



IBM Developer  
SKILLS NETWORK

# Winning Space Race with Data Science

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# Executive Summary

## Methodologies

### Data Collection

- REST API
- Web Scraping

### Data Wrangling

- Creation of binary Outcome Variable

### Visualization

- Basic Graphs
- Maps using Folium
- Dashboard

### Analysis

- Basic Analysis
- Predictive Models

## Results

- Improved Success rate over the time
- KSC LC-39A is the most successful site
- ES-L1, GEO, HEO, and SSO have the highest success rate
- All sites are close to coastline
- Decision Trees model has the highest accuracy

# Introduction

## Project background and context

- SpaceX is a leader in the space industry. It has launched several missions in some of which first stage has been successful and unsuccessful in others. We need to analyze publicly available data for a competitor of SpaceX.

## Specific questions

- How various factors (launch site, payload, orbit etc) affect success/failure of first stage landing?
- Is there any trend in success rate of first stage landings over time?
- What ML model best predicts success?



Section 1

# Methodology

# Methodology



## **Data collection:**

Collected through REST API and Web Scraping

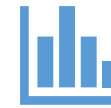


## **Perform data wrangling**

Filtering, missing values handling, one hot encode



## **Visualization and EDA with SQL**



## **Visual analytics using Folium and Plotly Dash**



## **Predictive Analytics**

Various classification models built and tested

# Data Collection

1

Request/Fetch  
Data

2

Convert to  
Data Frame

3

Filter Data

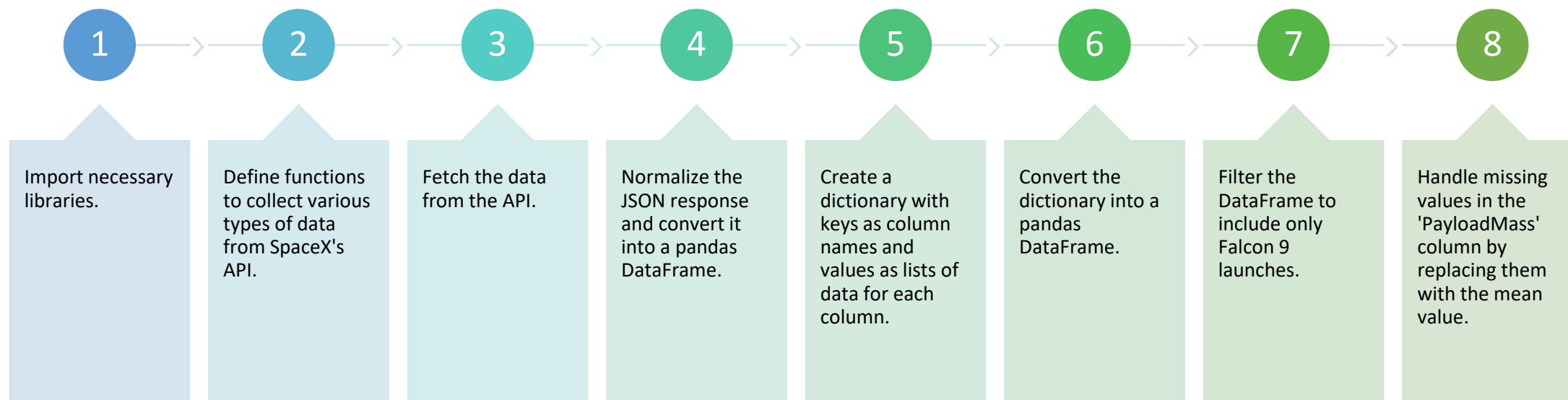
4

Replace  
missing values

5

Export to csv

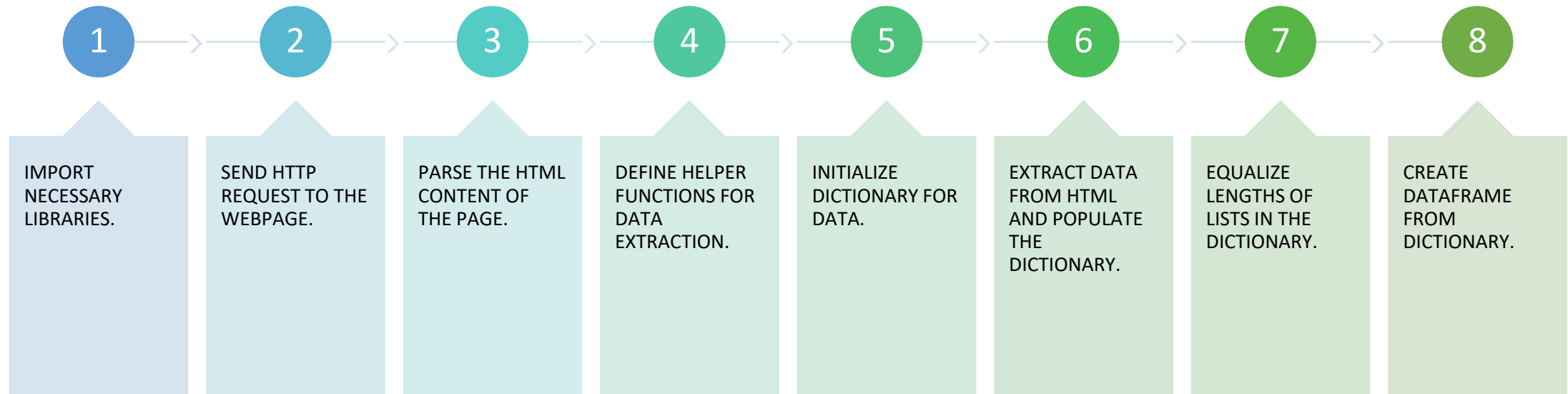
# Data Collection – SpaceX API



Github Link: [https://github.com/Atifhussain40/data\\_science\\_capstone/blob/main/api%20collection.ipynb](https://github.com/Atifhussain40/data_science_capstone/blob/main/api%20collection.ipynb)

# Data Collection - Scraping

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Github Link: [https://github.com/Atifhussain40/data\\_science\\_capstone/blob/main/web%20scraping.ipynb](https://github.com/Atifhussain40/data_science_capstone/blob/main/web%20scraping.ipynb)



# Data Wrangling



Import necessary libraries (**numpy**, **pandas**)



Load the dataset



**Data analysis:**

Missing values percentage  
Column types



**Calculations:**

Number of launches on each site  
Number and occurrence of each orbit  
Number and occurrence of mission outcome per orbit type  
New variable 'Class'



