

# Winning Space Race with Data Science

<Name> <Date>



## Outline

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

#### **Executive Summary**

- Collected data from public SpaceX API and SpaceX Wikipedia page. Created labels column 'class' which classifies successful landings. Explored data using SQL, visualization, folium maps, and dashboards. Gathered relevant columns to be used as features. Changed all categorical variables to binary using one hot encoding. Standardized data and used GridSearchCV to find best parameters for machine learning models. Visualize accuracy score of all models.
- Four machine learning models were produced: Logistic Regression, Support Vector Machine, Decision Tree Classifier, and K Nearest Neighbors. All produced similar results with accuracy rate of about 83.33%. All models over predicted successful landings. More data is needed for better model determination and accuracy.

#### Introduction

#### **Background:**

- Commercial Space Age is Here
- Space X has best pricing (\$62 million vs. \$165 million USD)
- Largely due to ability to recover part of rocket (Stage 1)
- Space Y wants to compete with Space X

#### **Problem:**

Space Y tasks us to train a machine learning model to predict if the first stage will land



## Methodology

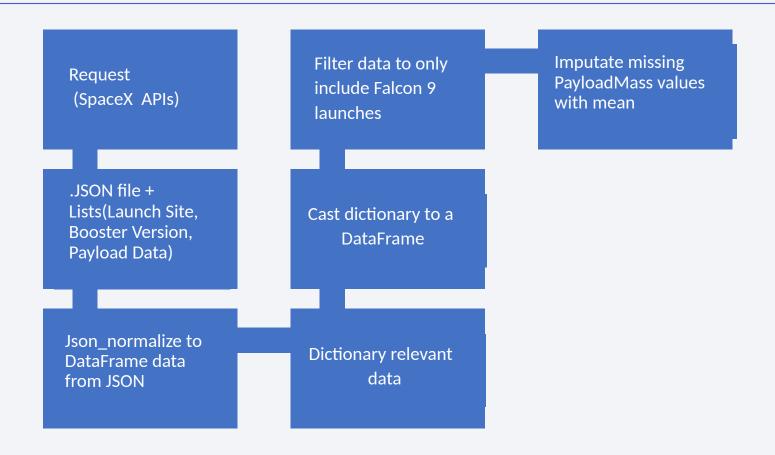
#### **Executive Summary**

- Data collection methodology:
  - Combined data from SpaceX public API and SpaceX Wikipedia page
- Perform data wrangling
  - Classifying true landings as successful and unsuccessful otherwise
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Tuned models using GridSearchCV

#### **Data Collection**

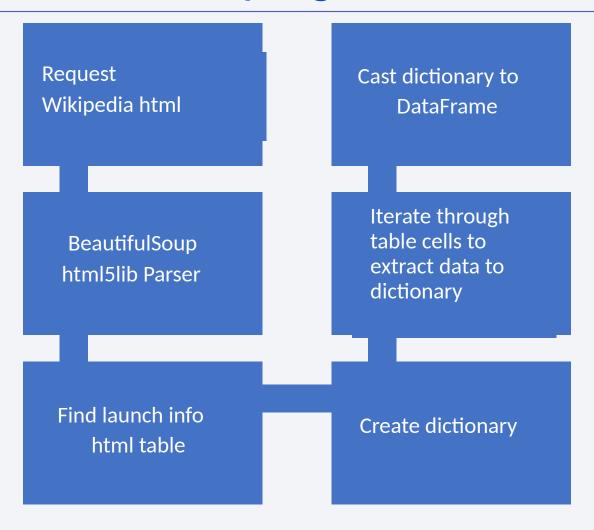
- Data collection process involved a combination of API requests from Space X public API and web scraping data from a table in Space X's Wikipedia entry.
- The next slide will show the flowchart of data collection from API and the one after will show the flowchart of data collection from webscraping.
- Space X API Data Columns:
  - FlightNumber, Date, BoosterVersion, PayloadMass, Orbit, LaunchSite, Outcome, Flights,
     GridFins,
  - o Reused, Legs, LandingPad, Block, ReusedCount, Serial, Longitude, Latitude
- Wikipedia Webscrape Data Columns:
  - Flight No., Launch site, Payload, PayloadMass, Orbit, Customer, Launch outcome,
     Version Booster, Booster landing, Date, Time

