

# Winning Space Race with Data Science

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#### **Outline**

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

#### **Executive Summary**

- Summary of methodologies
  - API Data Collection
  - Data Collection with Web Scaping
  - Exploratory Data Wrangling
  - Exploratory Data Analysis with SQL
  - Data Analysis with Data Visualization
  - Interactive Visual Analytics with Folium lab
  - Machine Learning Prediction
- Summary of all results
  - Data collection and data analysis results
  - O Visualize the data and extract meaningful patterns using data visualization
  - Machine learning model predictions

#### Introduction

• 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. If we can determine if the first stage will land, we can determine the cost of a launch. This information can be used if an alternate company wants to bid against SpaceX for a rocket launch.

• In a role of a data scientist for a new rocket company, Space Y, aiming to compete with SpaceX. The task involves determining launch prices, gathering SpaceX information, and predicting the reuse of the first stage using machine learning and public data,



# Methodology

#### **Executive Summary**

- Data collection methodology:
  - 2 sources used for Data collection:
  - SpaceX REST API (<a href="https://api.spacexdata.com/v4/rockets/">https://api.spacexdata.com/v4/rockets/</a>)
- Perform data wrangling
  - convert outcomes into training labels with 1 means the booster successfully landed 0 means it was unsuccessful.

# Methodology

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Collected data was standardized, Split into training data and test data. Then found the best Hyperparameter for SVM, Classification Trees and Logistic Regression and found which method performs the best

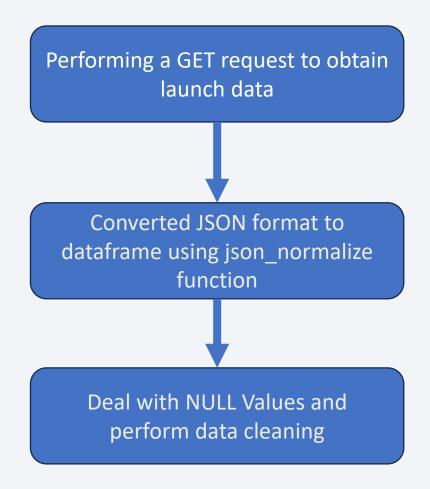
#### **Data Collection**

Data was collected using 2 sources: SpaceX REST API
 (<a href="https://api.spacexdata.com/v4/rockets/">https://api.spacexdata.com/v4/rockets/</a>) and wikipedia page
 (<a href="https://en.wikipedia.org/wiki/List">https://en.wikipedia.org/wiki/List of Falcon\ 9\ and Falcon Heavy launches</a>)

# Data Collection – SpaceX API

 Used SpaceX REST API to gather launch data, including rocket details, payload information, launch and landing specifications, and outcomes.

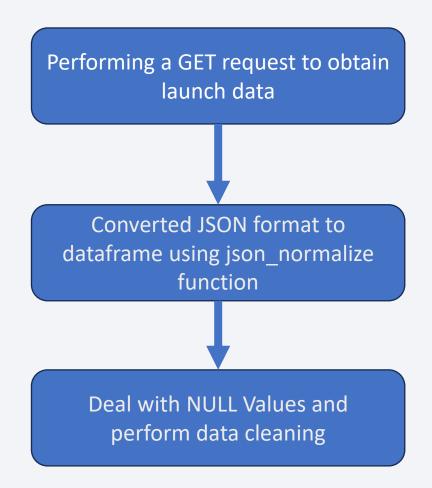
https://github.com/lsmail-Sadaqat/Applied-Data-Science-Capstone/blob/2bd2e2d33fc8616250bcb4548c1fb72a170428d7 /Data%20Collection%20API%20Lab.ipynb



#### **Data Collection - Scraping**

- Used Wikipedia pages for Falcon 9 launch records.
- Flowcharts present web scraping process

https://github.com/lsmail-Sadaqat/Applied-Data-Science-Capstone/blob/2bd2e2d33fc8616250bcb4548c1fb72a 170428d7/Data%20Collection%20with%20Web%20Sc raping.ipynb



# **Data Wrangling**

- Perform exploratory data analysis
  - Calculate the number of launches on each site
  - Calculate the number and occurrence of each orbit
  - Calculate the number and occurrence of mission outcome per orbit type
- Convert Landing outcomes into training labels with 1 means the booster successfully landed O means it was unsuccessful.

