

# Winning Space Race with Data Science

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### Outline

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

## **Executive Summary**

#### Summary of methodologies

- Data collection
- Data wrangling
- Data Analysis using SQL
- EDA with data visualization
- Visual Analytics with Folium
- Predictive analysis (Classification)

#### Summary of all results

- Interactive Visual Analytics
- Exploratory Data Analysis using SQL
- Predictive analysis using a machine learning model

#### Introduction

#### Project background and context

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 you can accurately predict the likelihood of the first stage rocket landing successfully, you can determine the cost of a launch. With the help of your Data Science findings and models, the competing start-up you have been hired by can make more informed bids against SpaceX for a rocket launch.

#### Problems you want to find answers

- Determined the price of each launch.
- Predict if the Falcon 9 first stage will land successfully.



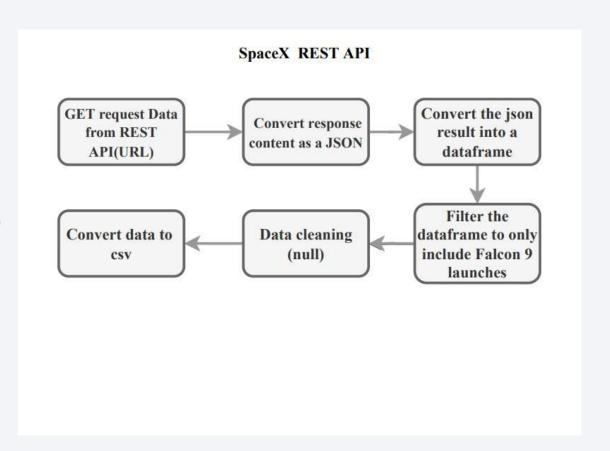
## Methodology

#### **Executive Summary**

- Data collection methodology:
  - Rest API and Web Scraping
- Perform data wrangling
  - Data are clean of null values, and irrelevant columns and transformed into one Hot Encoding to be applied to Machine Learning.
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Use RL, KNN, SVM, DT models have been built and evaluated for the best classifier.

#### **Data Collection**

- •Data that is collected launch data from SpaceX REST API.
- The SpaceX REST API endpoints, or URL, starts with api.spacexdata.com/v4/.