# Data Collection – SpaceX API



Github: Data Collection API

# Data Collection - Scraping



Github: Web Scraping

# **Data Wrangling**

- Create a training label with landing outcomes where successful = 1 & failure = 0.
- Outcome column has two components: 'Mission Outcome' 'Landing Location'
- New training label column 'class' with a value of 1 if 'Mission Outcome' is True and 0 otherwise. Value Mapping:
- True ASDS, True RTLS, & True Ocean set to -> 1
- None None, False ASDS, None ASDS, False Ocean, False RTLS

GitHub: <u>Data Wrangling</u>

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

- Exploratory Data Analysis performed on variables Flight Number, Payload Mass, Launch Site, Orbit, Class and Year.
- Plots Used:
- Flight Number vs. Payload Mass, Flight Number vs. Launch Site, Payload Mass vs. Launch Site, Orbit vs. Success Rate, Flight Number vs. Orbit, Payload vs Orbit, and Success Yearly Trend
- Scatter plots, line charts, and bar plots were used to compare relationships between variables to decide if a relationship exists so that they could be used in training the machine learning model
- GitHub: <u>EDA with Data Visualization</u>

#### **EDA** with SQL

- Loaded data set into IBM DB2 Database.
- Queried using SQL Python integration.
- Queries were made to get a better understanding of the dataset.
- Queried information about launch site names, mission outcomes, various pay load sizes
  of customers and booster versions, and landing outcomes
- GitHub url: <u>EDA with SQL</u>

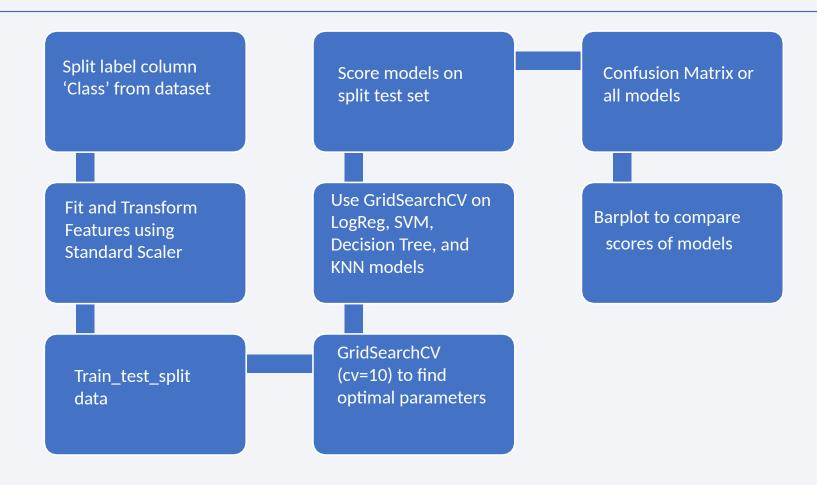
#### Build an Interactive Map with Folium

- Folium maps mark Launch Sites, successful and unsuccessful landings, and a proximity example to key locations: Railway, Highway, Coast, and City.
- This allows us to understand why launch sites may be located where they are. Also visualizes successful landings relative to location.
- GitHub: Interactive Map with Folium

## Build a Dashboard with Plotly Dash

- Dashboard includes a pie chart and a scatter plot.
- Pie chart can be selected to show distribution of successful landings across all launch sites and can be selected to show individual launch site success rates.
- Scatter plot takes two inputs: All sites or individual site and payload mass on a slider between 0 and 10000 kg.
- The pie chart is used to visualize launch site success rate.
- The scatter plot can help us see how success varies across launch sites, payload mass, and
- booster version category.
- GitHub:

# Predictive Analysis (Classification)



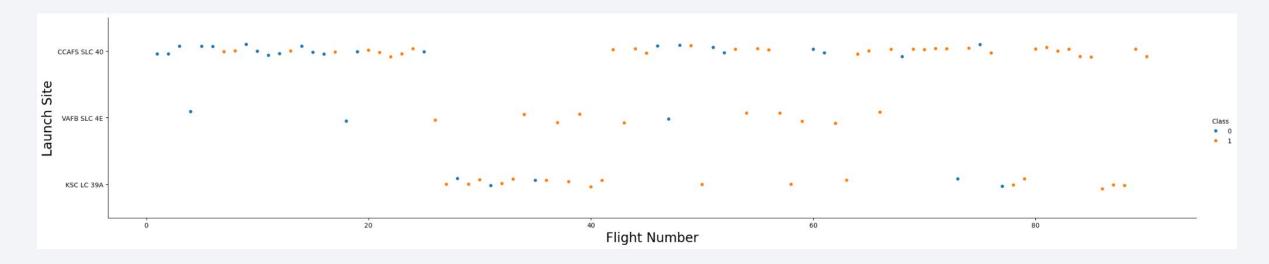
GitHub: Predictive Analysis (Classification)

#### Results

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

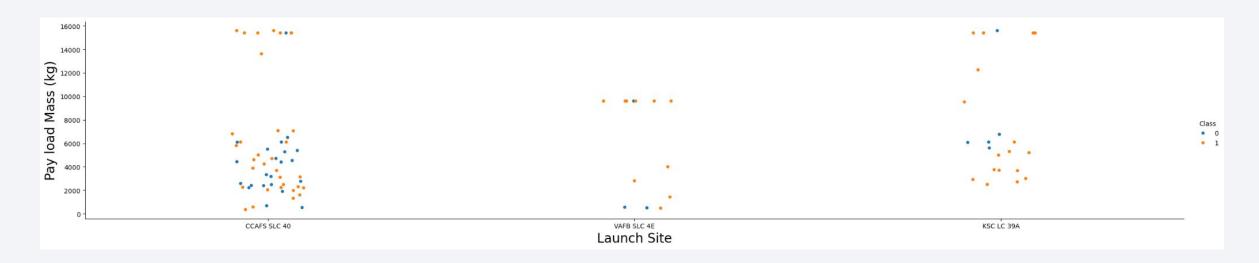


## Flight Number vs. Launch Site



Graphic suggests an increase in success rate over time (indicated in Flight Number). Likely a big breakthrough around flight 20 which significantly increased success rate. CCAFS appears to be the main launch site as it has the most volume.

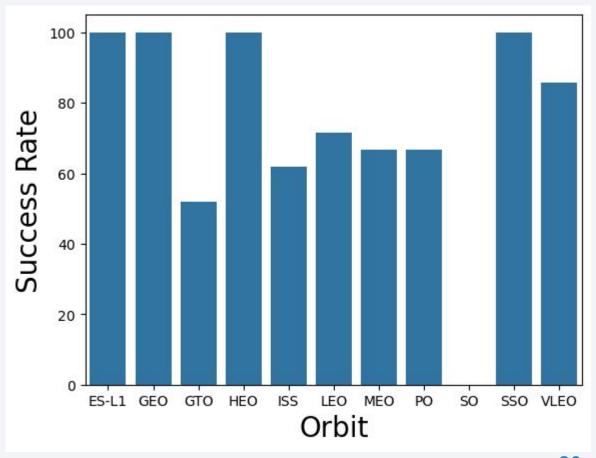
# Payload vs. Launch Site



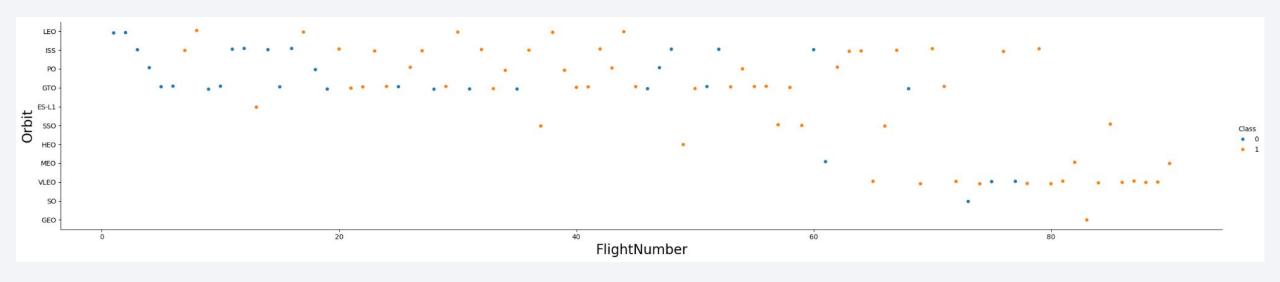
Payload mass appears to fall mostly between 0-6000 kg. Different launch sites also seem to use different payload mass.