**Exploratory Data Analysis** Calculate number of occurrence of different data **Training Labels** 

https://github.com/lsmail-Sadaqat/Applied-Data-Science-Capstone/blob/2bd2e2d33fc8616250bcb4548c1fb72a170428d7/Data%20Wrangling.ipynb

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

- Used scatter graph to Visualize and find relationship between:
  - Flight Number and Payload Mass
  - Flight Number and Launch Site
  - Payload and Launch Site
  - Flight Number and Orbit type
  - Payload and Orbit type
- Created a bar chart to visualize the relationship between success rate of each orbit type and a line chart to visualize the launch success yearly trend
- Used features engineering for success prediction in the future module.

#### **EDA** with SQL

#### Performed the following SQL queries:

- o 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 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 records which will 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 (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order.

#### Build an Interactive Map with Folium

- Markers, circles, lines created and added to a folium map:
  - Mark all launch sites on a map
  - Mark the success/failed launches for each site on the map
  - Calculate the distances between a launch site to its proximities

#### Build a Dashboard with Plotly Dash

- Plots, graphs and interactions added to a dashboard:
  - Pie charts Percentage of launches by certain launch site
  - Scatter graphs relationship between Outcome and Payload Mass

# Predictive Analysis (Classification)

- Four classification models prepared and compared:
  - Logistic Regression
  - Support Vector machines
  - Decision Tree Classifier
  - K-nearest neighbors
- The flowchart presents the model development process.

Preprocessing: standardize and splitting data into Train and test Train the models and perform Grid Search Find the hyperparameters and determine the model with the best accuracy

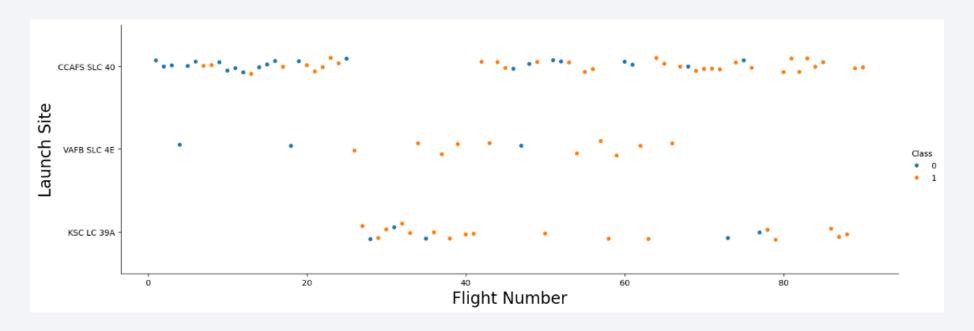
https://github.com/lsmail-Sadaqat/Applied-Data-Science-Capstone/blob/2bd2e2d33fc8616250bcb4548c1fb72a170428d7/Machine%20Learning%20Prediction.ipynb

#### Results

- Exploratory data analysis results
- Interactive analytics demo in screenshots
- Predictive analysis results

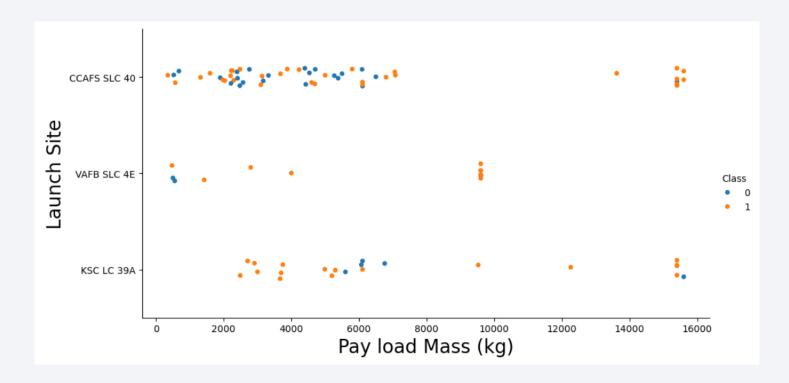


#### Flight Number vs. Launch Site



- The Scatter graph shows that the most successful Launch site is the CCAFS SLC 40
- The Scatter graph also shows that the larger the flight number the greater the success rate.