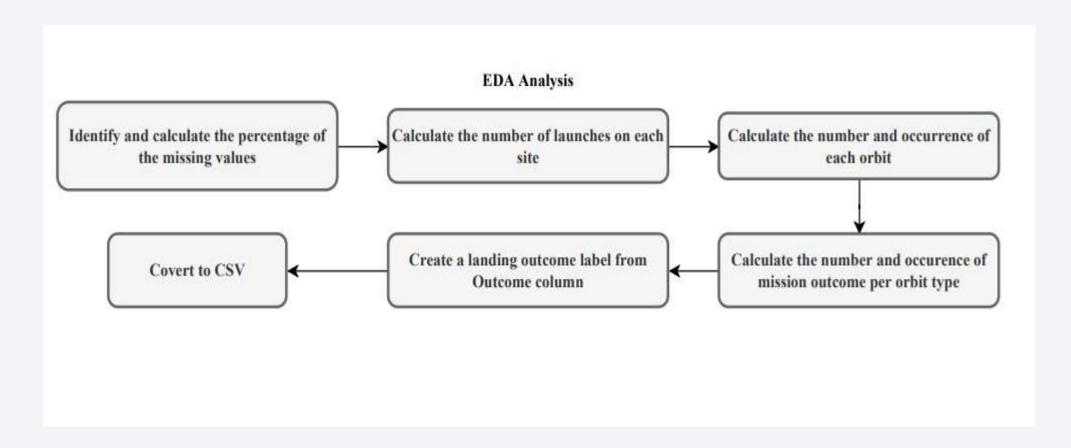
# Data Collection - SpaceX API

8

# Data Collection - Scraping

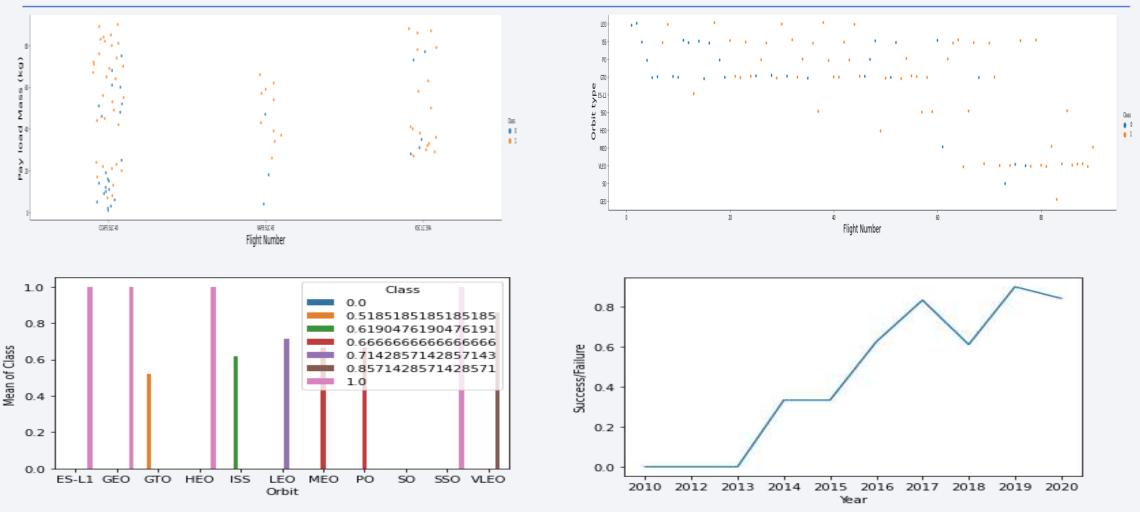
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## **Data Wrangling**



Github: **Data Wrangling** 

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



Github: Exploratory Data Analysis

#### **EDA** with SQL

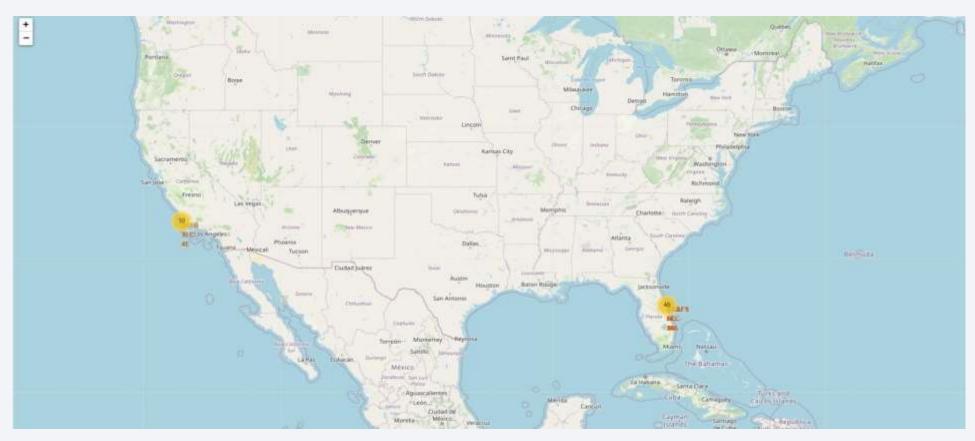
#### SQL queries you performed include:

- 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 acheived
- 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 the names of the booster\_versions which have carried the maximum payload mass. Use a subquery
- List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order

Github: **EDA with SQL** 

## Build an Interactive Map with Folium

By using Marker Cluster, I'm able to add the markers having the same coordinate for launch outcomes to each site.



