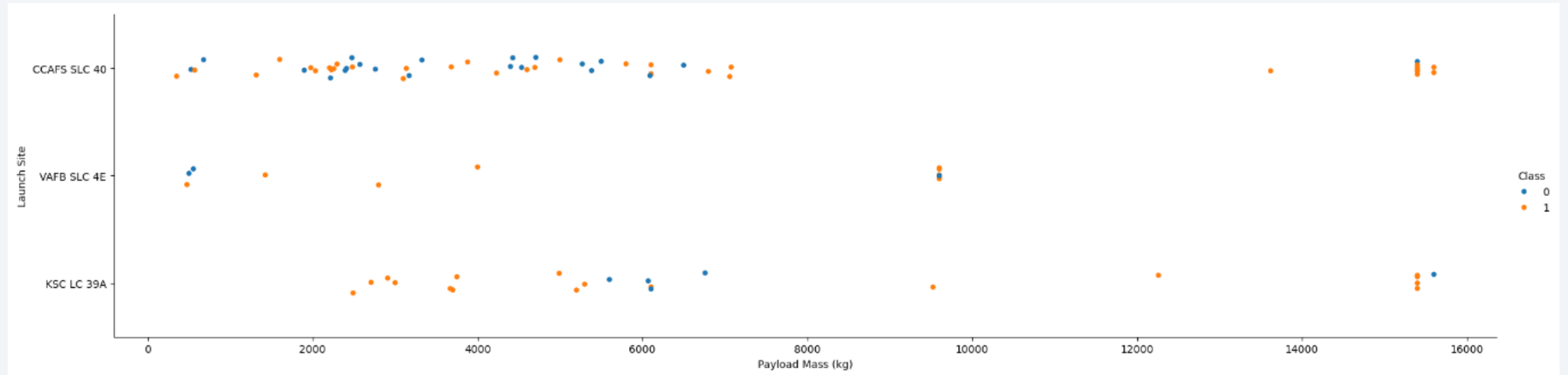


## Explanation:

- **Blue dots** represents the landing is **not successful**, **Orange dots** represents **successful landing**
- The higher the flight number, the later the flight is launched
- Results show the higher the flight number, the higher the success rate (More orange dots appear)
- **CCAFS SLC 40 has the highest success rate, compared to KSC LC 39A and VAFB SLC 4E**
- It infers the success landing rate is improved over time



# Payload vs. Launch Site

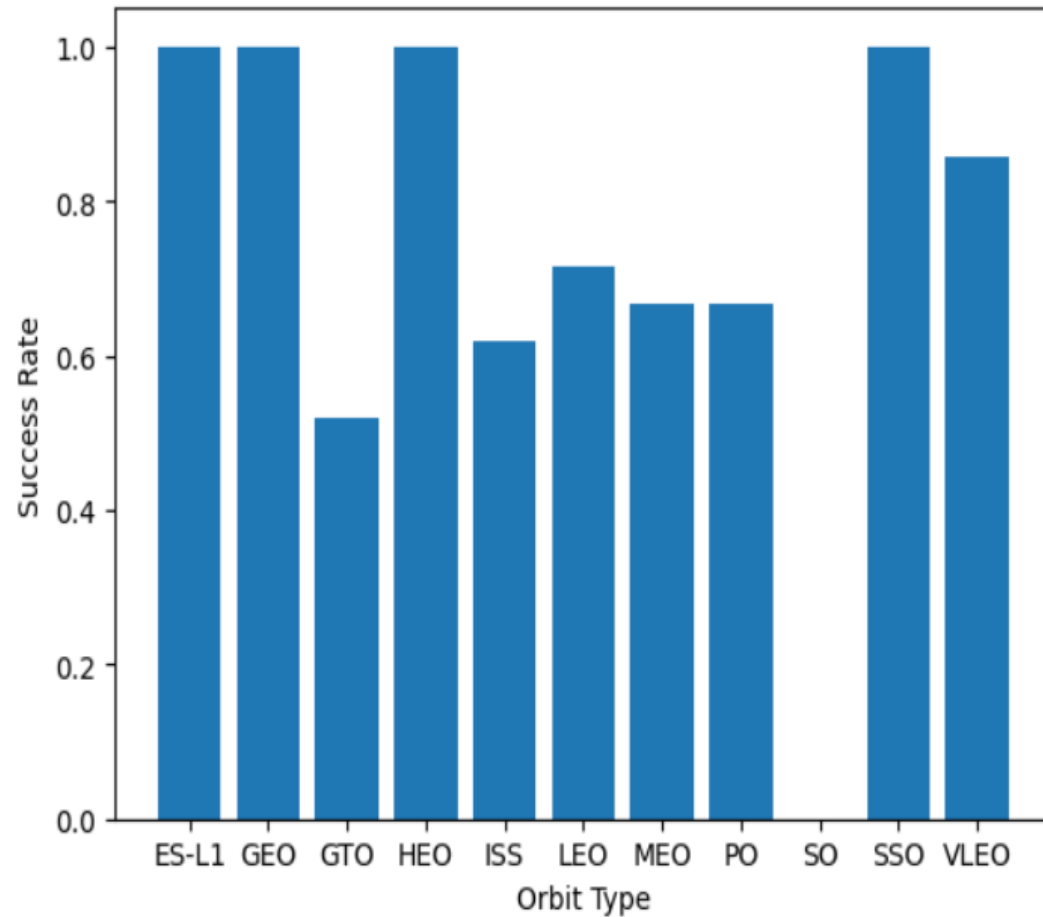


## Explanation:

- Results show the higher the payload mass, the higher the success rate
- The landing is **almost successful** when the payload mass is **more than 7000 kg**.
- The maximum payload mass for VAFB SLC 4E is about 9000 kg, while for CCAFS SLC 40 and KSC LC 35A, the maximum payload mass can reach to around 15000 kg.