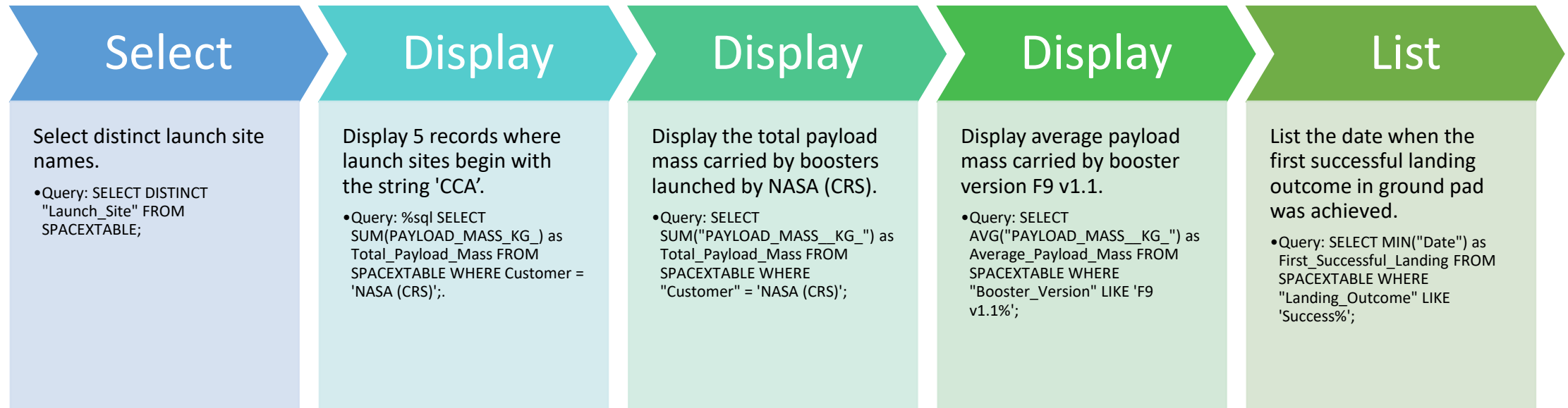
Export the dataframe to a CSV file

Github Link: [https://github.com/Atifhussain40/data\\_science\\_capstone/blob/main/data%20wrangling.ipynb](https://github.com/Atifhussain40/data_science_capstone/blob/main/data%20wrangling.ipynb)

# EDA with Data Visualization

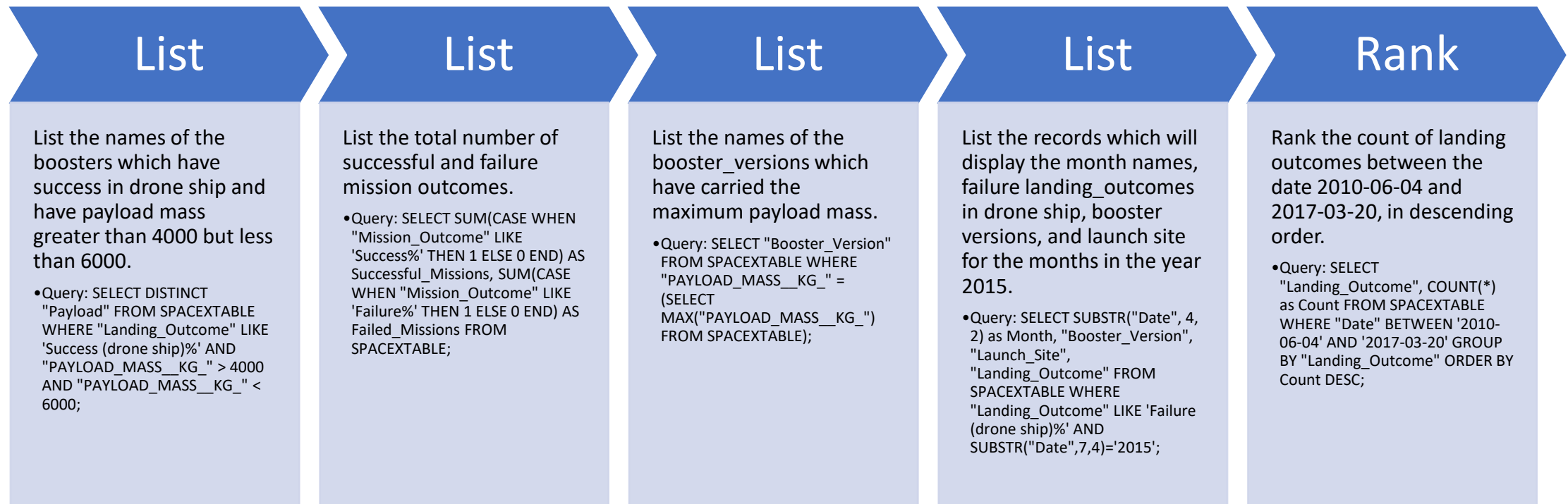
Plot	Purpose
FlightNumber vs. PayloadMass	Investigate the impact of flight number and payload mass on landing success
FlightNumber vs. LaunchSite	Examine the influence of flight number and launch site on success rates
Payload vs. LaunchSite	Determine the relationship between payload mass and choice of launch site
Success Rate vs. Orbit Type	Identify which orbit types have higher success rates
FlightNumber vs. Orbit Type	Explore the correlation between flight number and orbit type
Payload vs. Orbit Type	Analyze how payload mass and orbit type affect landing success
Yearly Success Trend	Assess the trend of launch success rate over time

# EDA with SQL 1/2



Github Link: [https://github.com/Atifhussain40/data\\_science\\_capstone/blob/main/sql%20analysis.ipynb](https://github.com/Atifhussain40/data_science_capstone/blob/main/sql%20analysis.ipynb)

# EDA with SQL 2/2



# Build an Interactive Map with Folium

Object Added	Purpose
<code>folium.map.Marker</code>	Mark the exact location of each launch site with a text label marker. The marker has the launch site name.
<code>folium.PolyLine</code>	Draw a line from the launch site to the coastline.
<code>folium.PolyLine</code>	Draw a line from a nearby city to the launch site and display the distance.
<code>folium.PolyLine</code>	Draw a line from a nearby railway to the launch site and display the distance.
<code>folium.PolyLine</code>	Draw a line from a nearby highway to the launch site and display the distance.