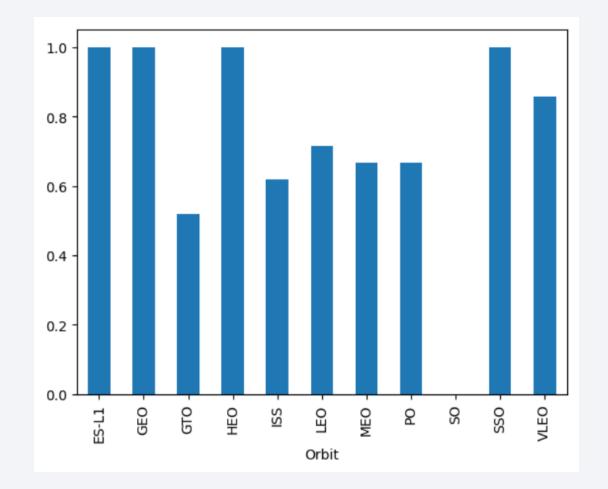
#### Payload vs. Launch Site



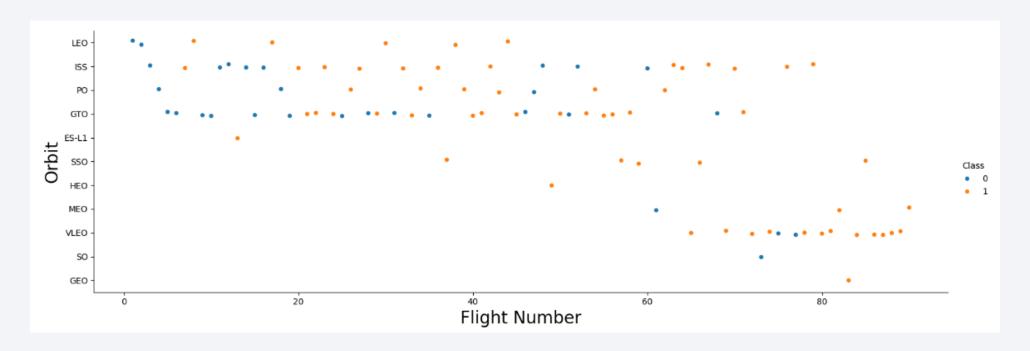
- Payload above 8000 kg the probability of success rate increases
- The scatter graph suggests that payload above 10000kgs is only possible in CCAFS SLC 40 and KSC LC 39A

# Success Rate vs. Orbit Type

- The graph shows the probability of success full landing outcome depending on the orbit type
- Orbits with 100% success rate:
  - o ES-L1
  - o GWO
  - o HEO
  - o SSO

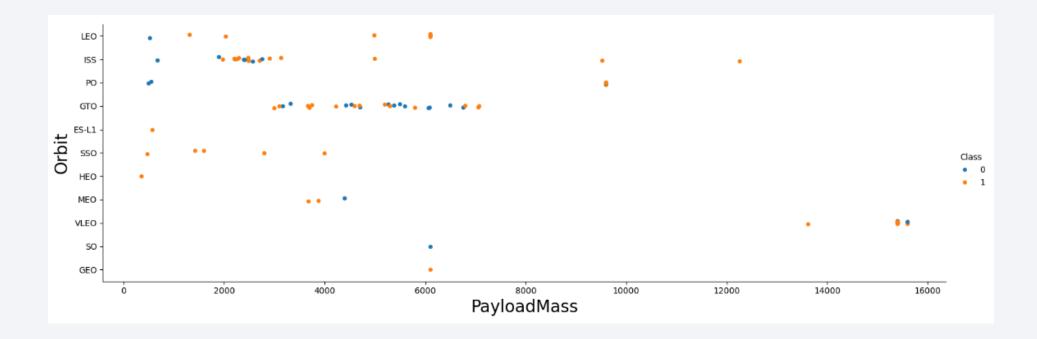


# Flight Number vs. Orbit Type



- The scatter graph suggest that most of the Flights were done in LEO, ISS,
   PO and GTO
- Success rate is also increased with the increase in Flight Number