# Success Rate vs. Orbit Type

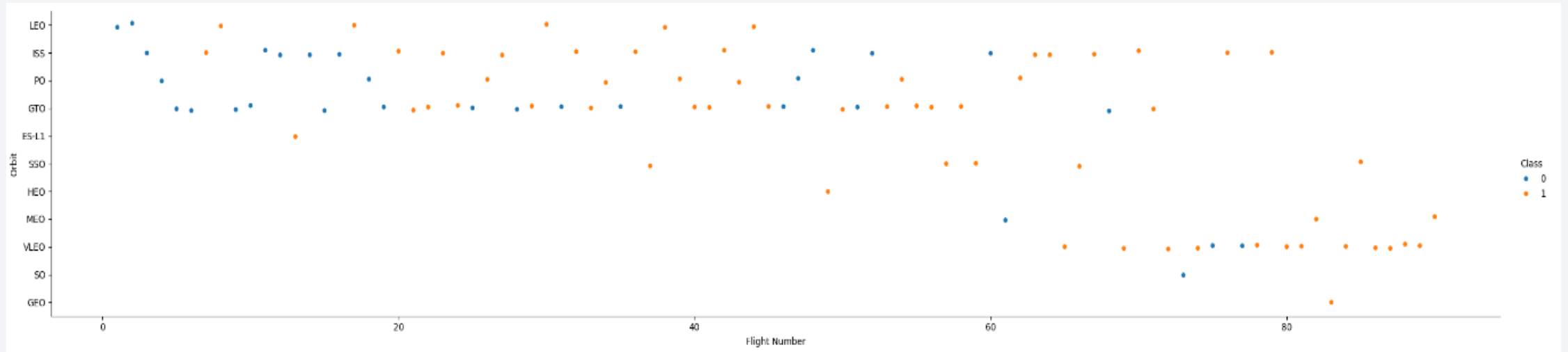
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## Explanation:

- **100% success rate:** ES-L1, GEO, HEO, and SSO
- **>80% success rate:** VLEO
- **Lowest success rate (0%):** SO

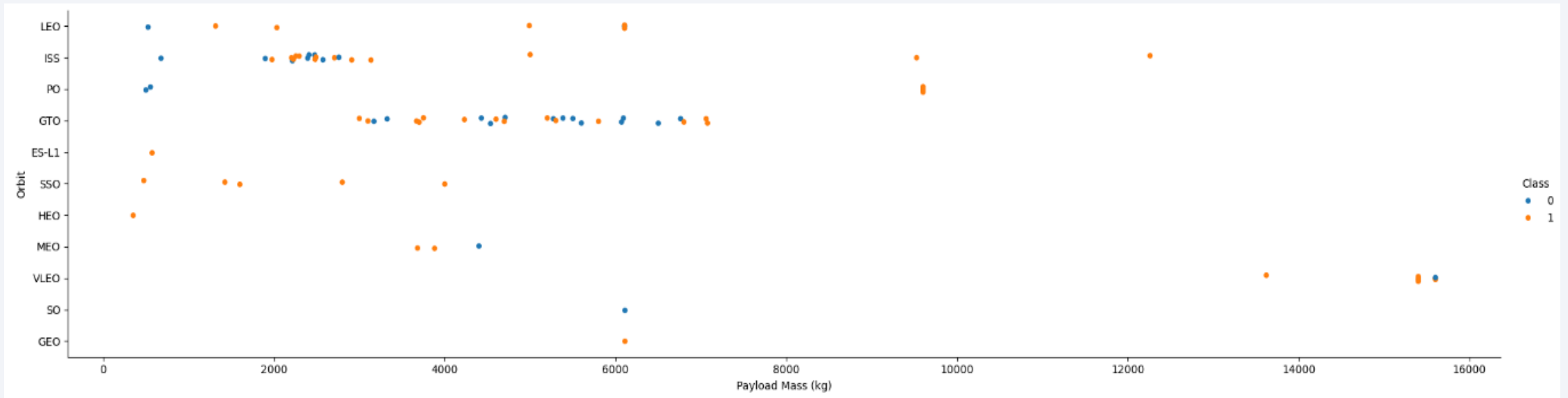
# Flight Number vs. Orbit Type



## Explanation:

- Some orbits has higher success rate, but the total flight number in those orbits is less
- It could be observed that the later the flight, the success rate is improved
- VLEO has a higher frequency in late stage, with high rate of success