Github Link: [https://github.com/Atifhussain40/data\\_science\\_capstone/blob/main/Folium.ipynb](https://github.com/Atifhussain40/data_science_capstone/blob/main/Folium.ipynb)



# Build a Dashboard with Plotly Dash

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Chart	Purpose
Chart- Successful Launches by Site	To see which site has the highest share in the successful launches
Chart – KSC LC 39A Success Failure comparison	To see the percentage of successful launches from the site
Scatter Chart	To see correlation between payload and launch success

# Predictive Analysis (Classification)

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- Data Loaded
  - Models Built
  - Models tuned using GridSearchCV
  - Models compared using accuracy score
  - Best model selected
- 
- Github Link: [https://github.com/Atifhussain40/data\\_science\\_capstone/blob/main/Predictive.ipynb](https://github.com/Atifhussain40/data_science_capstone/blob/main/Predictive.ipynb)

# Main Results

- Success rate has improved over the time
- KSC LC 39A has the best success rate
- Decision Trees based model performs best in predicting outcome

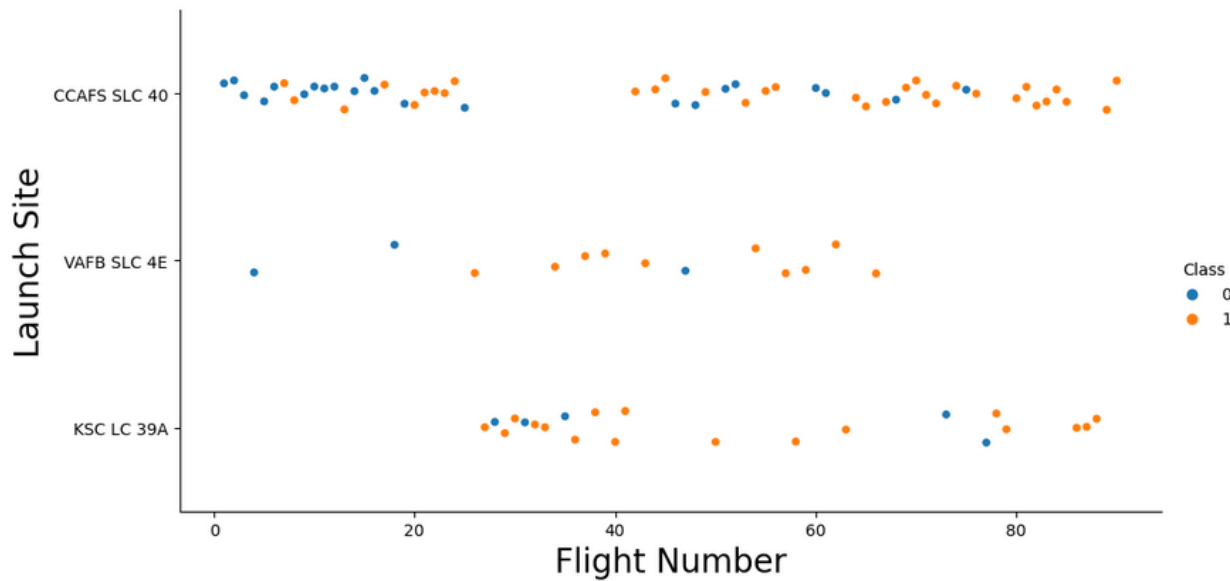


The background of the slide is a complex, abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks and lines in shades of red and cyan. These lines vary in thickness and opacity, creating a sense of depth and movement. A faint, light blue grid pattern is also visible, particularly in the lower-left quadrant. The overall effect is a high-tech, digital aesthetic.

Section 2

# Insights drawn from EDA



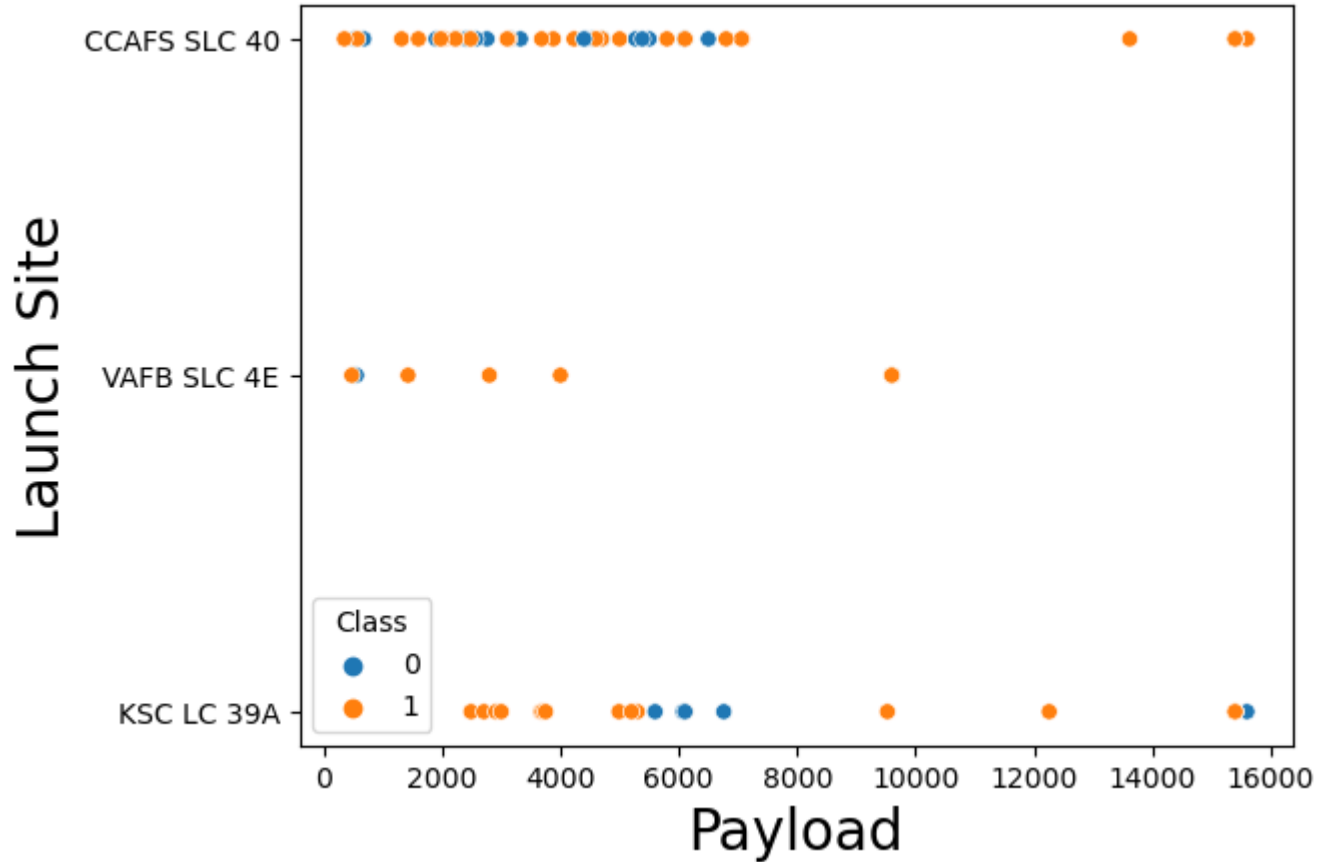


## Flight Number vs. Launch Site

- CCAFS SLC 40 has the highest number of flights, however, many of those have been unsuccessful
- KSC LC 39A has very few unsuccessful flights