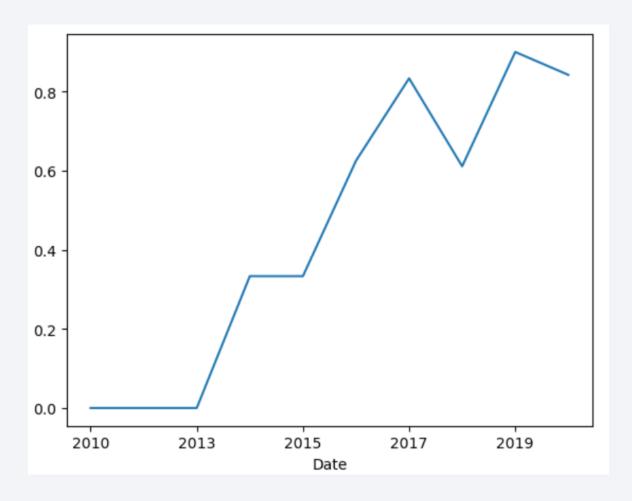
# Payload vs. Orbit Type



• There seem to be no relationship between payload and orbit Type

# Launch Success Yearly Trend

- The line chart shows a clear increase in yearly average success rate starting from 2013 till 2017.
- After 2017 there has been few decrease in the success rate, but overall, as time passes success rate increases.



#### All Launch Site Names

• There are four unique launch sites:

# Launch\_Site CCAFS LC-40 VAFB SLC-4E KSC LC-39A CCAFS SLC-40

They were obtained by using the DISTINCT keyword

# Launch Site Names Begin with 'CCA'

• Five records where launch sites begin with `CCA`:

Date	Time (UTC)	Booster_Ver sion	Launch_Site	Payload	PAYLOAD_ MASSKG_	Orbit	Customer	Mission_Ou tcome	Landing_Ou tcome
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-	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

# **Total Payload Mass**

Total payload carried by boosters from NASA:

Total\_Payload 111268

• This was obtained by summing all payload mass whose codes contain 'CRS'

# Average Payload Mass by F9 v1.1

Average payload mass carried by booster version F9 v1.1:

Average\_Payload\_Mass 2928.4

This was obtained by averaging all payload mass where booster version is F9 v1.1

# First Successful Ground Landing Date

• Date of the first successful landing outcome on ground pad:

min(Date)

2015-12-22

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

 Names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 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

• Total number of successful and failure mission outcomes:

Mission_Outcome	count(*)
Failure (in flight)	1
Success	98
Success	1
Success (payload status unclear)	1

# **Boosters Carried Maximum Payload**

• Names of the booster which have carried the maximum payload mass:

#### 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

#### 2015 Launch Records

• Failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015

Months	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

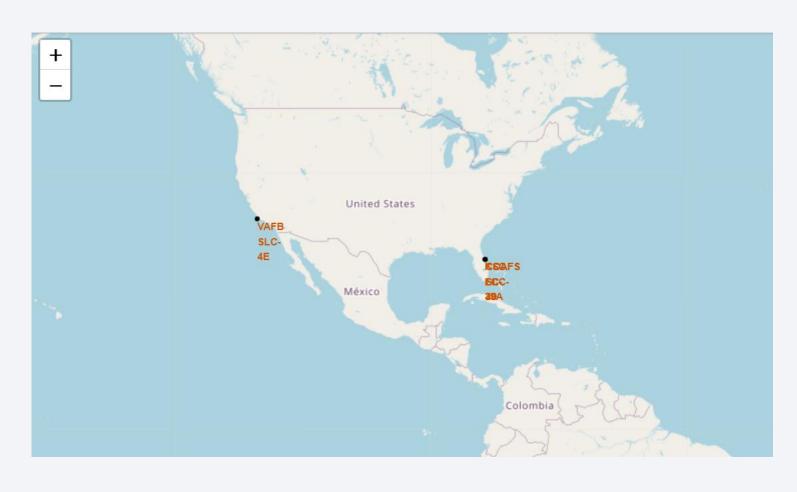
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

• Count of landing outcomes between the date 2010-06-04 and 2017-03-20, in ranked in descending order