# Payload vs. Orbit Type

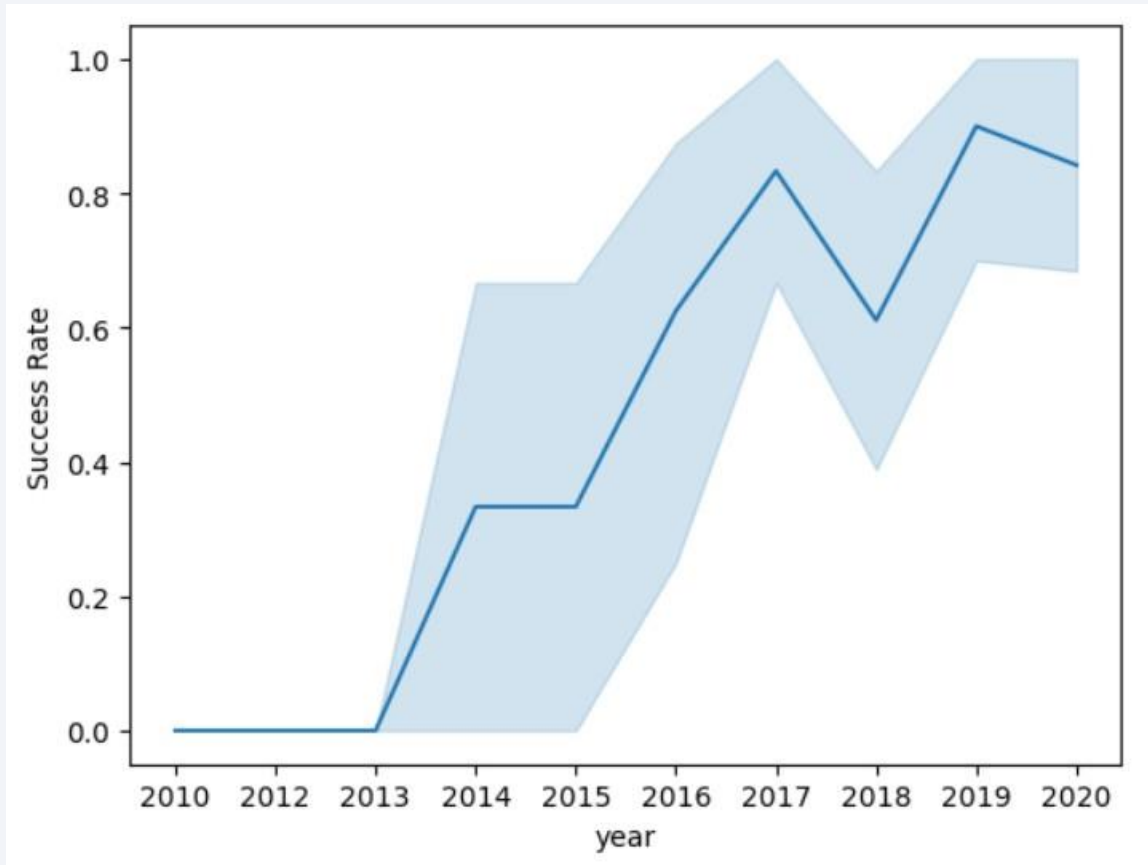


## Explanation:

- The payload does not show significant difference compared to orbit type
- The higher payload mass brings higher success rate in LEO and ISS

# Launch Success Yearly Trend

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## Explanation:

- The success rate shows improvement throughout the year starting from 2013 to 2020
- The success rate from 2010 to 2013 remains 0%, it could be inferred that the launching is still in development



# All Launch Site Names

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## Launch Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

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**Query:**

```
%sql SELECT DISTINCT(LAUNCH_SITE) FROM SPACEXTBL;
```

# Launch Site Names Begin with 'CCA'

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

## Query:

```
%sql SELECT * FROM SPACEXTBL WHERE LAUNCH_SITE LIKE 'CCA%' LIMIT 5;
```

**\*\*Only first 5 rows are shown in this query, to show the Launch Site 'CCAFS-LC 40'**

# Total Payload Mass

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**Total\_Payload\_Mass\_kg**

45596

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**Query:**

```
%sql SELECT SUM(PAYLOAD_MASS__KG_) AS Total_Payload_Mass_kg FROM SPACEXTBL WHERE CUSTOMER = 'NASA (CRS)';
```

# Average Payload Mass by F9 v1.1

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AVG(PAYLOAD_MASS__KG_)
------------------------

2928.4
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---

Query:

```
%sql SELECT AVG(PAYLOAD_MASS__KG_) FROM SPACEXTBL WHERE BOOSTER_VERSION = 'F9 v1.1';
```

# First Successful Ground Landing Date

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**MIN(DATE)**

2015-12-22

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**Query:**

**%sql SELECT MIN(DATE) \**

**FROM SPACEXTBL \**

**WHERE Landing\_Outcome = 'Success (ground pad)'**



# Successful Drone Ship Landing with Payload between 4000 and 6000

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## Payload

JCSAT-14

JCSAT-16

SES-10

SES-11 / EchoStar 105

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## Query:

```
%sql SELECT PAYLOAD \
FROM SPACEXTBL \
WHERE Landing_Outcome = 'Success (drone ship)' \
AND PAYLOAD_MASS__KG_ BETWEEN 4000 AND 6000;
```