# Success Rate vs. Orbit Type

- ES-L1 (1), GEO (1), HEO (1)
   have 100% success rate
   (sample sizes in
   parenthesis) SSO (5) has
   100% success rate
- VLEO (14) has decent success rate and attempts
- SO (1) has 0% success rate
- GTO (27) has the around 50% success rate but largest sample



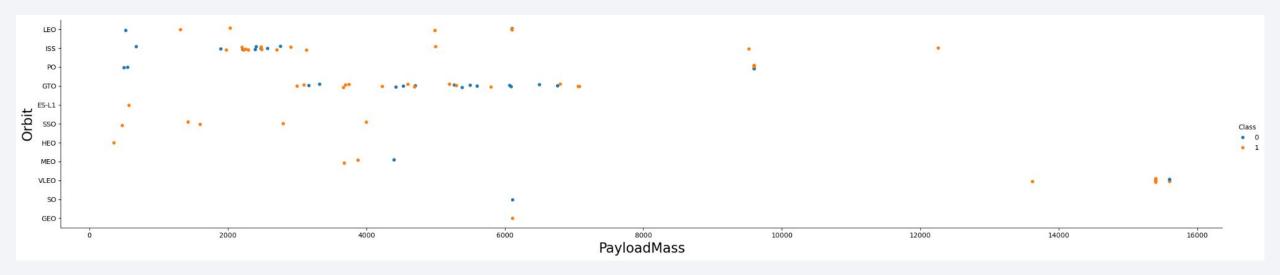
# Flight Number vs. Orbit Type



Launch Orbit preferences changed over Flight Number. Launch Outcome seems to correlate with this preference.

SpaceX started with LEO orbits which saw moderate success LEO and returned to VLEO in recent launches SpaceX appears to perform better in lower orbits or Sun-synchronous orbits

## Payload vs. Orbit Type



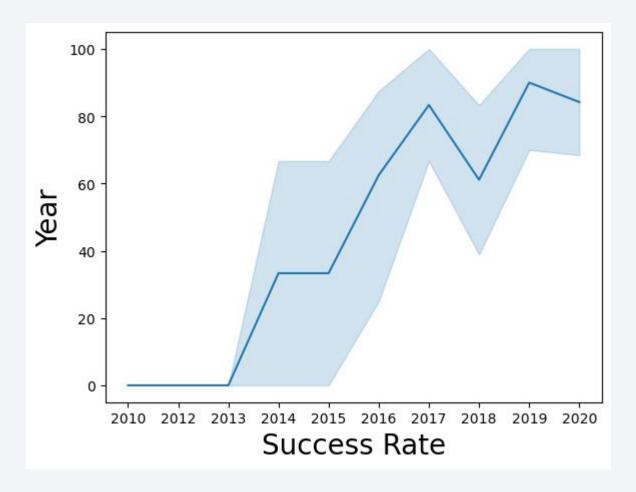
Payload mass seems to correlate with orbit

LEO and SSO seem to have relatively low payload mass

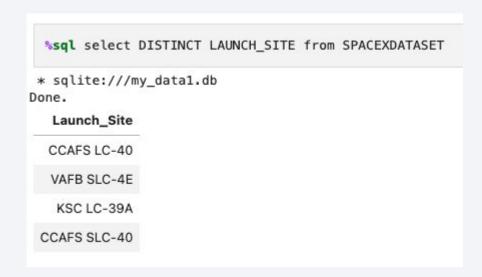
The other most successful orbit VLEO only has payload mass values in the higher end of the range

# Launch Success Yearly Trend

- Success generally increases over time since 2013 with a slight dip in 2018
- Success in recent years at around 80%



#### All Launch Site Names



- Query unique launch site names from database.
- CCAFS SLC-40 and CCAFSSLC-40 likely all represent the same
- launch site with data entry errors.
- CCAFS LC-40 was the previous name. Likely only 3 unique launch\_site values: CCAFS SLC-40, KSC LC-39A, VAFB SLC-4E

