

# Winning Space Race with Data Science

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## **Outline**

- Executive Summary
- Introduction
- Methodology
- Results
- Conclusion

# **Executive Summary**

- Summary of methodologies
  - Data collection using API and wrangling
  - Exploratory Data Analysis including SQL
  - Data Visualization (Graphs, Map with Folium and Dashboard with Plotly Dash)
  - Predictive Analysis
- Summary of all results
  - Exploratory data analysis results
  - Interactive analytics demo in screenshots
  - Predictive analysis results

## Introduction

#### Project background and context

• The objective of this project is to predict if the Falcon 9 first stage will land successfully. SpaceX advertises Falcon 9 rocket launches on its website with a cost of 62 million dollars; other providers cost upward of 165 million dollars each, much of the savings is because SpaceX can reuse the first stage. Therefore, if it is determined that the first stage will land, it would be possible to determine the cost of a launch.

#### Problems you want to find answers

- On which variables does the success of the first stage landing depend?
- What is the relationship between rockets variables and the success or failure of a landing?
- Based on the information available, how good are the predictions of landing success?



# Methodology

#### **Executive Summary**

- Data collection methodology:
  - SpaceX REST API
  - Web scraping from Wiki pages
- Perform data wrangling
  - Clean the original data (null values, irrelevant data)
  - One Hot Encoding
- Perform exploratory data analysis (EDA) using different visualization tools and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Logistic Regression, SVM, Tree Classifier, and KNN models have been built and evaluated (train and test accuracy, score, confusion matrix)

### **Data Collection**

#### Rest SpaceX API

 Data about launches, including information about the rocket used, payload delivered, launch specifications, landing specifications, and landing outcome were obtained from the API (URL api.spacexdata.com/v4/)



- Web scrapping from Wikipedia
  - Data about launches, landings, and payloads were obtained from <a href="https://en.wikipedia.org/wiki/List\_of\_Falcon\ 9\ and Falcon\_Heavy\_launches">https://en.wikipedia.org/wiki/List\_of\_Falcon\ 9\ and Falcon\_Heavy\_launches</a>.



# Data Collection – SpaceX API

#### Link

## 1-Getting Response from API

- spacex\_spacex\_url=https://a pi.spacexdata.com/v4/launc hes/past url=https://api.spacexdata.co m/v4/launches/past
- response = requests.get(spacex\_url)

## 2- Convert Response to JSON into DataFrame

data =
pd.json\_normalize(response.json())

## 3-Cleaning and Transforming Data

- getBoosterVersion(data)
- getLaunchSite(data)
- getPayloadData(data)
- getCoreData(data)

#### 5-Export dataframe to file

data\_falcondata\_falcon9.to\_csv('
dataset\_part\_1.csv', \_1.csv',
index=False)

# 4-Create dictionary with data and store in DataFrame

- launch\_dict = {'FlightNumber': list(data['flight\_number']), 'Date': list(data['date']), 'BoosterVersion':BoosterVersion,'PayloadMass':PayloadMass, 'Orbit':Orbit,'LaunchSite':LaunchSite, 'Outcome':Outcome, 'Flights':Flights, 'GridFins':GridFins, 'Reused':Reused, 'Legs':Legs,'LandingPad':LandingPad,'Block':Block, 'ReusedCount':ReusedCount, 'Serial':Serial,'Longitude': Longitude, 'Latitude': Latitude}
- df = pd.DataFrame(launch\_dict)

# **Data Collection - Scraping**

#### Link

## 1-Getting Response from HTML

- static\_url =
   https://en.wikipedia.org/w/i
   ndex.php?title=List of Falco
   n 9 and Falcon Heavy lau
   nches&oldid=1027686922
- response =requests.get(static\_url)

## 2-Creating a BeatifulSoup Object

soup =
BeautifulSoup(response.text,'html5lib')

#### 3-Finding tables

html\_tables =soup.findAll('table')

# 6-Filling up the dictionary with records from the tables and converting it into a DataFrame

df=pd.DataFrame(launch\_dict)