Github: Folium

# Build a Dashboard with Plotly Dash



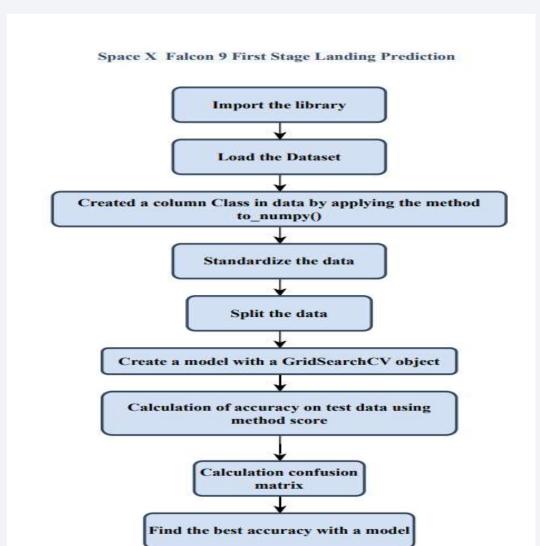


Github: Dash App

## Predictive Analysis (Classification)

- Load the Dataset
- Standardize the data
- Split the data into training and testing data
- Create a model with a GridSearchCV object
- Calculation of accuracy on test data using method score
- Calculation confusion matrix
- Find the best accuracy with a model

**Github**: ML Prediction



#### Results

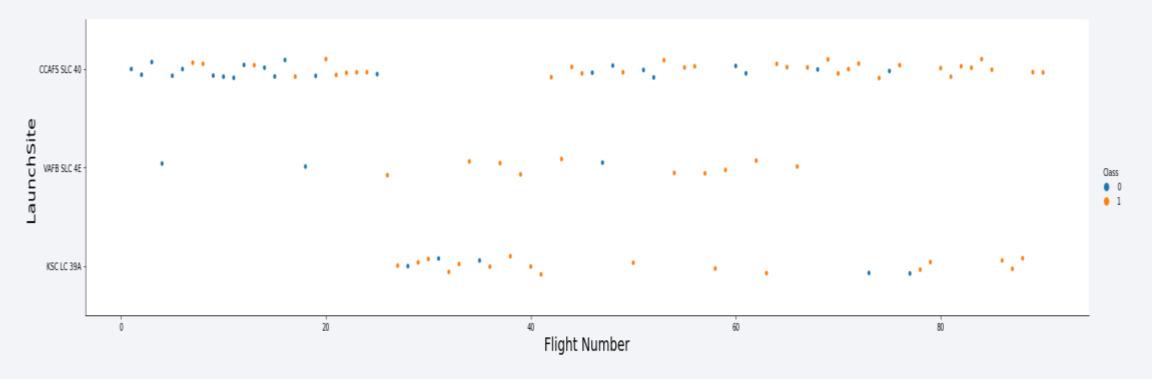
- •After analysis data the CCAFS SLC-40 site and KSCLC-39A site are has most successful launches from all the sites.
- Orbit GEO,HEO,SSOES L1 has the best Success Rate.
- •Decision Tree model are the best in terms of prediction accuracy for this dataset.



## Flight Number vs. Launch Site

#### Visualize the relationship between Flight Number and Launch Site

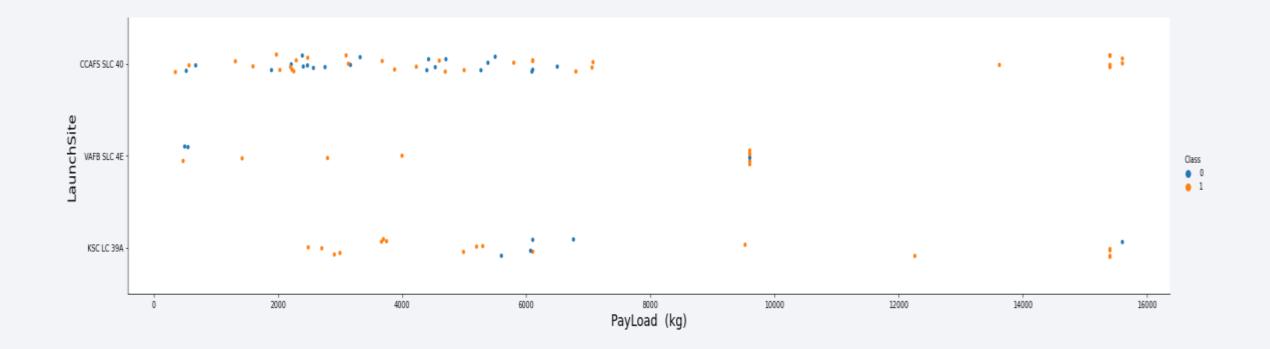
We see that different launch sites have different success rates. But as we increase the number of flights the success rate increase.