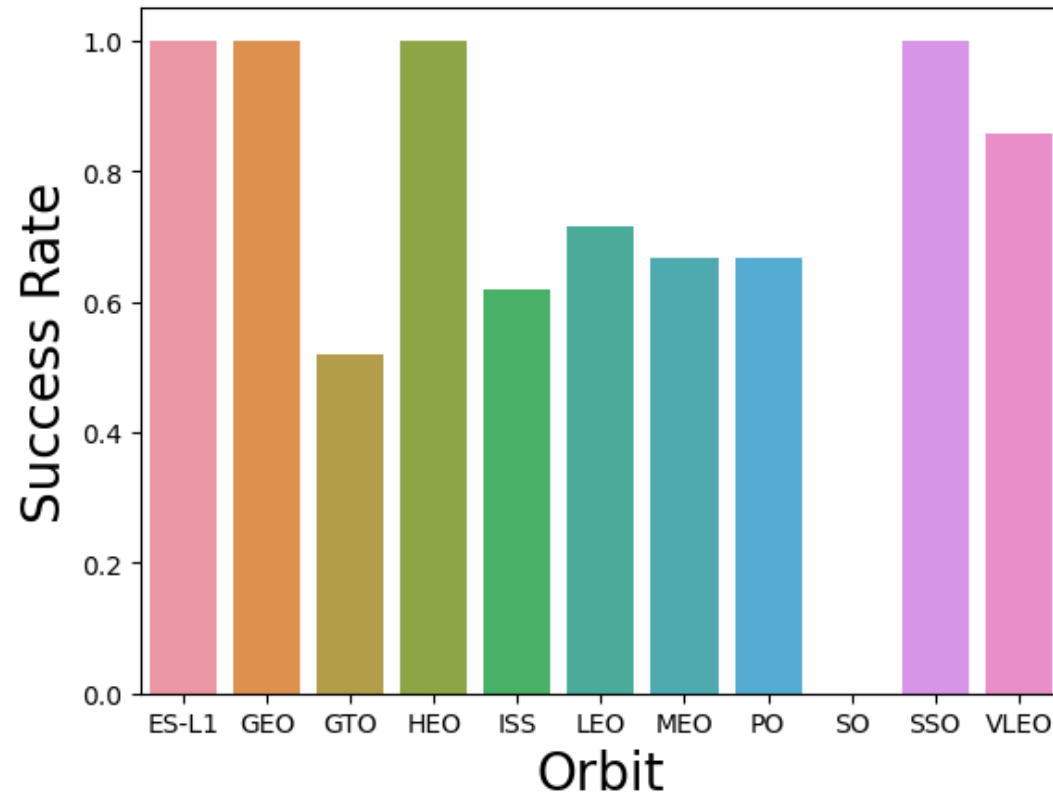


# Payload vs. Launch Site



- CCAFS SLC 40 has more success with greater mass loads
- KSC LC 39A also has greater success with small mass loads

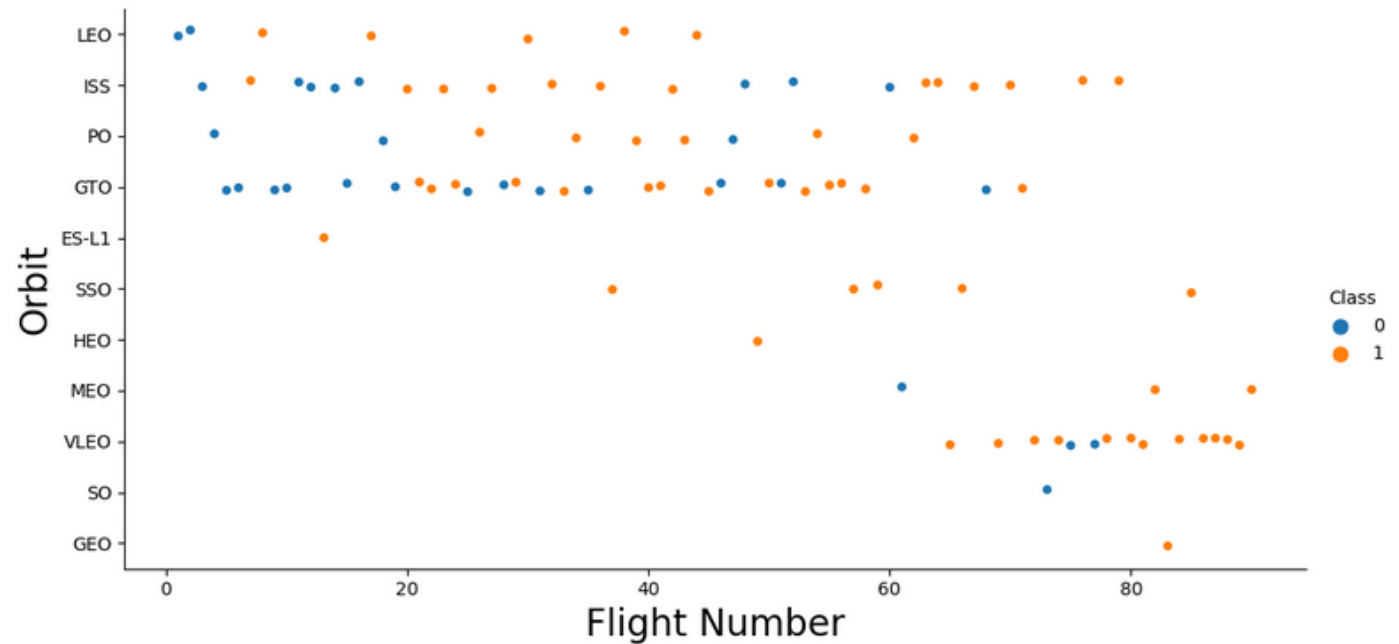
# Success Rate vs. Orbit Type



- ES-L1, GEO, HEO and SSO have the highest success rates
- GTO orbit has the lowest success rate