#### **6-Exporting the DataFrame**

df.to\_csv('spacex\_web\_scraped.csv'
, index=False)

## 5-Creating dictionary to store data

launch\_dict= dict.fromkeys(column\_names)
launch\_dict['Flight No.'] = []

launch dict['Launch site'] = []

launch dict['Payload'] = []

launch\_dict['Payload mass'] = []

launch dict['Orbit'] = []

launch\_dict['Customer'] = []

launch\_dict['Launch outcome'] = []

# Added some new columns

launch\_dict['Version Booster']=[]

launch\_dict['Booster landing']=[]

launch\_dict['Date']=[]

launch\_dict['Time']=[]

## 4- Getting the tables columns

for th in

first\_launch\_table.find\_all('th'):

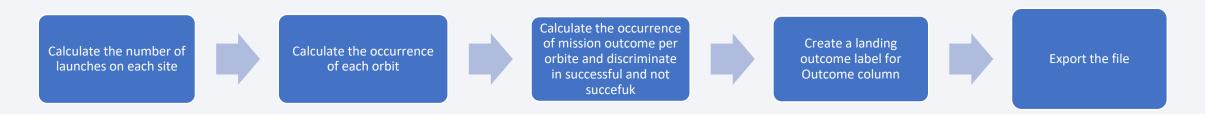
name =

extract\_column\_from\_header(th)

if name is not None and len(name)>0:

column\_names.append(name)

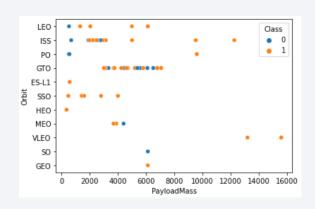
- Data presents different cases where the booster did not land successfully identify which different string labels. The idea was to simplify them into a categorical variable.
  - 1 successful
  - O failure



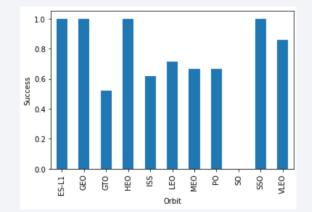
## **EDA** with Data Visualization

Link

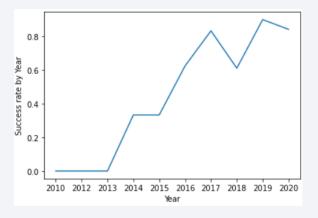
- Scatter plot can suggest relationships (correlation) between variables.
  - PayloadMass vs Flight Numbers
  - Flight Numbers vs Lunch Site
  - Payload vs Launch Site
  - Flight Number vs Orbit type
  - Payload vs Orbit type



- Bar Graph is used to compare data among categories.
  - Success vs Orbit type



- Line plot is used to track changes over short and long periods of time.
  - Launch success vs Years



#### SQL queries performed:

- Display the names of the unique launch sites in the space mission
- Display 5 records where launch sites begin with the string 'CCA'
- Display the total payload mass carried by boosters launched by NASA (CRS)
- 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 ships and have payload mass greater than 4000 but less than 6000
- List the total number of successful and failed mission outcomes
- List the names of the booster\_versions which have carried the maximum payload mass. Use a subquery
- List the records which will display the month names, failure landing\_outcomes in drone ship ,booster versions, launch\_site for the months in the year 2015.
- Rank the count of successful landing\_outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

• Maps objects added to folium map centred on NASA, Texas

- Marker all lunch sites with the corresponding name
- Marker the success (green) and failed (red) lunches for each site
- Line with the distances between a launch site to its proximities (coastline, railways, highways)
- These maps help to understand the problem and the data. It is an easy way to visualize the situation.

# Build a Dashboard with Plotly Dash



- Dashboard has
  - Dropdown which allows selecting one specific launch site or all.
  - Pie plot that shows total success/failure for the selected launch site.
  - Rangeslider that allows choosing a mass payload range.
  - Scatter plot of success vs payload showing their relationship.