## Payload vs. Launch Site

#### Visualize the relationship between Payload and Launch Site

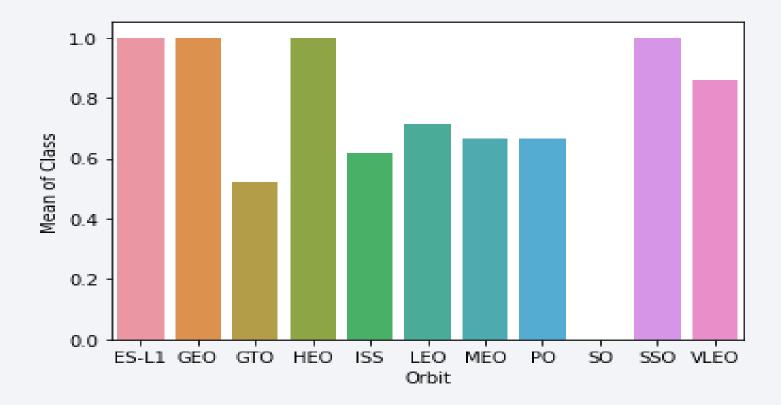
As well, if increase the number of we Pay Load Mass (kg) the success rate increase.



## SSuccess Rate was COntbit Trypee

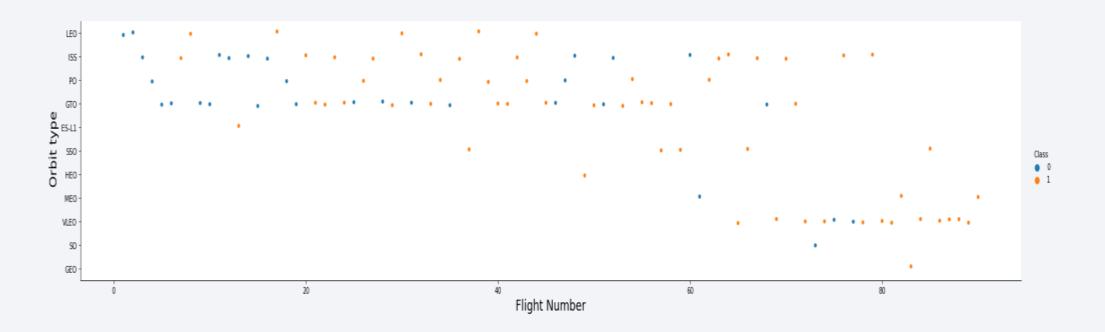
#### Visualize the relationship between success rate of each orbit type

• AS we can see ES-L1, GEO, HEO and SSO have a success rates 100%.



# Flight Number vs. Orbit Type

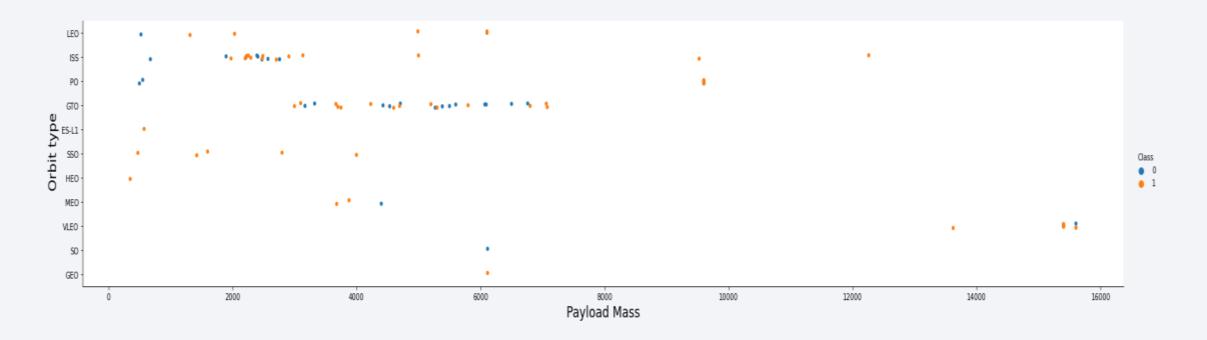
#### Visualize the relationship between FlightNumber and Orbit type