# Total Number of Successful and Failure Mission Outcomes

---

<b>Mission_Outcome</b>	<b>total_number</b>
Failure (in flight)	1
Success	98
Success	1
Success (payload status unclear)	1

## Query:

```
%sql SELECT MISSION_OUTCOME, COUNT(*) as total_number \
FROM SPACEXTBL \
GROUP BY MISSION_OUTCOME;
```

# Boosters Carried Maximum Payload

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## 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

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## Query:

```
%sql SELECT BOOSTER_VERSION \
FROM SPACEXTBL \
WHERE PAYLOAD_MASS__KG_ = (SELECT MAX(PAYLOAD_MASS__KG_) FROM SPACEXTBL);
```

# 2015 Launch Records

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month	Date	Booster_Version	Launch_Site	Landing_Outcome
01	2015-01-10	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
04	2015-04-14	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

---

## Query:

```
%sql SELECT substr(Date,6,2) AS month, DATE,BOOSTER_VERSION, LAUNCH_SITE, [Landing_Outcome] \
FROM SPACEXTBL \
WHERE [Landing_Outcome] = 'Failure (drone ship)' and substr(Date,0,5)='2015';
```

**substr(Date,6,2) AS month** → This extracts a substring from the "Date" column starting at the 6th character and taking 2 characters. It is used to get the month part of the date.

**substr(Date,0,5) ='2015'** → This extracts a substring from the "Date" column starting at the 0th character (the beginning) and taking 5 characters. It is used to compare the first 5 characters of the date to check if they are equal to '2015'. This condition filters the results to include only those records where the year in the "Date" column is '2015'.

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

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Landing_Outcome	count_outcomes
No attempt	10
Success (drone ship)	5
Failure (drone ship)	5
Success (ground pad)	3
Controlled (ocean)	3
Uncontrolled (ocean)	2
Failure (parachute)	2
Precluded (drone ship)	1

## Query:

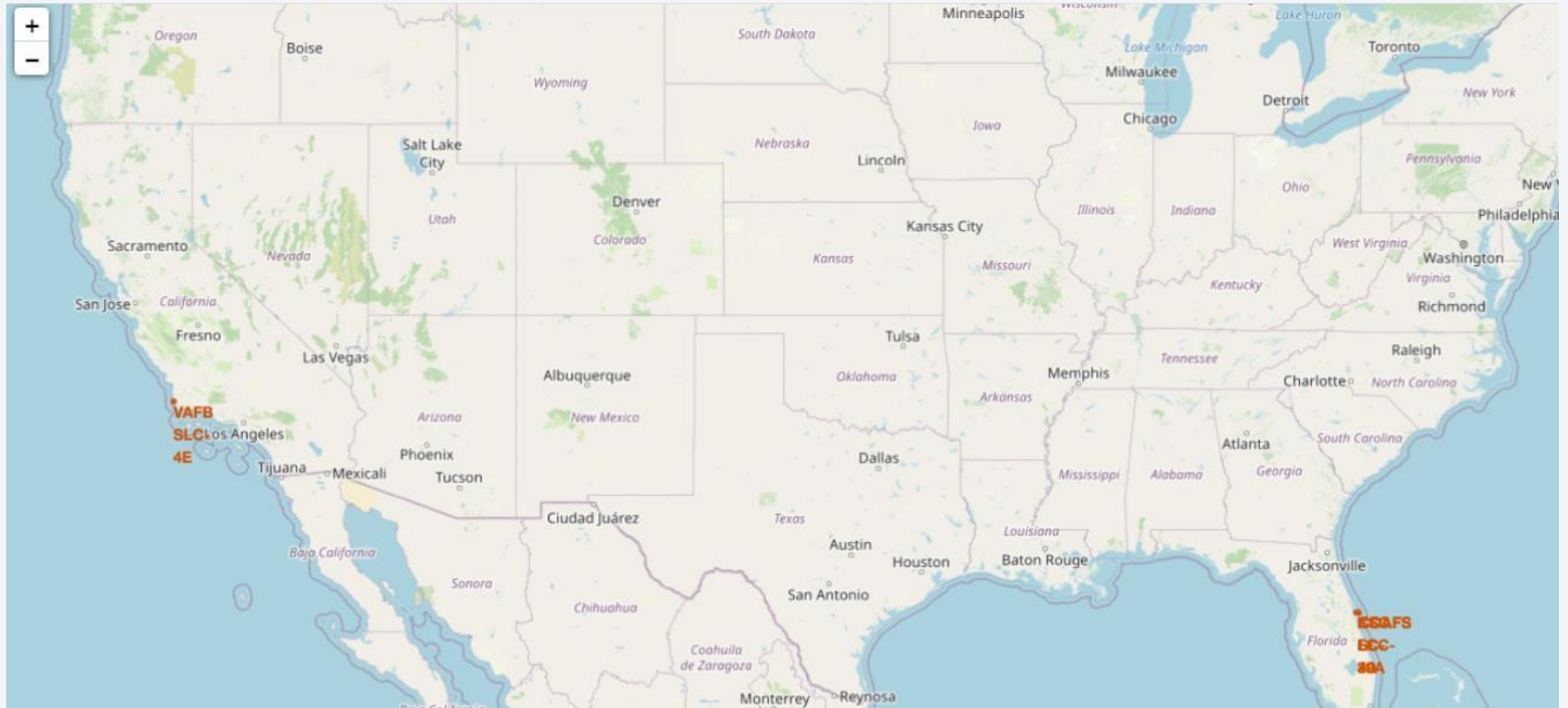
```
%sql SELECT Landing_Outcome, COUNT(*) as count_outcomes FROM SPACEXTBL \
WHERE DATE between '2010-06-04' and '2017-03-20' \
GROUP BY Landing_Outcome \
ORDER BY count_outcomes DESC;
```