# Predictive Analysis (Classification)

Link

#### Data Preparation

- Load preprocessed data
- Normalization data
- Divide into train and test data

#### Model Preparation

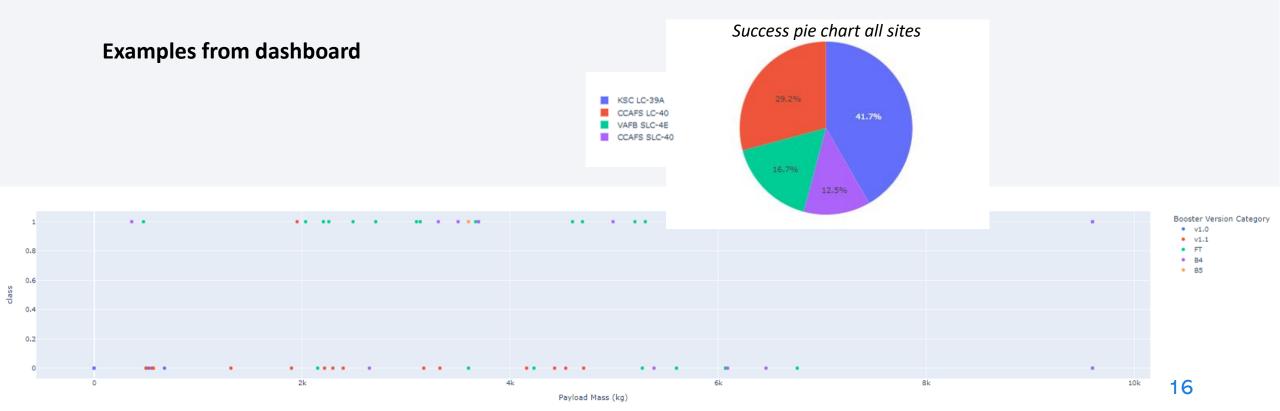
- Selecting 4 models (Logistic Regression, SVM, KNN and Tree Classification)
- Set parameters for each model, and use GridSearchCV
- Training models

#### Model Evaluation and comparison

- Get the best parameters for each model
- Calculate accuracy with test dataset and plot confusion Matrix
- Comparison of the accuracy of the models

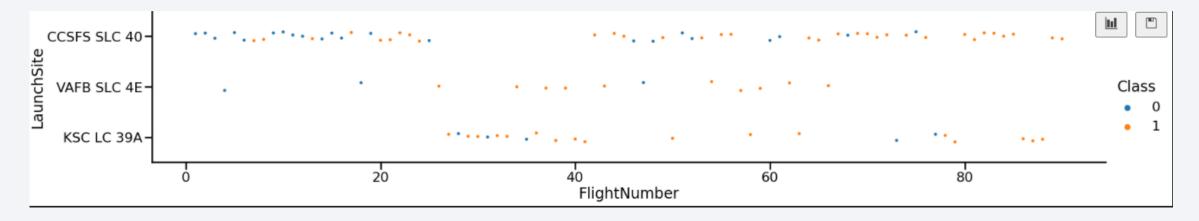
## Results

- All models have similar accuracy (80 %)
- The evolution of success rates show an increase over the years
- Different launch sites have different success rates



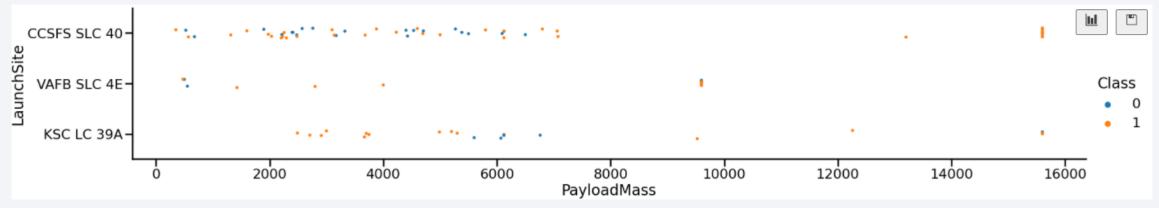


# Flight Number vs. Launch Site



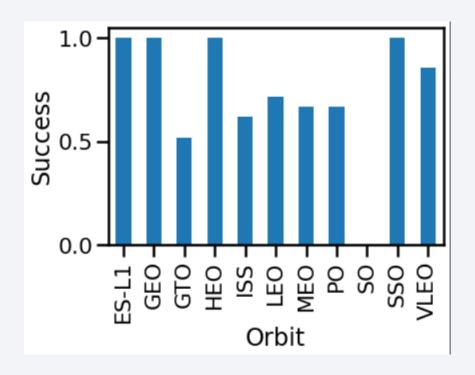
It is observed that the success rate increases, which is more clear in the launch site CCSFC SLC 40