Landing_Outcome	count
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



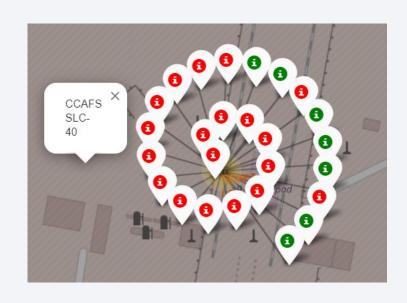
#### **Launch Sites**



 From the map we can see that all launch sites are in proximity to the Equator line

 We can also see all launch sites are in very close proximity to the coast

### launch outcomes



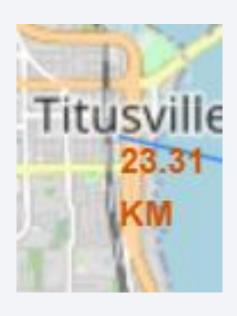


 The color-labeled markers identify which launch sites have relatively high success rates.

 Green color-labeled markers indicates successful and red colorlabeled markers indicates falure

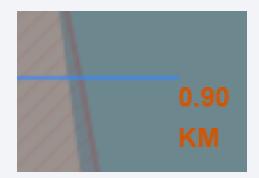
#### launch site distance calculated from landmarks





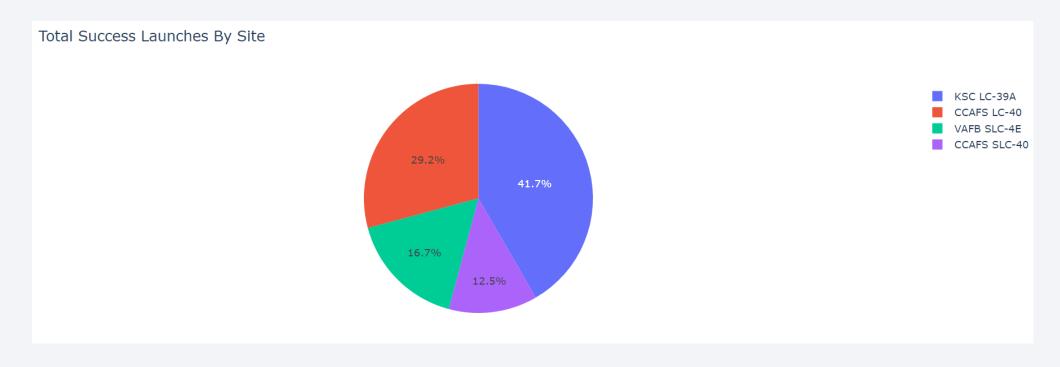


- The launch sites are in close proximity to railways, highways and coastline?
- The launch sites keeps certain distance away from cities