A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The background is a deep blue gradient.

Section 3

# Launch Sites Proximities Analysis

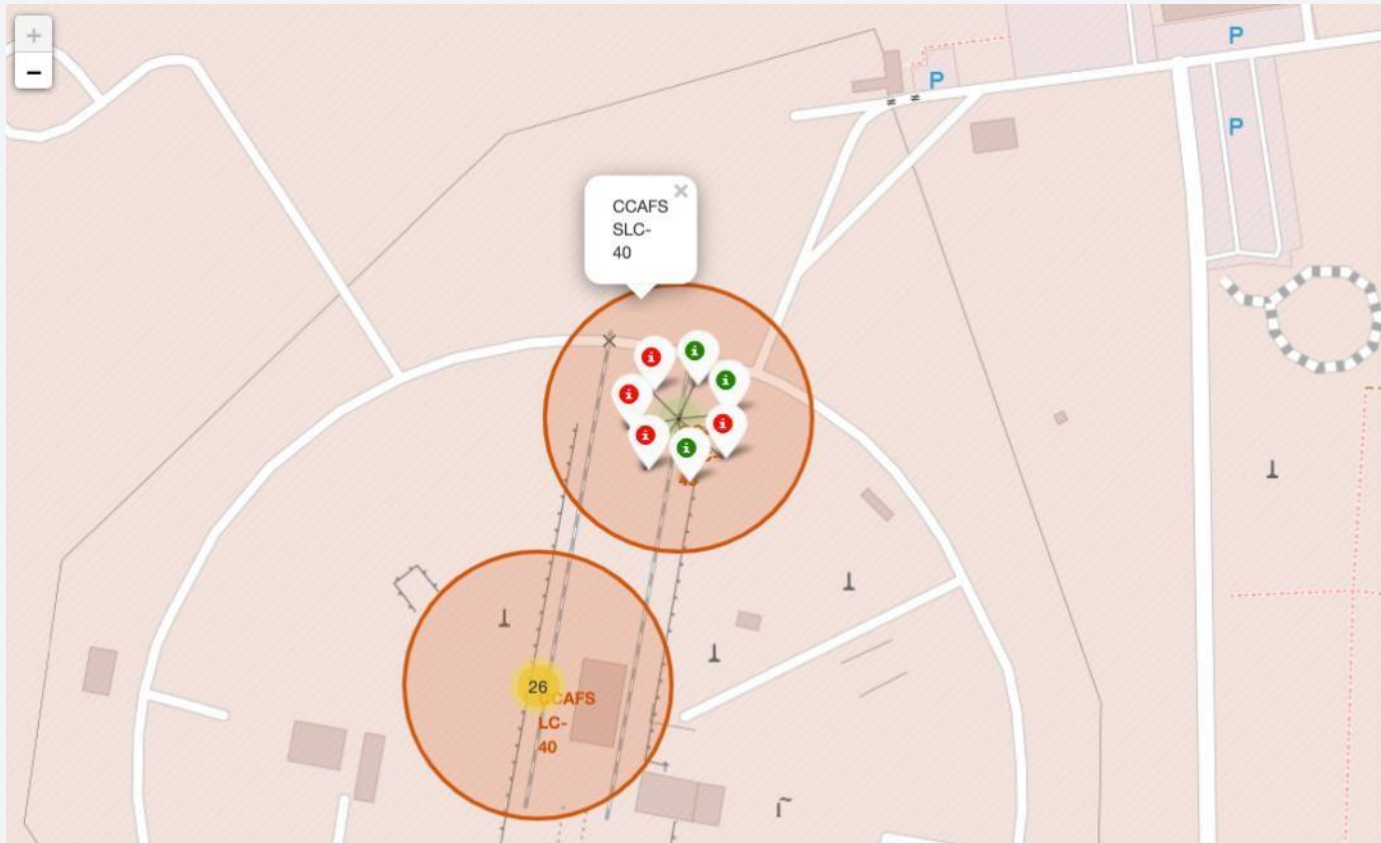
# Launch Sites



- All launch sites were in seashore area → for plannings if the landing is not successful