# Payload vs. Launch Site



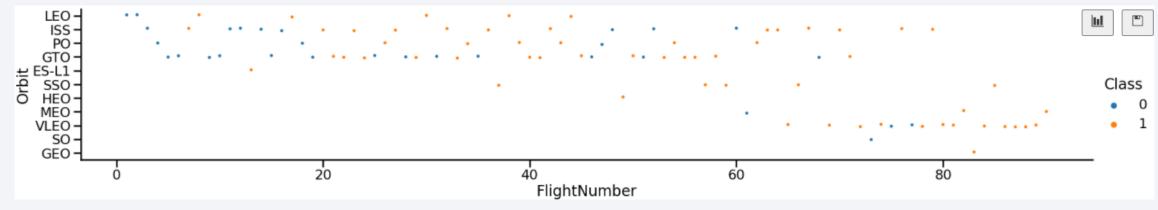
For some launch sites, the heavier payload, the larger the successful landing. However, if the payload is too large, there are some failed landings.

# Success Rate vs. Orbit Type



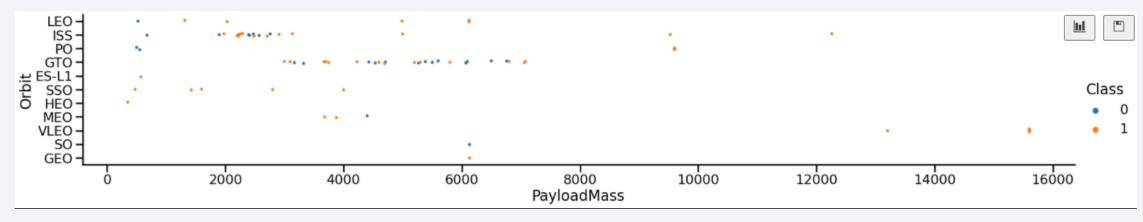
- This graph shows the success rates for different orbit types.
- Some of them (ES-L1, GEO, HEO, and SSO) have the highest values.

# Flight Number vs. Orbit Type



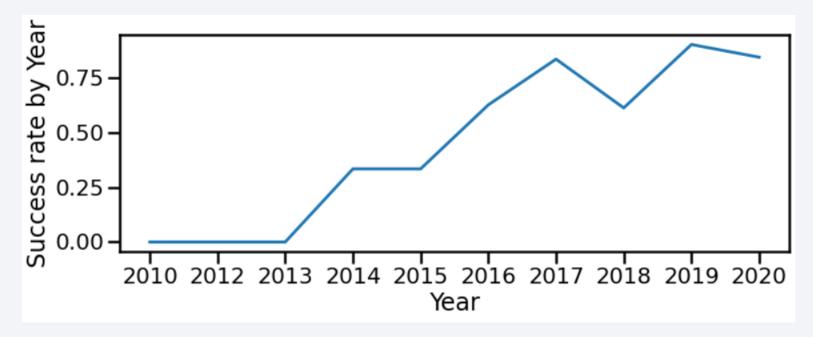
It is observed that the success appears related to the number of flights in the case of LEO orbit, and on the other hand, there seems to be no relationship between both variables in the GTO orbit.

# Payload vs. Orbit Type



The payload has an impact on the success at least for some orbits. Heavy payloads are more successful for Polar, Leo, and ISS orbits

# Launch Success Yearly Trend



This graph shows the evolution of the success, and from 2013 it increases over the years.