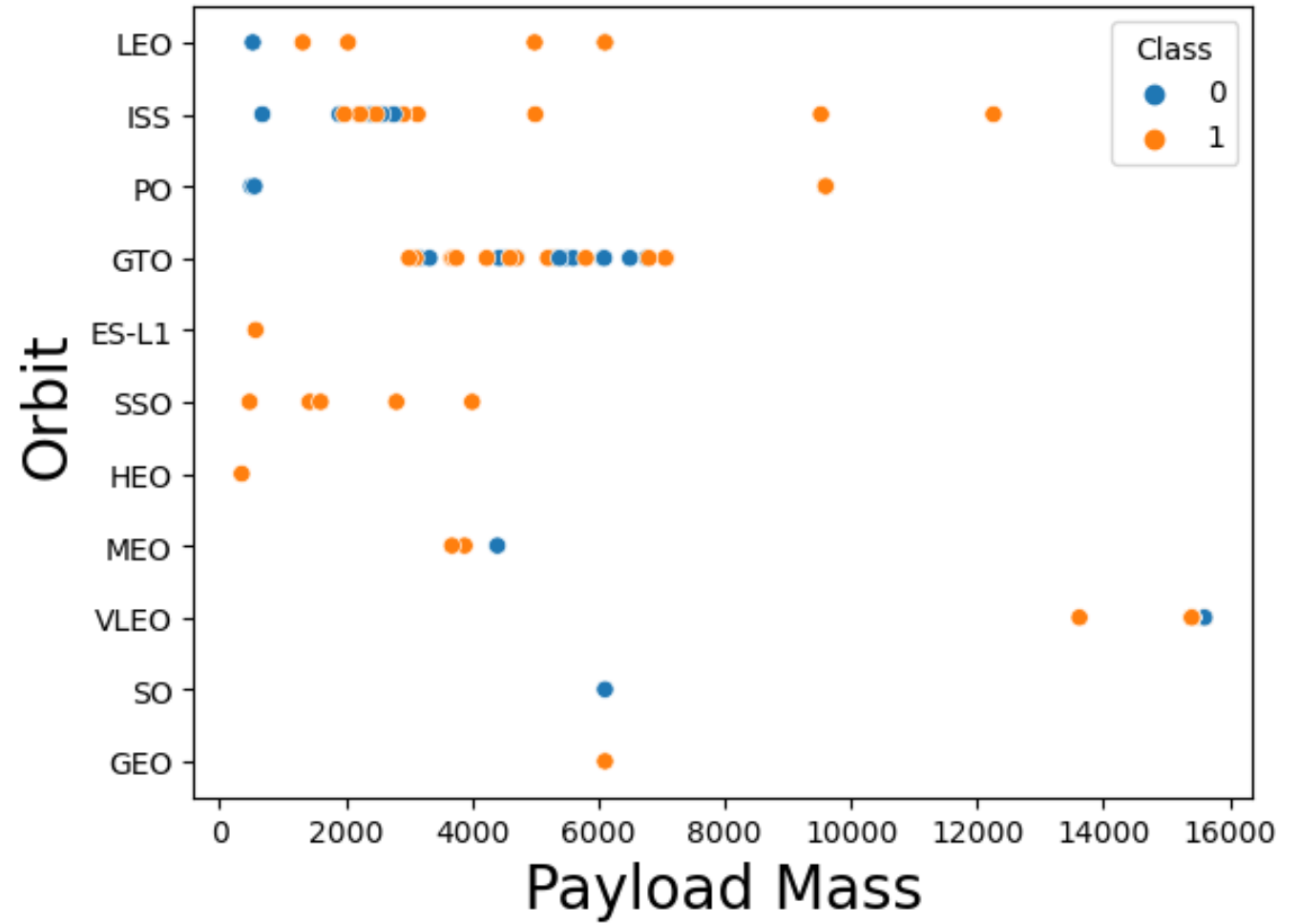
## Flight Number vs. Orbit Type

- LEO orbit, success is related to the flight number whereas
- In the GTO orbit, there is no relationship between flight number and the success



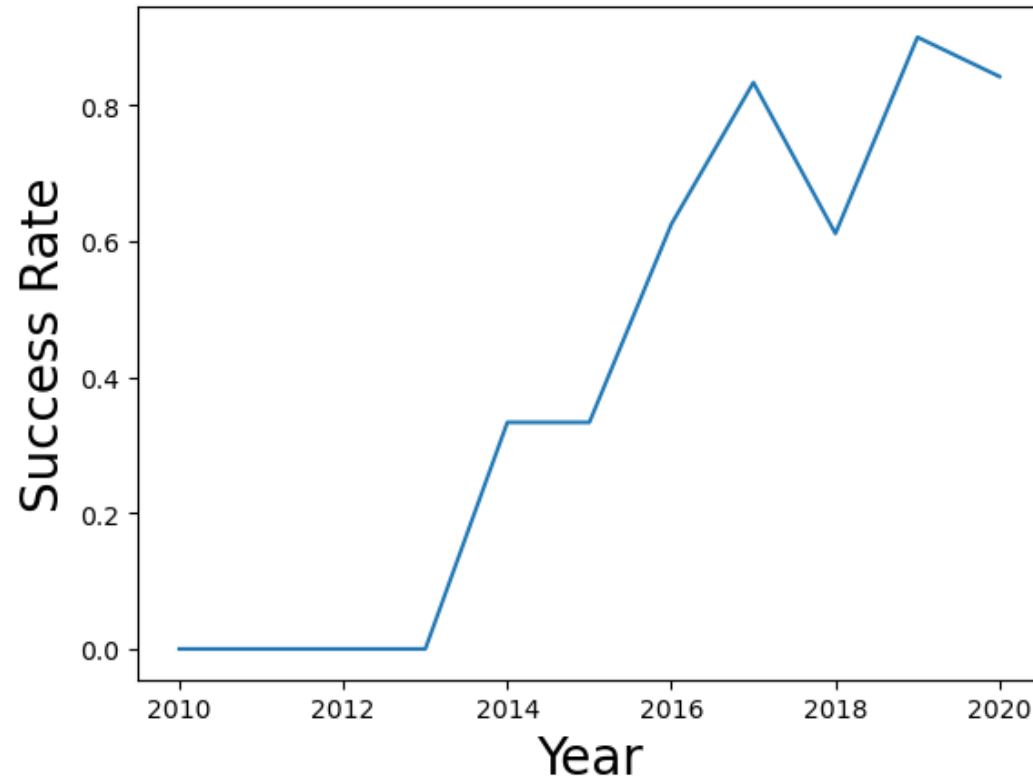
## Payload vs. Orbit Type

In ISS orbit, success is associated with higher payloads



# Launch Success Yearly Trend

Success rate has generally increased over the years.