## Launch Site Names Begin with 'CCA'

\*sql select \* from SPACEXDATASET where launch\_site like 'CCA%' limit 5 \* sqlite:///my\_data1.db Done. Booster\_Version Launch\_Site Payload PAYLOAD\_MASS\_\_KG\_ Orbit Customer Mission\_Outcome Landing\_ Date Dragon 2010-CCAFS LC-Spacecraft 06-18:45:00 F9 v1.0 B0003 LEO SpaceX Success Failure (g Qualification 04 Unit Dragon demo flight 2010-C1, two NASA CCAFS LC-LEO 12-15:43:00 F9 v1.0 B0004 CubeSats, (COTS) Success Failure (g 08 barrel of NRO Brouere cheese 2012-Dragon CCAFS LC-LEO NASA 05-7:44:00 F9 v1.0 B0005 demo flight Success (COTS) 22 2012-CCAFS LC-LEO NASA SpaceX 10-0:35:00 F9 v1.0 B0006 Success CRS-1 (CRS) 08 2013-CCAFS LC-LEO NASA SpaceX 15:10:00 F9 v1.0 B0007 Success (ISS) 40 CRS-2 (CRS) 01

First five entries in database with Launch Site name beginning with CCA.

## **Total Payload Mass**

```
%sql select sum(payload_mass__kg_) as sum from SPACEXDATASET where customer like 'NASA (CRS)'

* sqlite://my_data1.db
Done.
    sum
    45596
```

- This query sums the total payload mass in kg where NASA was the customer.
- CRS stands for Commercial Resupply Services which indicates that these payloads were sent to the International Space Station (ISS).

# Average Payload Mass by F9 v1.1

- This query calculates the average payload mass or launches which used booster version F9 v1.1
- Average payload mass of F9 1.1 is on the low end of our payload mass range

## First Successful Ground Landing Date

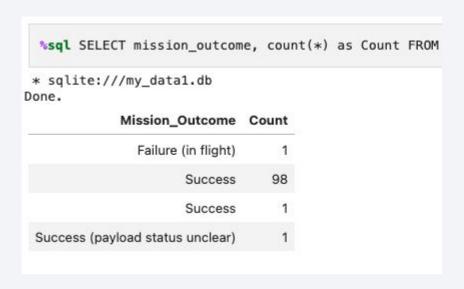
- This query returns the first successful ground pad landing date.
- First ground pad landing wasn't until the end of 2015.
- Successful landings in general appear starting 2014.

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



This query returns the four booster versions that had successful drone ship landings and a payload mass between 4000 and 6000 non inclusively.

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



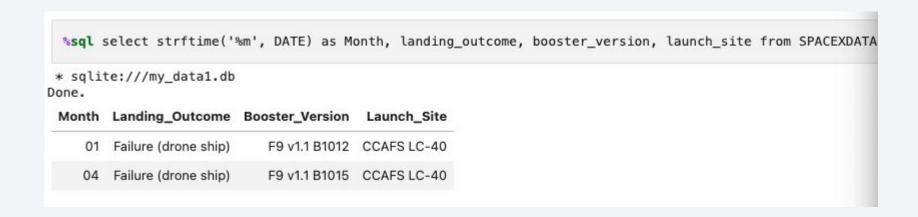
- This query returns a count of each mission outcome.
- SpaceX appears to achieve its mission outcome nearly 99% of the time.
- This means that most of the landing failures are intended.
- Interestingly, one launch has an unclear payload status and unfortunately one failed in flight.

# **Boosters Carried Maximum Payload**

```
maxm = %sql select max(payload mass kg )
 maxv = maxm[0][0]
 %sql select booster version from SPACEXDAT
* sqlite:///my_data1.db
Done.
* sqlite:///my_data1.db
Done.
 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
```

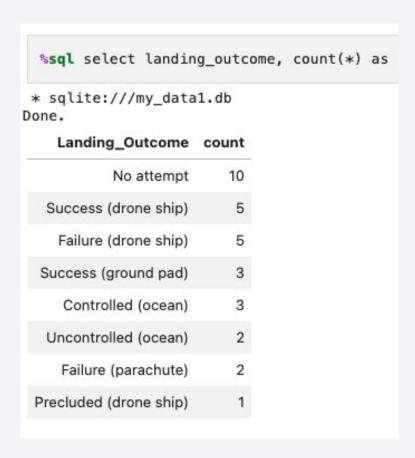
- This query returns the booster versions that carried the highest payload mass of 15600 kg.
- These booster versions are very similar and all are of the F9 B5 B10xx.x variety.
- This likely indicates payload mass correlates with the booster version that is used.