### All Launch Site Names

SQL query

```
%sql select DISTINCT "Launch_Site" from SPACEXTBL
```

Results



This query lists all distinct variables in the column "Lauch site" from the table SPACEXTBL.

# Launch Site Names Begin with 'CCA'

SQL query

```
%sql select * from SPACEXTBL where "Launch_Site" like 'CCA%' limit 5
```

Results

| Date           | Time<br>(UTC) | Booster_Version | Launch_Site     | Payload  | PAYLOAD_MASS_KG_ | Orbit        | Customer           | Mission_Outcome | Landing<br>_Outcome    |
|----------------|---------------|-----------------|-----------------|--|------------------|--------------|--------------------|-----------------|------------------------|
| 04-06-<br>2010 | 18:45:00      | F9 v1.0 B0003   | CCAFS LC-<br>40 | Dragon Spacecraft Qualification<br>Unit                          | 0                | LEO          | SpaceX             | Success         | Failure<br>(parachute) |
| 08-12-<br>2010 | 15:43:00      | F9 v1.0 B0004   | CCAFS LC-<br>40 | Dragon demo flight C1, two<br>CubeSats, barrel of Brouere cheese | 0                | LEO<br>(ISS) | nasa (cots)<br>Nro | Success         | Failure<br>(parachute) |
| 22-05-<br>2012 | 07:44:00      | F9 v1.0 B0005   | CCAFS LC-<br>40 | Dragon demo flight C2  | 525              | LEO<br>(ISS) | NASA (COTS)        | Success         | No attempt             |
| 08-10-<br>2012 | 00:35:00      | F9 v1.0 B0006   | CCAFS LC-<br>40 | SpaceX CRS-1   | 500              | LEO<br>(ISS) | NASA (CRS)         | Success         | No attempt             |
| 01-03-<br>2013 | 15:10:00      | F9 v1.0 B0007   | CCAFS LC-<br>40 | SpaceX CRS-2   | 677              | LEO<br>(ISS) | NASA (CRS)         | Success         | No attempt             |

• This query finds firstly all row where column "Launch\_site" starts with CCA and "limit 5" indicates to list only the first 5 lines.

# **Total Payload Mass**

• SQL query

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

Results

```
sum ("PAYLOAD_MASS_KG_")
45596
```

• This query returns the sum of payloads masses filtering where the costumer is NASA (CRS).

# Average Payload Mass by F9 v1.1

SQL query

```
%sql select avg ("PAYLOAD_MASS__KG_") from SPACEXTBL where "Booster_Version" =='F9 v1.1'
```

Results

```
avg ("PAYLOAD_MASS_KG_")
2928.4
```

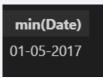
• This query returns the average of payloads masses filtering where the booster version is F9 v1.1

# First Successful Ground Landing Date

SQL query

```
%sql select min(Date) from SPACEXTBL where "Landing _Outcome" == 'Success (ground pad)'
```

Results



• This query returns the oldest successful landing considering only ground landing.

#### Successful Drone Ship Landing with Payload between 4000 and 6000

SQL query

```
%sql select "Booster_Version" from SPACEXTBL where ("Landing _Outcome" == 'Success (drone ship)') AND ("PAYLOAD_MASS__KG_">4000 AND "PAYLOAD_MASS__KG_"<6000)
```

Results

```
sum ("PAYLOAD_MASS_KG_")
45596
```

• This query returns the Bosster Veersion filtering by two columns. Only returns Booster version where Landing Outcomes is Success drone ship, and where Payload masses in between 4000 and 6000 kg.

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

SQL query

```
%sql SELECT(SELECT COUNT("MISSION_OUTCOME") FROM SPACEXTBL WHERE "MISSION_OUTCOME" LIKE '%Success%') as SUCCESS,\
   (SELECT COUNT("MISSION_OUTCOME") FROM SPACEXTBL WHERE "MISSION_OUTCOME" LIKE '%Failure%') as Failure
```

Results



• In this case subqueries were used. The first one counts the success, the second one the failure, using the filter "where".