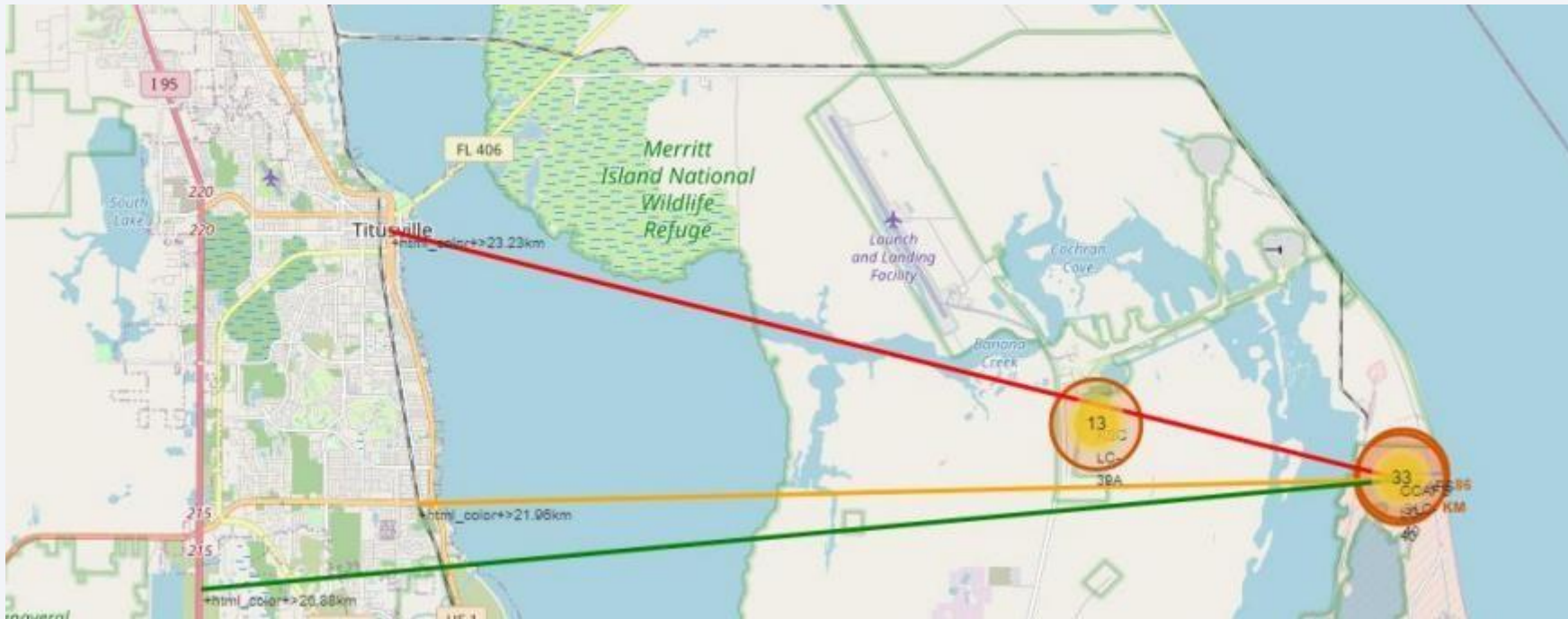
# Launch Outcomes by Site



- Showing one of the marker clusters in CCAFS SLC-20, it could be observed that the success rate of this site is 3 per 7.
- Another site could be analysed for the outcome through click the centre of the site, e.g. 26.

# Distance to Proximities

## Example: CCAFS SLC-20



- It is built very near to the coastline, while far away from the city, highway, and railway
- To ensure the safety of the citizens, and avoid causing further damage when the landing is unsuccessful





Section 4

# Build a Dashboard with Plotly Dash

# Total Success Launches by Site

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Total Success Launches by Site



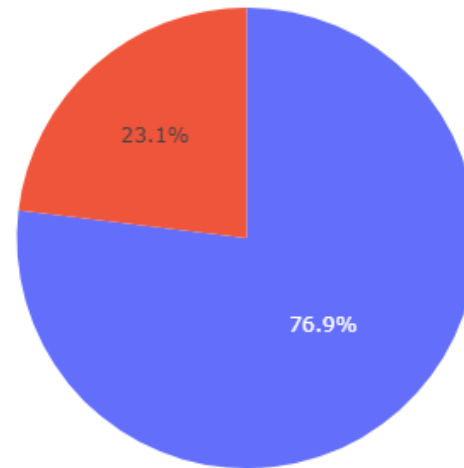
By comparing the total success launches by site,

- **KSC LC-39A has the highest success rate (41.2%)**
- **CAFS LC-40 has the lowest success rate (14.4%)**

# Launch Success Analysis for KSC LC-39A

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Total Launches for site KSC LC-39A