# All Launch Site Names

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Query has selected  
distinct/unique launch site  
names from SPACEXTABLE

In [16]:

```
%sql SELECT DISTINCT "Launch_Site" FROM SPACEXTABLE;
```

```
* sqlite:///my_data1.db  
Done.
```

Out[16]:

Launch_Site
-------------

CCAFS LC-40
-------------

VAFB SLC-4E
-------------

KSC LC-39A
------------

CCAFS SLC-40
--------------

# Launch Site Names Begin with 'CCA'

Selected records where  
launch sites began  
with `CCA` using LIKE

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG_	Orbit	Customer	Mission
06/04/2010	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0.0	LEO	SpaceX	
12/08/2010	15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0.0	LEO (ISS)	NASA (COTS) NRO	
22/05/2012	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525.0	LEO (ISS)	NASA (COTS)	
			CCAFS	SpaceX		LEO	NASA	

# Total Payload Mass

```
In [22]: %sql SELECT SUM("PAYLOAD_MASS__KG_") as Total_Payload_Mass FROM SPACEXTABLE WHERE "Customer" = 'NASA (CRS)';  
* sqlite:///my_data1.db  
Done.
```

```
Out[22]: Total_Payload_Mass  
         45596.0
```

Total payload carried by boosters from NASA is 45,596 Kgs

# Average Payload Mass by F9 v1.1

```
In [12]: %sql SELECT AVG("PAYLOAD_MASS_KG_") FROM SPACEXTABLE WHERE "Booster_Version" = 'F9 v1.1';
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
Out[12]: AVG("PAYLOAD_MASS_KG_")
```

```
2928.4
```

**Average Payload Mass carried by F9 v1.1 is 2534.67 Kgs**

# First Successful Ground Landing Date

```
: %sql SELECT MIN("Date") FROM SPACEXTABLE WHERE "Landing_Outcome" = 'Success (ground pad)';
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
: MIN("Date")
```

```
01/08/2018
```

**The first successful ground landing was on 01/07/2020**



# Successful Drone Ship Landing with Payload between 4000 and 6000

```
In [27]: %sql SELECT DISTINCT "Payload" FROM SPACEXTABLE WHERE "Landing_Outcome" LIKE 'Success (drone ship)%' AND "PAY
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
Out[27]:
```

Payload
JCSAT-14
JCSAT-16
SES-10
SES-11 / EchoStar 105

Names of the boosters with payloads between 4000 and 6000 and successful landing are JCSAT-14, JCSAT-16, SES-10, SES-11/EchoStar 105

# Total Number of Successful and Failure Mission Outcomes

List the total number of successful and failure mission outcomes

```
In [28]: %sql SELECT SUM(CASE WHEN "Mission_Outcome" LIKE 'Success%' THEN 1 ELSE 0 END) AS Successful_Missions, SUM(CA
```

```
* sqlite:///my_data1.db  
Done.
```

```
Out[28]: Successful_Missions Failed_Missions
```

100

1

100 missions have been successful and 1  
unsuccessful

# Boosters Carried Maximum Pa yload

```
In [15]: %sql SELECT "Booster_Version", "PAYLOAD_MASS__KG_" FROM SPACEXTABLE WHERE "PAYLOAD_MASS__KG_" = (SELECT MAX("PAYLOAD_MASS__KG_"
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
Out[15]:
```

Booster_Version	PAYLOAD_MASS_KG_
F9 B5 B1048.4	15600.0
F9 B5 B1049.4	15600.0
F9 B5 B1051.3	15600.0
F9 B5 B1056.4	15600.0
F9 B5 B1048.5	15600.0
F9 B5 B1051.4	15600.0
F9 B5 B1049.5	15600.0
F9 B5 B1060.2	15600.0
F9 B5 B1058.3	15600.0
F9 B5 B1051.6	15600.0
F9 B5 B1060.3	15600.0
F9 B5 B1049.7	15600.0

Booster_Version	PAYLOAD_MASS_KG_
F9 B5 B1048.4	15600.0
F9 B5 B1049.4	15600.0
F9 B5 B1051.3	15600.0
F9 B5 B1056.4	15600.0
F9 B5 B1048.5	15600.0
F9 B5 B1051.4	15600.0
F9 B5 B1049.5	15600.0
F9 B5 B1060.2	15600.0
F9 B5 B1058.3	15600.0
F9 B5 B1051.6	15600.0
F9 B5 B1060.3	15600.0
F9 B5 B1049.7	15600.0

**Boosters carrying the  
maximum payload**

# 2015 Launch Records

```
In [36]: %sql SELECT SUBSTR("Date", 4, 2) as Month, "Booster_Version", "Launch_Site", "Landing_Outcome" FROM SPACEXTAB
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
Out[36]:
```

Month	Booster_Version	Launch_Site	Landing_Outcome
10	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
04	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