# **Boosters Carried Maximum Payload**

#### SQL query

```
%sql select "Booster_Version", "PAYLOAD_MASS__KG_" from SPACEXTBL where "PAYLOAD_MASS__KG_" ==(select max("PAYLOAD_MASS__KG_") from SPACEXTBL)
```

#### Results

| Booster_Version | PAYLOAD_MASS_KG_ |
|-----------------|------------------|
| F9 B5 B1048.4   | 15600            |
| F9 B5 B1049.4   | 15600            |
| F9 B5 B1051.3   | 15600            |
| F9 B5 B1056.4   | 15600            |
| F9 B5 B1048.5   | 15600            |
| F9 B5 B1051.4   | 15600            |
| F9 B5 B1049.5   | 15600            |
| F9 B5 B1060.2   | 15600            |
| F9 B5 B1058.3   | 15600            |
| F9 B5 B1051.6   | 15600            |
| F9 B5 B1060.3   | 15600            |
| F9 B5 B1049.7   | 15600            |

In this case, two subqueries were used again. The main is for listing the Boster\_version and Payload masses, and the second one is for filtering the maximum payload.

## 2015 Launch Records

SQL query

```
%sql select substr(Date, 4, 2) as Month ,"Landing _Outcome","Booster_Version","Launch_Site" from SPACEXTBL where substr(Date,7,4)='2015' and "Landing _Outcome" = 'Failure (drone sl
```

Results



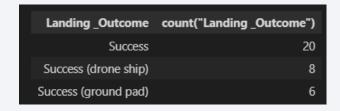
• This query returns the month, landing outcome, booster version, and launch site where the landing was unsuccessful and took place in 2015.

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

SQL query

```
%sql select "Landing _Outcome", count("Landing _Outcome") from SPACEXTBL where ("Date" Between '04-06-2010' and '20-03-2017') and "Landing _Outcome" like '%Success%'\
group by "Landing _Outcome" \
order by count("Landing _Outcome") desc
```

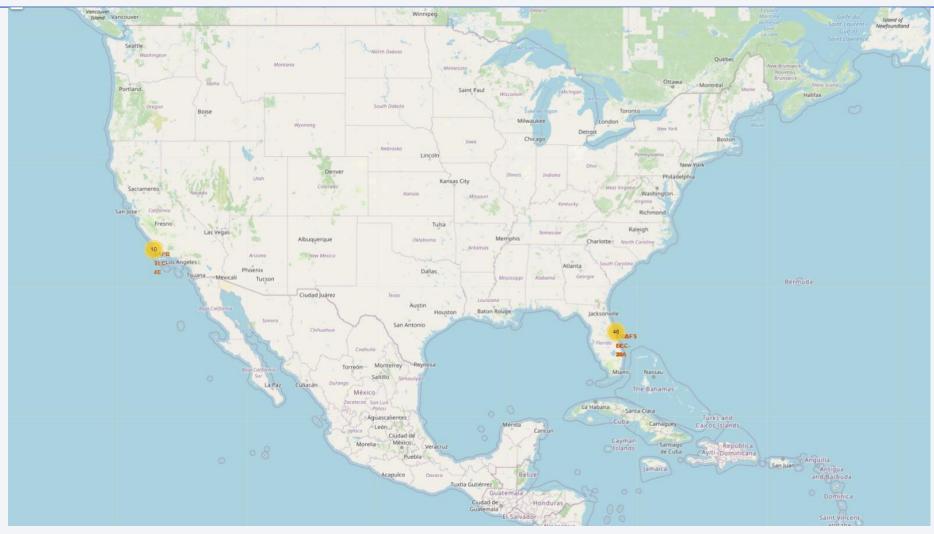
Results



• This query returns the Landing outcomes and their counts for dates between 04-06-2010 and 20-03-2017, also it groups them by the landing outcome, and finally shows them in decreasing order.