#### 2015 Launch Records



- This query returns the Month, Landing Outcome, Booster Version, Payload Mass (kg), and Launch site of 2015 launches where stage 1 failed to land on a drone ship.
- There were two such occurrences.

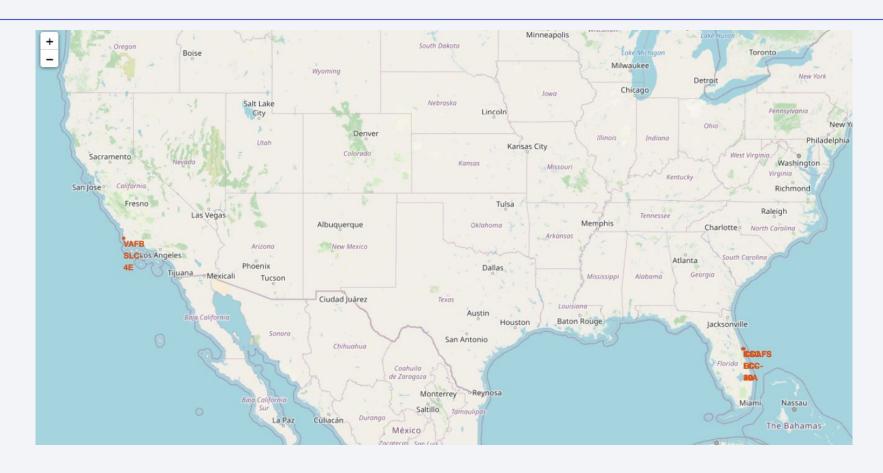
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20



- This query returns a list of successful landings and between 2010-06-04 and 2017-03-20 inclusively.
- There are two types of successful landing outcomes: drone ship and ground pad landings.
- There were 8 successful landings in total during this time period



#### **Launch Site Locations**



The map shows all Launch Site Locations in US

## Success/Failed Launch Markers



Clusters on Folium map can be clicked on to display each successful landing (green icon) and failed landing (red icon). This example shows 3 successful landings and 4 failed landings.

# **Key Location Proximities**

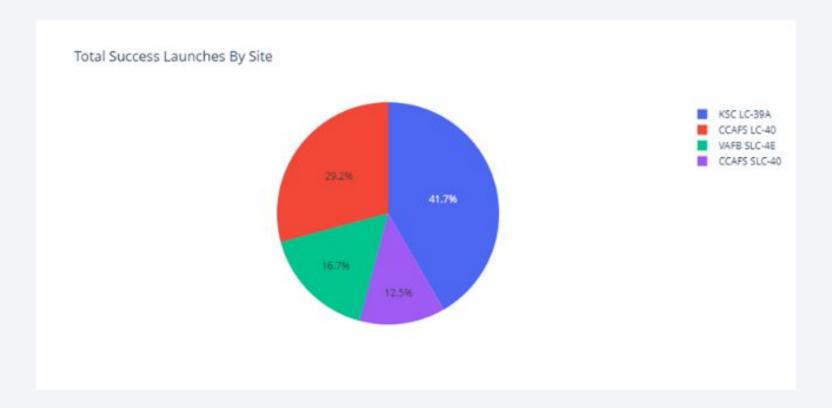


In this example, a line connects a launch site and the closest coastline.



## Total Success Launches By Site

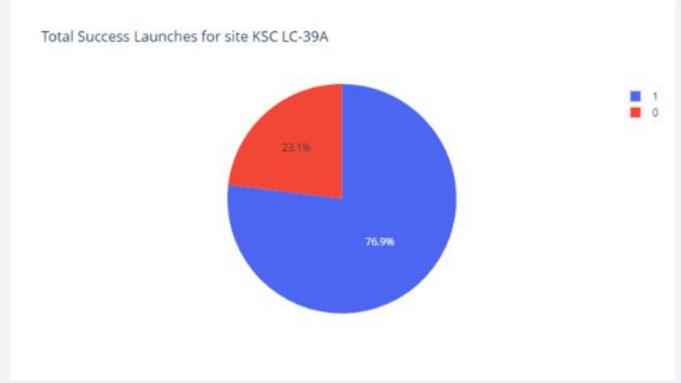
KSC LC-39A is the site with the higher success launches followed by CCAFS LC-40