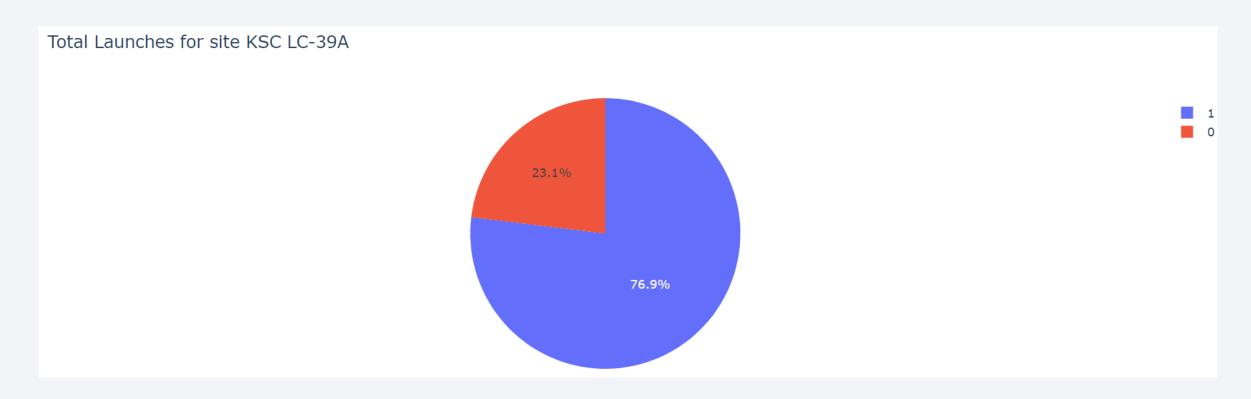


### launch success for all sites



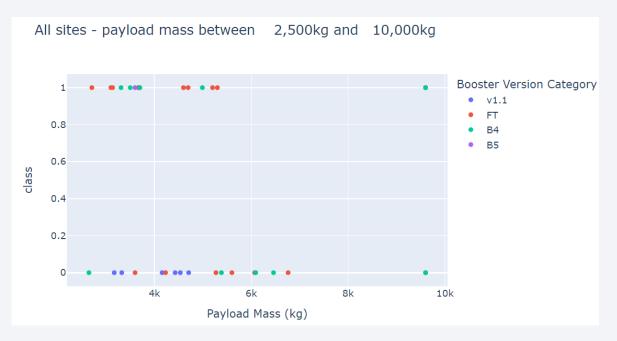
• KSC LC-39A launch site has a high launch success rate

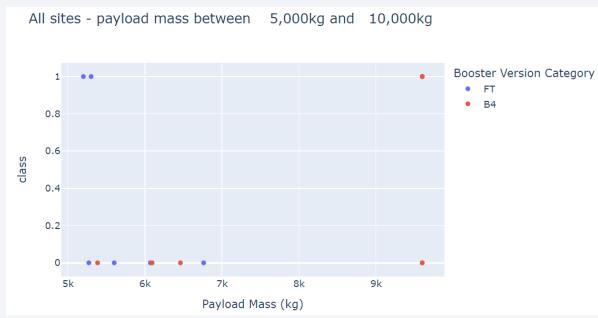
# launch site with highest launch success



• 76.9% of launches done in KSC LC-39A are successful

### Payload vs. Launch Outcome





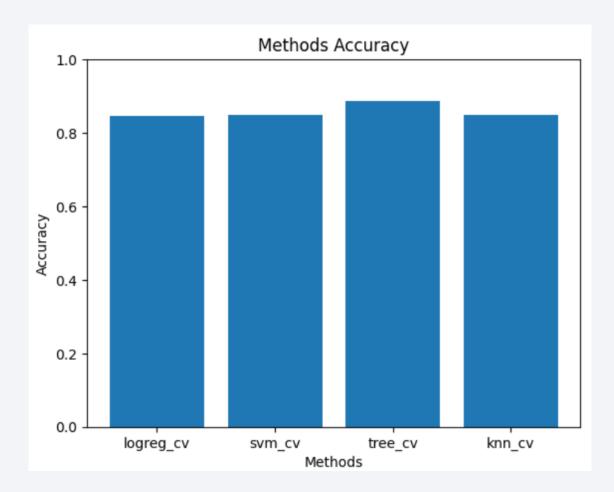
• As the payload mass increase, successful launch decreases



### Classification Accuracy

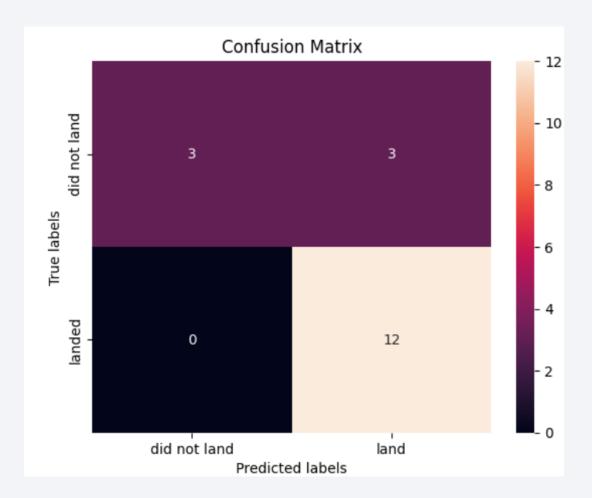
 The bar chart visualizes the built model accuracy for all built classification models

 Decision tree classifier was had the highest classification accuracy



### **Confusion Matrix**

Confusion matrix of the best performing model shows that the model is accurate by showing big numbers in the true positive and true negative



#### Conclusions

- Different sources were used for the analysis
- KSC LC-39A launch site has a high launch success rate
- As the payload mass increase, successful launch decreases
- Decision tree classifier is the best model for this dataset

## **Appendix**

• Include any relevant assets like Python code snippets, SQL queries, charts, Notebook outputs, or data sets that you may have created during this project