## Payload vs. Orbit Type

#### Visualize the relationship between Payload and Orbit type

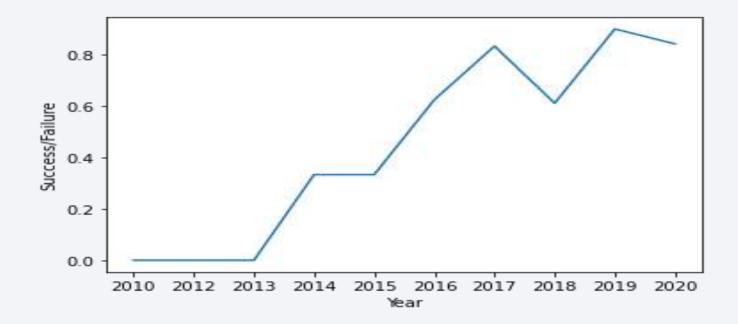
There is a connection between ISSand payload in the range of 2000 to 3000. Also between GTE and Payload at 4000 to 8000.



## Launch Success Yearly Trend

#### Visualize the launch success yearly trend

We can note in the figure that the launch sites begin to increase in success rate from 2013 to approximately 2018, decreases slightly, and then returns to increase with the passage of the year



#### All Launch Site Names

%sql SELECT DISTINCT(launch\_site) FROM Spacex



# Laundh Se Nammes Begin wath COXX

• %sql SELECT \* FROM Spacex WHERE launch\_site LIKE'CCA%' LIMIT 5

landing_outcome	mission_outcome	customer	orbit	payload_masskg_	payload	launch_site	booster_version	timeutc_	DATE
Failure (parachute	Success	SpaceX	LEO	0	Dragon Spacecraft Qualification Unit	CCAFS LC- 40	F9 v1.0 B0003	18:45:00	2010- 04-06
Failure (parachute	Success	NASA (COTS) NRO	LEO (ISS)	0	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	CCAFS LC- 40	F9 v1.0 B0004	15:43:00	2010- 08-12
No attemp	Success	NASA (CRS)	LEO (ISS)	500	SpaceX CRS-1	CCAFS LC- 40	F9 v1.0 B0006	00:35:00	2012- 08-10
No attemp	Success	NASA (CRS)	LEO (ISS)	677	SpaceX CRS-2	CCAFS LC- 40	F9 v1.0 B0007	15:10:00	2013- 01-03
No attemp	Success	SES	GTO	3170	SES-8	CCAFS LC- 40	F9 v1.1	22:41:00	2013- 03-12

## **Total Payload Mass**

%sql SELECT SUM(payload\_mass\_\_kg\_) FROM Spacex WHERE customer='NASA (CRS)'



# AXveragge Payboard Mass by FF99w11.11

 %sql SELECT AVG(payload\_mass\_\_kg\_) FROM Spacex WHERE
 booster\_version='F9 v1.1'



# First SSuccessful Gound Landing Date

 %sql SELECT MIN(DATE) FROM Spacex WHERE landing\_outcome='Success (ground pad)'

**1** 2015-12-22

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

%sql SELECT booster\_version FROM Spacex WHERE landing\_outcome='Success (drone ship)'
 AND payload\_mass\_kg\_BETWEEN 4000 AND 6000

booster\_version

F9 FT B1022

F9 FT B1026

F9 FT B1021.2

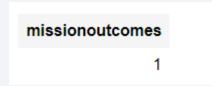
F9 FT B1031.2

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

• %sql SELECT COUNT(MISSION\_OUTCOME) AS missionoutcomes FROM Spacex WHERE mission\_outcome LIKE 'Success%'



%sql SELECT COUNT(MISSION\_OUTCOME) AS missionoutcomes FROM Spacex WHERE mission\_outcome LIKE 'Failure