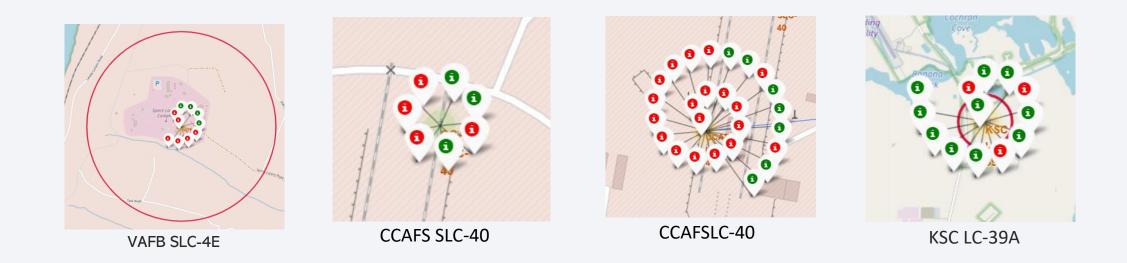


# Folium Map- Stations



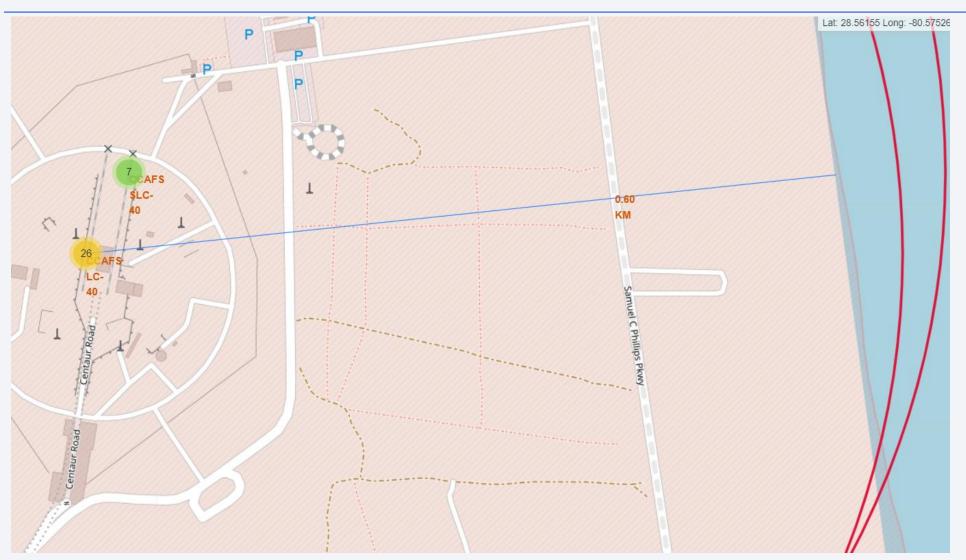
It can be observed the Space X launch sites on both coast.

# Folium Map- Succeed and Failure in color markers



Examples for each launch site were the succeed is indicated in green and failure in red

# Folium Map- Distance



Examples of the distance between CCAFS LC-40 launch site and the coast (0.6 km).