**Two Boosters in 2015 failing in  
drone ship**

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

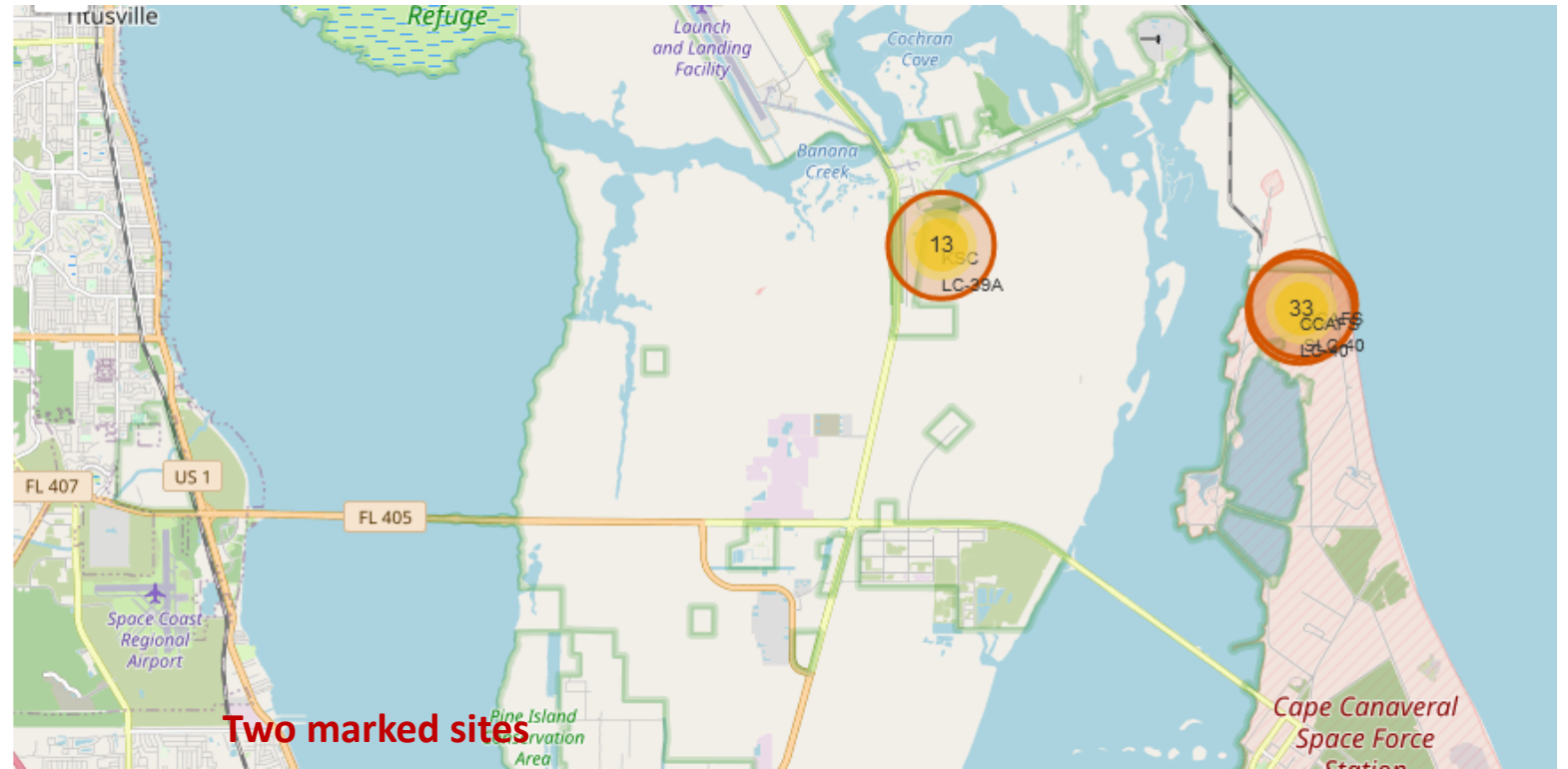
# Launch Sites Proximities Analysis

# Launch Sites in USA

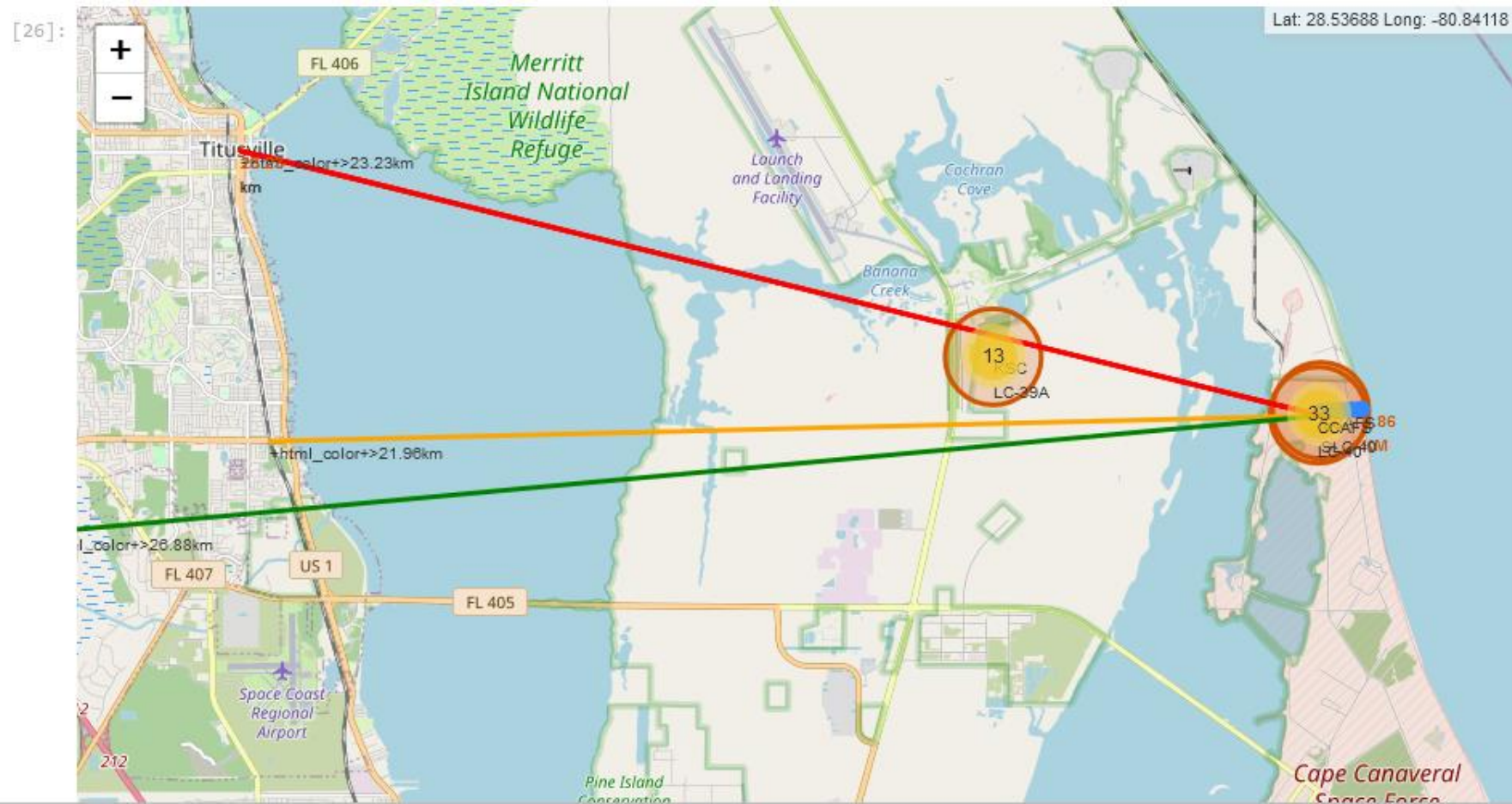




# Marked Sites



# Distances from city, railway, and highway







Section 4

# Build a Dashboard with Plotly Dash

# Successful Launches by site

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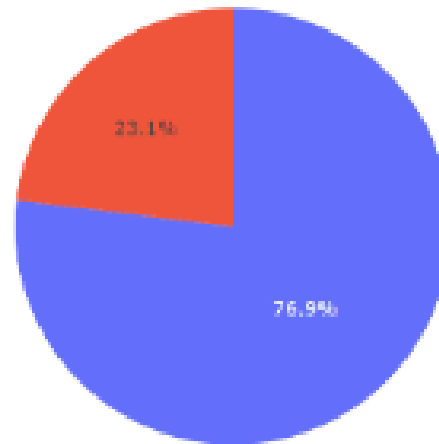
Total Success Launches by Site