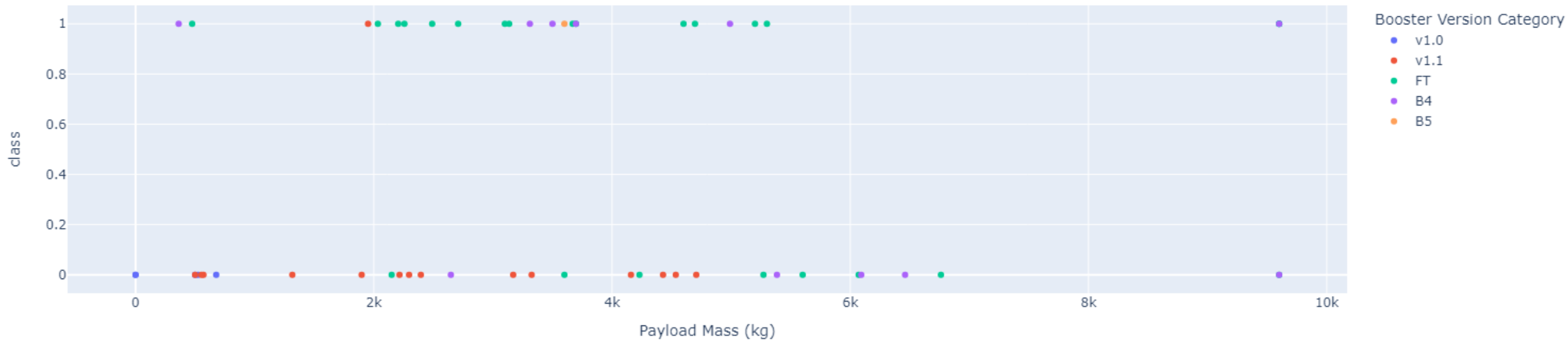
■ 1  
■ 0

In total success launches by site KSC LC-39A,

- **The success rate reaches 76.9%**

# Payload vs Success Rate

Correlation Between Payload and Success for All Sites



- In the graph, 1 = success, 0 = not success
- **In successful launches**, the payload mass is normally between 2000kg and 6000kg
- **In successful launches**, FT Booster has the highest prevalence, showing the success rate is the highest when FT Booster is used