## **Boosters Carried Maximum Payload**

%sql SELECT booster\_version AS Maxboosterversion FROM Spacex WHERE payload\_mass\_\_kg\_=(SELECT MAX(payload\_mass\_\_kg\_) FROM Spacex)

maxboosterversion
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

#### 2015 Launch Records

%sql SELECT landing\_\_outcome,booster\_version,launch\_site,DATE FROM Spacex WHERE landing\_\_outcome='Failure (drone ship)' AND EXTRACT(YEAR FROM DATE)='2015'

landing_outcome	booster_version	launch_site	DATE
Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40	2015-10-01
Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40	2015-04-14

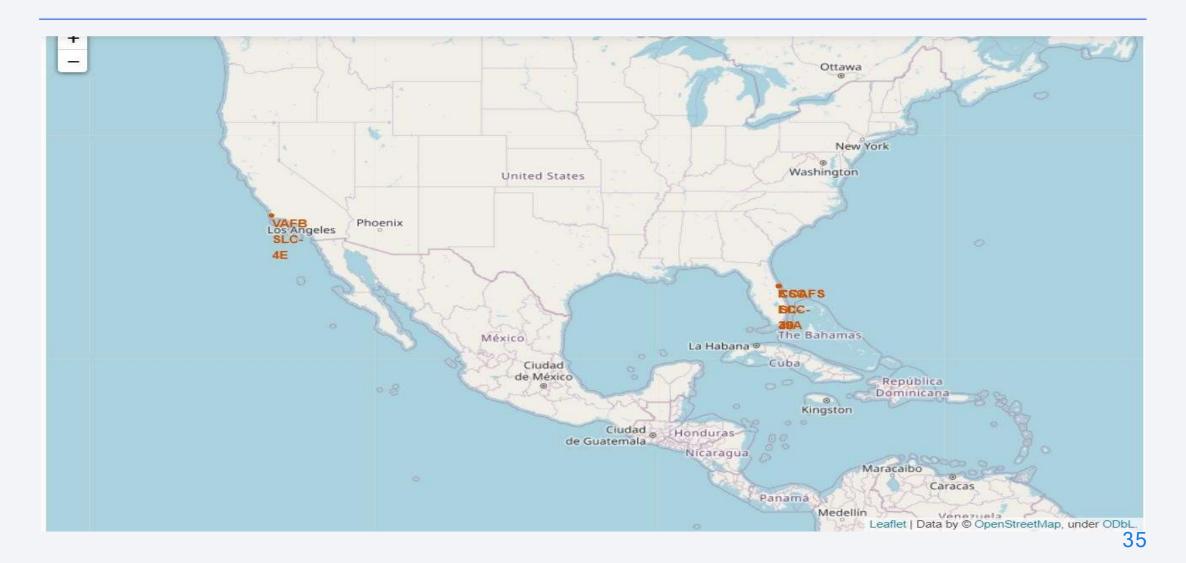
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

%sql SELECT landing\_\_outcome , COUNT(landing\_\_outcome) FROM Spacex WHERE DATE BETWEEN
 '2010-06-04' AND '2017-03-20' GROUP BY landing\_\_outcome ORDER BY COUNT(landing\_\_outcome)
 DESC

landing_outcome	2
No attempt	10
Success (drone ship)	6
Failure (drone ship)	5
Success (ground pad)	5
Controlled (ocean)	3
Uncontrolled (ocean)	2
Failure (parachute)	1
Precluded (drone ship)	1