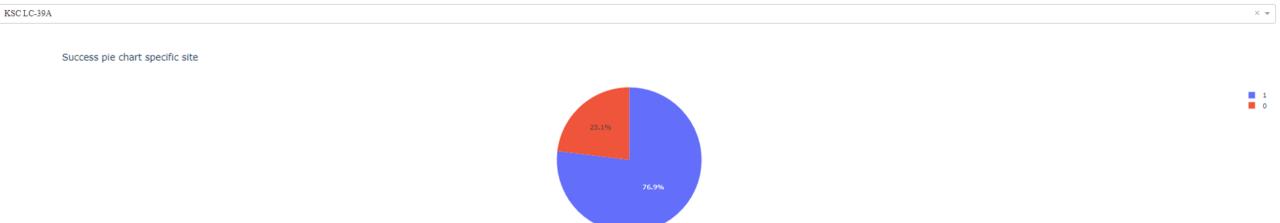


# Dashboard Total success by site



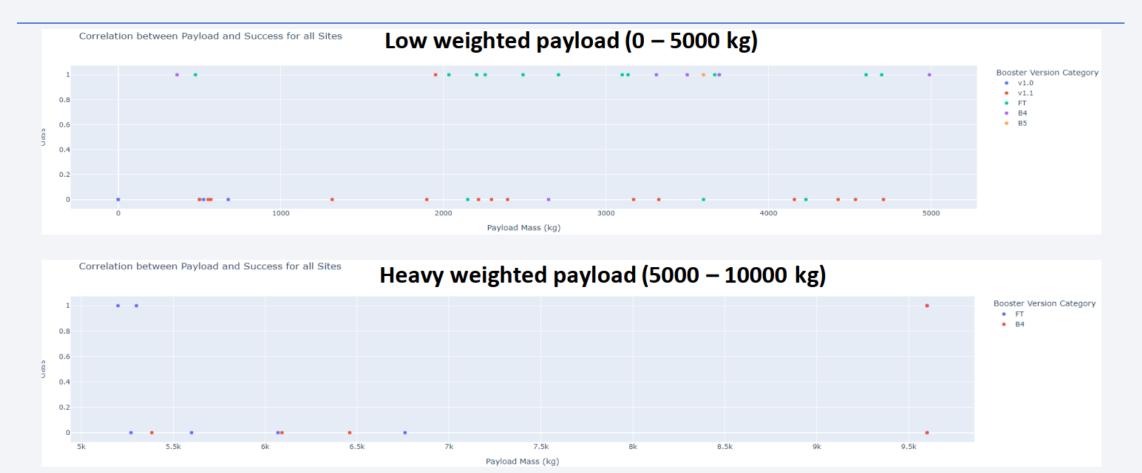
It can be seen that the KSC LC 39A has the best performance.

## Dashboard Total Success for site KSC LC 39A



On KSC LC 39A, the 76.9 % of launches were successful and 23.1 % unsuccessful.

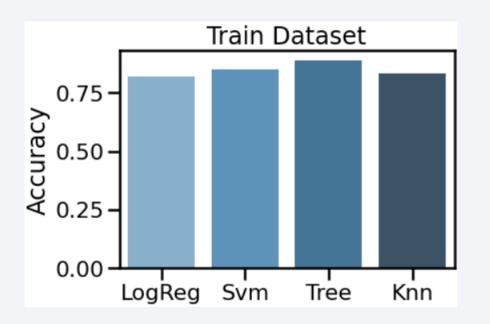
# Dashboard Payload vs Launch outcome for all sites

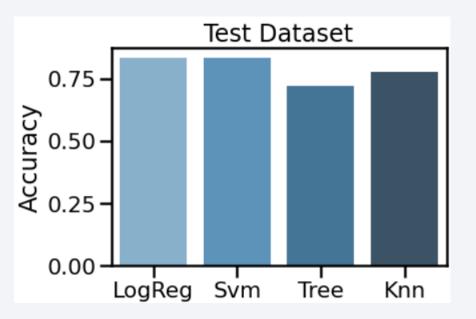


The success rate is better for low weighted payload.



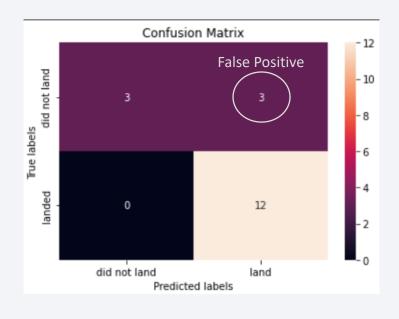
# **Classification Accuracy**





- Based on this data the models have a similar accuracy (train Dataset)
- LogReg and Svm have slightly higher accuracy than Tree and Knn

## **Confusion Matrix**



- This is the confusion Matrix for the SVM model.
- The problem with this model is the false positives.

## **Conclusions**

- From this analysis it can be concluded that the success of landing depends on many factors such as the launch site, the orbit, and the number of previous launches.
- ES-L1, GEO, HEO, and SSO are the orbits with the highest values of success.
- In general terms, the success rate is better for a low-weighted payload.
- An increase in the success rate since 2013 was observed.
- 4 models were used to predict future missions (Logistic Regression, SVM, KNN, and Tree Classifier), and LR and SVM have a slight best accuracy (on test datasets)