KSC LC 39A has the highest successful launches

# KSC LC 39A Success vs Failure

Total Success Launches for Site KSC LC-39A



■ 0  
■ 1

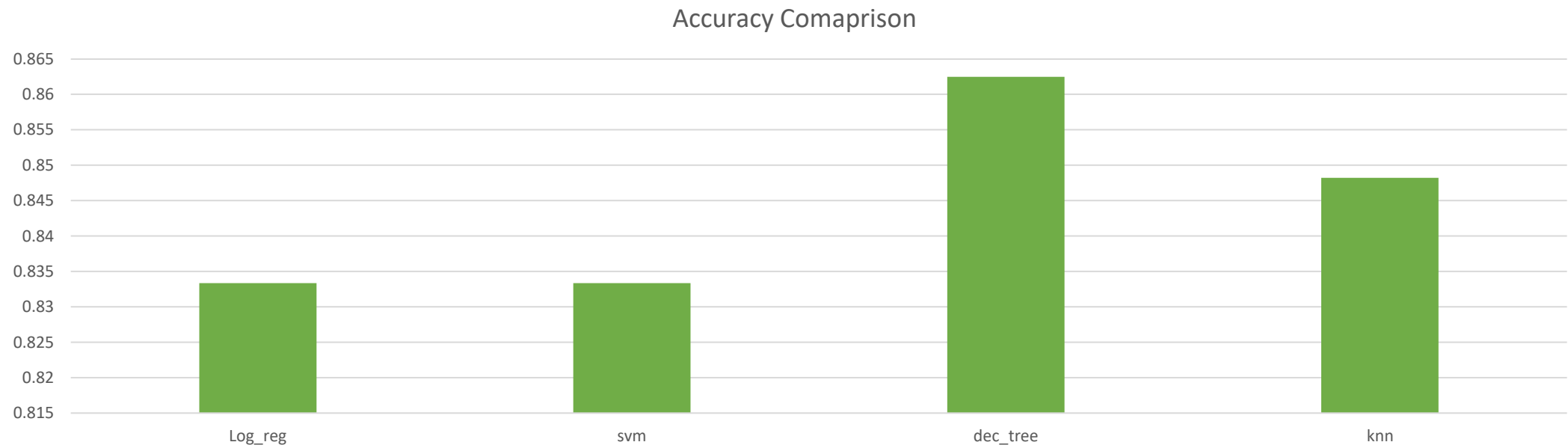
Class 0 = Fail  
Class 1 = Success

KSC LC 39A has had more success than failure

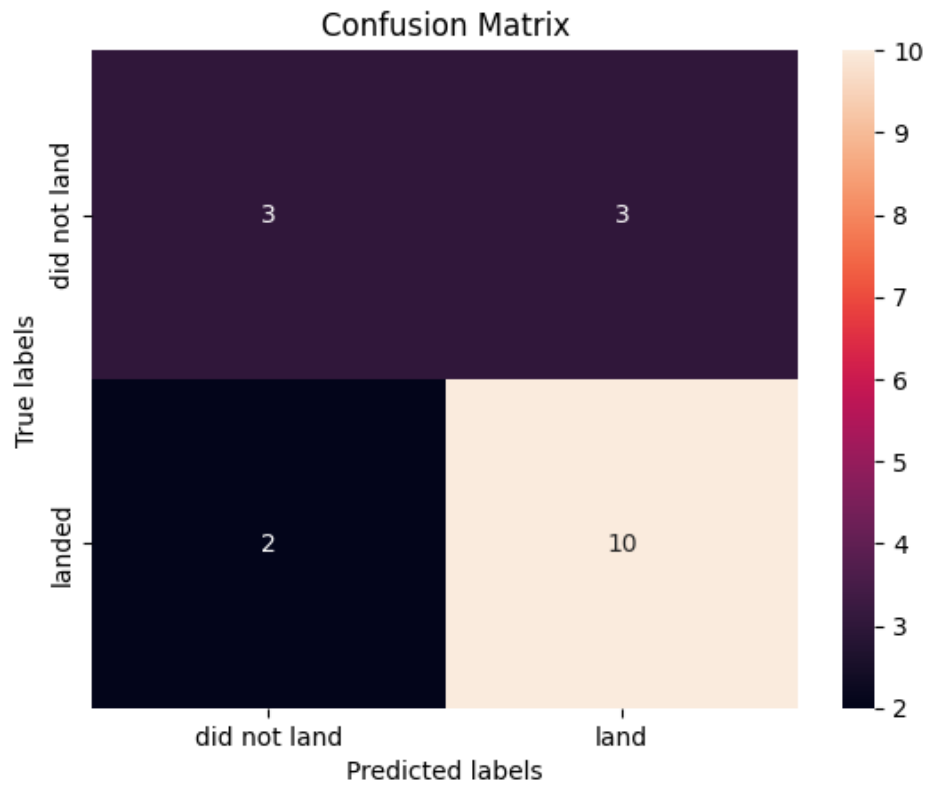
Section 5

# Predictive Analysis (Classification)

# Classification Accuracy



# Confusion Matrix of Decision Tree



The matrix shows that the classifier misclassified 3 failed landings as successful

# Conclusions



Launch success rate has generally increased over the time.



Orbits ES-L1, GEO, HEO, SSO, VLEO had the most success



KSC LC-39A has enjoyed the most success.



The Decision tree classifier is the best classifier in the current context





# Conclusions

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- Launch success rate has generally increased over the time.
- Higher payloads are associated with greater success
- Orbits ES-L1, GEO, HEO, SSO, VLEO had the most success
- KSC LC-39A has enjoyed the most success.
- Sites are near the coastline
- The Decision tree classifier is the best classifier in the current context



Thank you!