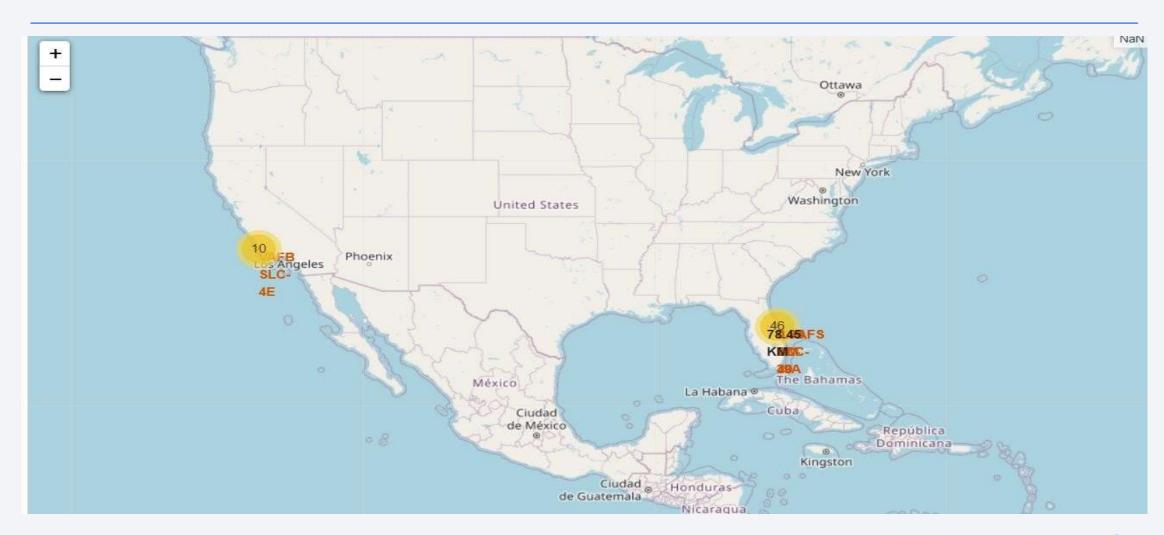


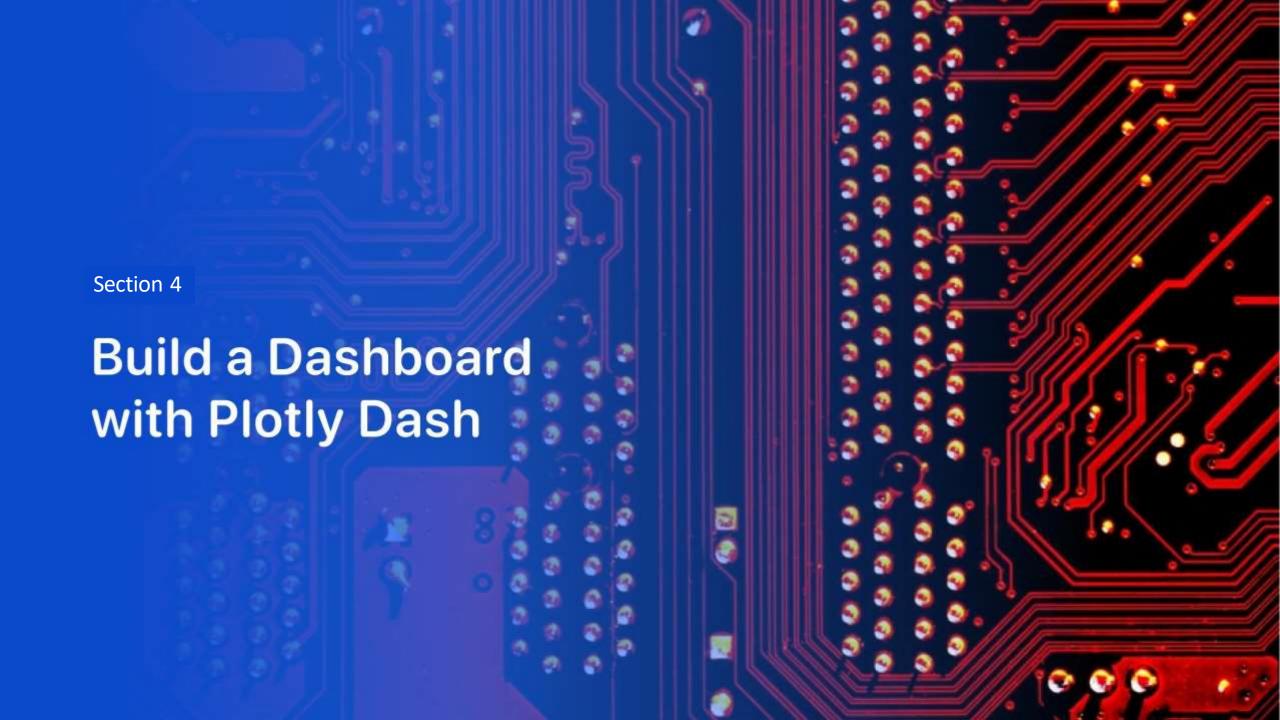
# All launch sites on the site map



# launch sites by using Marker Cluster

#### Marker distance between the coastline point and the launch site

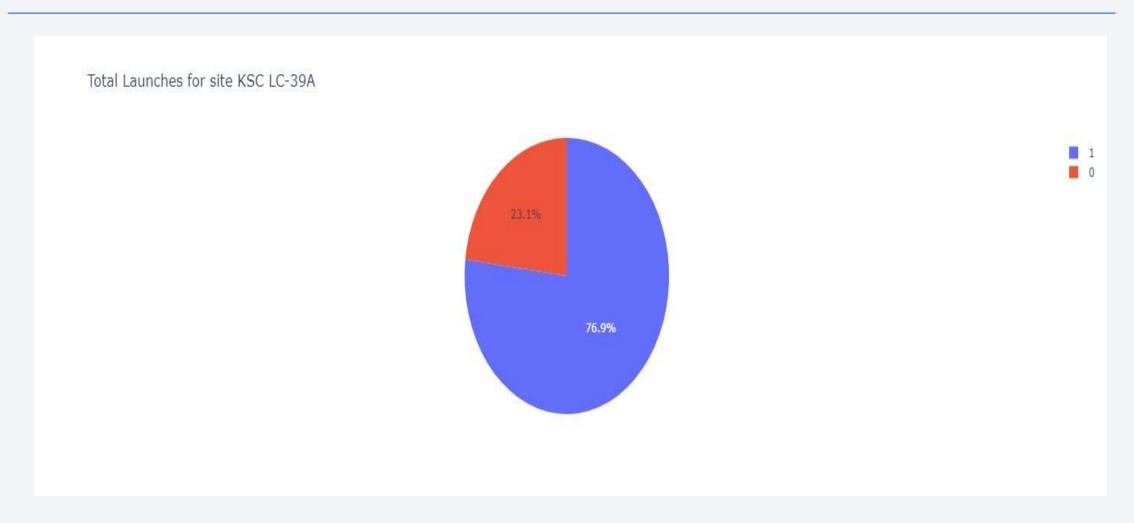




### **Total Success Launches**



# Launches for Hight site Score



# Payload vs. All Launch

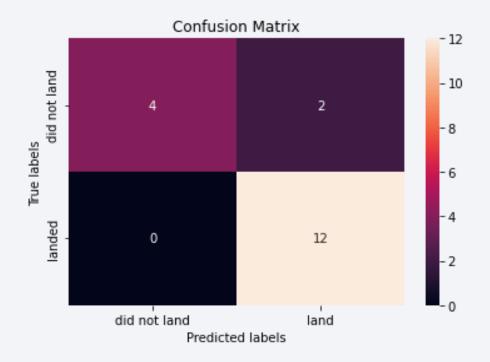




## Classification Accuracy

As we can see Decision Tree has the highest accuracy with almost 0.89, then comes the remaining models with almost the same accuracy of 0.84.

## **Confusion Matrix**



Measure	Derivations	Result
Precision	PPV = TP / (TP + FP)	0.67
Accuracy	ACC = (TP + TN) / (P + N)	0.89
F1 Score	F1 = 2TP / (2TP + FP + FN)	0.80

#### Conclusions

- After analysis data the CCAFS SLC-40 site and KSC LC-39A site are has most successful launches from all the sites.
- Orbit GEO, HEO, SSOES L1 has the best Success Rate.
- The payload of O kg to 5000 kg was more diverse than 6000 kg to 10000
- The Decision Tree model is the best in terms of prediction accuracy for this dataset.

#### References

- Confusion Matrix : https://onlineconfusionmatrix.com/
- Matplotlib Bar Plot : <a href="https://www.tutorialspoint.com/matplotlib/matplotlib">https://www.tutorialspoint.com/matplotlib/matplotlib</a> bar plot.htm
- https://towardsdatascience.com/7-points-to-create-better-histograms-with-seaborn-5fb542763169
- https://www.researchgate.net/figure/Bar-chart-showing-the-performance-evaluation-in-our-data-loading-tests fig4 268150621