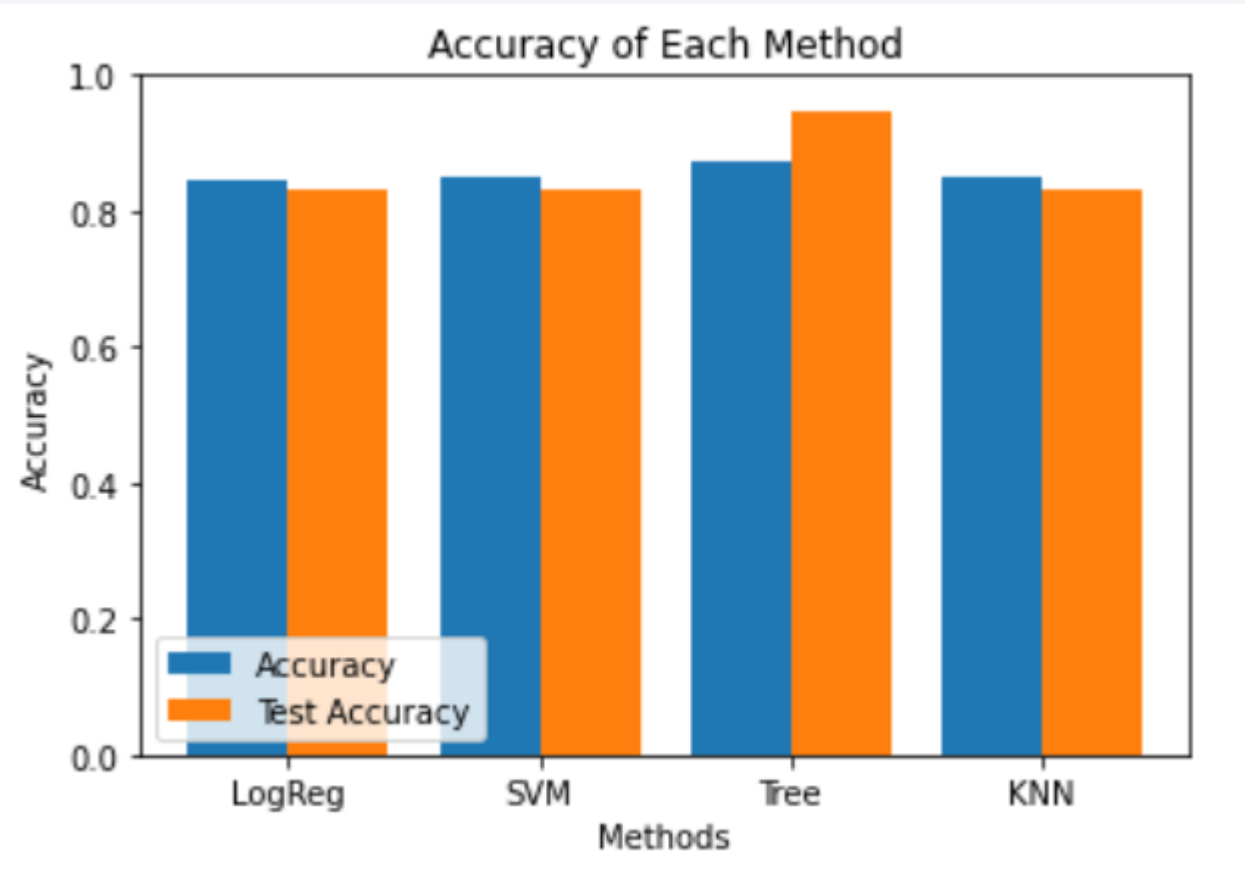


Section 5

# Predictive Analysis (Classification)



# Classification Accuracy

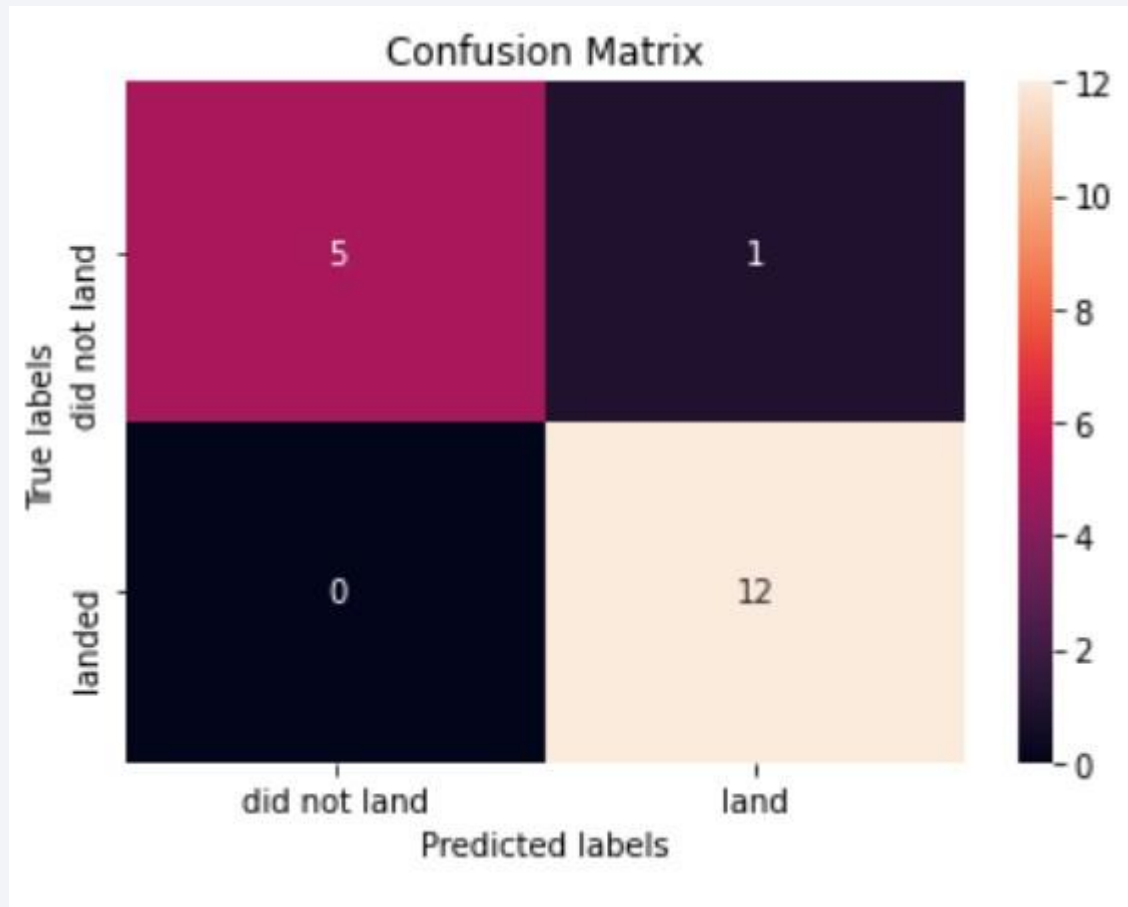


- Through the accuracy of the models, the decision tree model has the highest values among others.
- The graph is plotted based on the table below, by showing accuracy score using each ML method:

	ML Method	Accuracy Score (%)
0	Support Vector Machine	83.333333
1	Logistic Regression	83.333333
2	K Nearest Neighbour	94.444444
3	Decision Tree	83.333333



# Confusion Matrix



## Example: Decision Tree Classifier

- The Decision Tree Classifier shows the highest accuracy as it shows the highest value of true positive and true negative values.
- The false positive (Type I) is 1, while the false negative (Type II) is 0.

# Conclusions

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- The analysis reveals that CCAFS SLC 40 boasts the highest success rate among launch sites, and there is a positive correlation between payload mass and mission success.
- Certain orbits consistently achieve a 100% success rate, but some high-performing orbits have fewer overall flights.
- Over the years, there is a noticeable improvement in success rates. Interactive analytics highlight strategic launch site locations for safety, with KSC LC-39A having the highest success rate and CCAFS LC-40 the lowest.
- The Decision Tree Classifier stands out in predictive analysis, demonstrating the highest accuracy.
- Overall, the findings provide insights into the factors influencing space mission success, encompassing launch site performance, payload characteristics, and orbital considerations.

# Appendix

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- GitHub Project Repository!

[https://github.com/rhuang7/DataScience\\_IBM](https://github.com/rhuang7/DataScience_IBM)

Thank you!