#### KSC LC-39A

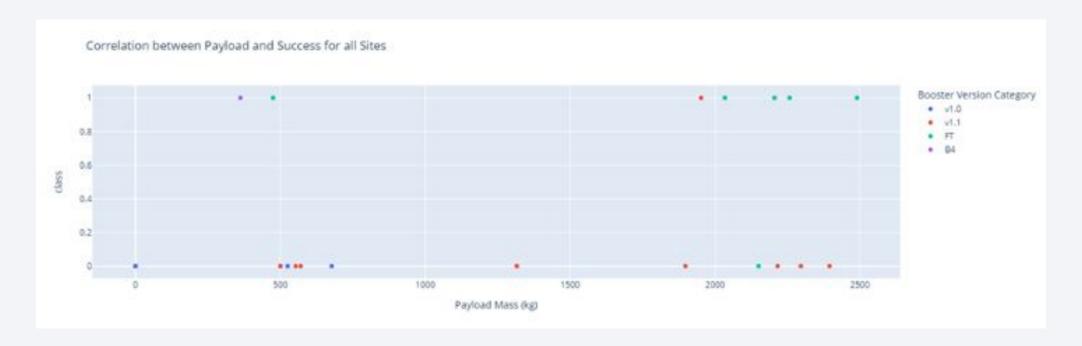
• The pie chart for the launch site KSC LC-39A shows the site with highest launch

success ratio.



## Payload vs Launch Outcome

• Scatter plot for all sites with 2500 kg, 5000 kg and 10000 kg payload ranges.

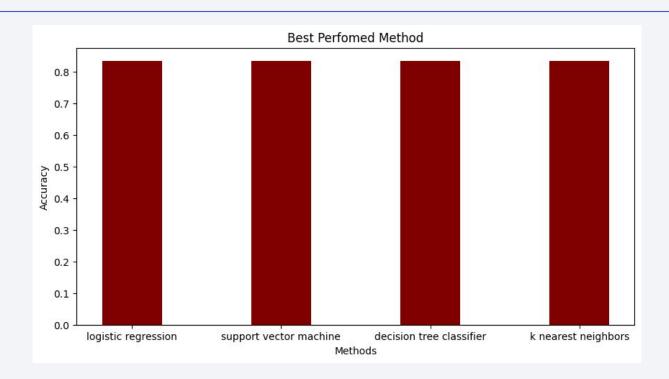


# Payload vs Launch Outcome





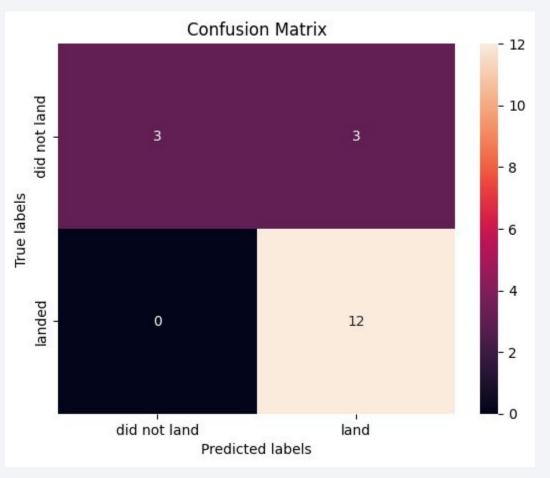
### **Classification Accuracy**



- All models had virtually the same accuracy on the test set at 83.33% accuracy.
   It should be noted that test size is small at only sample size of 18.
- We likely need more data to determine the best model.

#### **Confusion Matrix**

 Correct predictions are on a diagonal from top left to bottom right.



#### Conclusions

- As all the algorithms are giving the same accuracy, they all perform practically the same.
- By using our machine learning model, we can predict if the first stage of our competitor will land and determine the cost of a launch.

# **Appendix**

- For all notebooks follow this link
- <u>GitHub Repository</u>

