

Winning Space Race with Data Science

Todd Kurtz July 2, 2023



Outline

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

Executive Summary

- This capstone will try to predict if Space X Falcon 9 first stage will land successfully using machine learning and Data Collection, Data Wrangling and Formatting
- The process for the project used the following steps:
 - Data Collection
 - Exploratory data analysis
 - Visualization
 - Machine Learning Prediction
- Based on different aspects of the rocket launch, we will try and determine whether the outcome
 will be a success or failure

Introduction

- At a cost of 62 million dollars per launch, Space X provides a substantial savings over its competitors. The savings is achieved mostly by reusing the first stage of the rocket. By predicting if the first stage will land, we can therefore determine the cost of each launch.
- We are hoping to solve the main problem of determining if the first stage of the rocket will land successfully. Given a set of features, is the outcome predictable?



Methodology

Executive Summary

- Data collection methodology:
 - Data was collected using the SpaceX API and Web scraping
- Perform data wrangling
 - Data was cleansed looking for nulls and normalization
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Decision Tree
 - Logistic Regression
 - Support vector machine
 - K-nearest neighbors

Data Collection

- Data sets were collected via the SpaceX API and web scraping
 - SpaceX API: https://api.spacexdata.com/v4/rockets/
 - Web scraping was performed from Wikipedia
 - https://en.wikipedia.org/w/index.php?title=List of Falcon 9 and Falcon Heavy launches&oldid=10276
 86922

Data Collection - SpaceX API

- Using the GET request, we used the API to import Booster Version, Launchpad, Payload, and Rocket Core information
- Data was parsed from the json output, Falcon 9 Boosters were filtered and null values were replaced using the mean of the column
- Total data set is 90 rows and 17 columns.
- Applied-Data-Science-Capstone/01 Capstone Lab 1 Collecting the Data.ipynb at main · mberringer/Applied-Data-Science-Capstone (github.com)

- 1. Get Rocket
- 2. Get Launchpad
- 3. Get Payloads
- 4. Get Cores
- 5. Parse and normalize the response
- 6. Filter for Falcon 9 only
- 7. Replace nulls with column mean

Data Collection - Scraping

 Using Beautiful Soup to extract HTML tables from Wikipedia

Applied-Data-Science-Capstone/01a
 Capstone Lab 1a Complete the Data
 Collection with Web Scraping lab.ipynb at main · mberringer/Applied-Data-Science-Capstone (github.com)

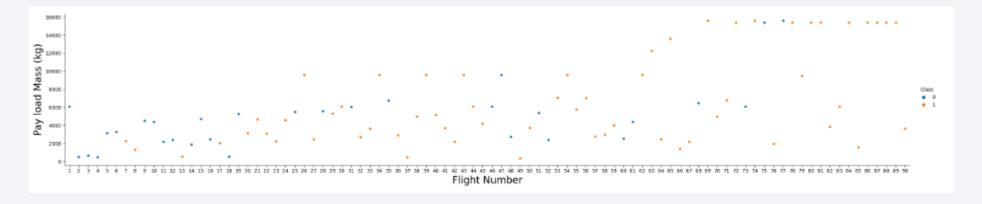
- 1. Extract the HTML tables from Wikipedia
- 2. Parse the table and convert to Pandas data frame

Data Wrangling

- The object is to determine if the booster successfully landed or if it failed
- We performed exploratory Data analysis and determined training labels
- We first imported out data set from the prior steps
- We identified missing values for each attribute
- Calculated number of launches per site, the number of occurrences of each orbit, number and occurrence per orbit type and mission outcome
- Created the landing outcome label
- <u>Applied-Data-Science-Capstone/02 Capstone Lab 2 Data Wrangling.ipynb at main · mberringer/Applied-</u> Data-Science-Capstone · GitHub

EDA with Data Visualization

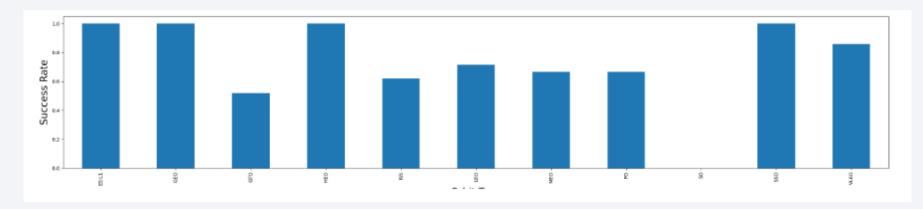
 We plotted scatter plots of Flight Number and Payload Mass to determine success rate



• <u>Applied-Data-Science-Capstone/04 Capsone Lab 4 EDA with Visualization Lab.ipynb at main · mberringer/Applied-Data-Science-Capstone · GitHub</u>

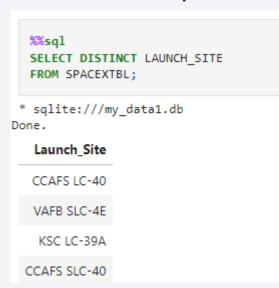
EDA with Data Visualization

• We plotted bar charts to determine success rate of each orbit

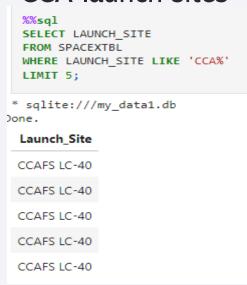


• <u>Applied-Data-Science-Capstone/04 Capsone Lab 4 EDA with Visualization Lab.ipynb at main · mberringer/Applied-Data-Science-Capstone · GitHub</u>

• Names of unique launch sites



CCA launch sites



Total payload mass for NASA boosters

```
%%sql
SELECT SUM(PAYLOAD_MASS__KG_) as PayloadMass
FROM SPACEXTBL
WHERE Customer = 'NASA (CRS)';

* sqlite:///my_data1.db
Done.

PayloadMass

45596.0
```

Average payload mass for booster Falcon 9

```
Display average payload mass carried by booster version F9 v1.1

%%sql
SELECT AVG(PAYLOAD_MASS__KG_) as AvgPayloadMass
FROM SPACEXTBL
WHERE Booster_Version LIKE 'F9 v1.0%';

* sqlite:///my_data1.db
Done.

AvgPayloadMass

340.4
```

- First successful landing outcome on the ground pad
- Name of boosters which have success in drone ship and have a payload mass between 4,000 and 6,000
- Total Number of successful and failure mission outcomes
- Booster version names that have carried the maximum payload mass
- Landing outcome rank between 6/4/2010 and 3/20/2017
- <u>Applied-Data-Science-Capstone/03 Capsone Lab 3 Complete the EDA with SQL.ipynb at main · mberringer/Applied-Data-Science-Capstone · GitHub</u>

Build an Interactive Map with Folium

- We marked all launch sites on a map and added the success rate for each site
- We calculated the distance between a launch site to its proximities
- We used circles and labels to display launch sites
- We used markers and lines to display the success rate of each site along with its proximity to other sites
- <u>Applied-Data-Science-Capstone/05 Capsone Lab 5 Interactive Visual Analytics with Folium.ipynb at main ·</u> mberringer/Applied-Data-Science-Capstone · GitHub

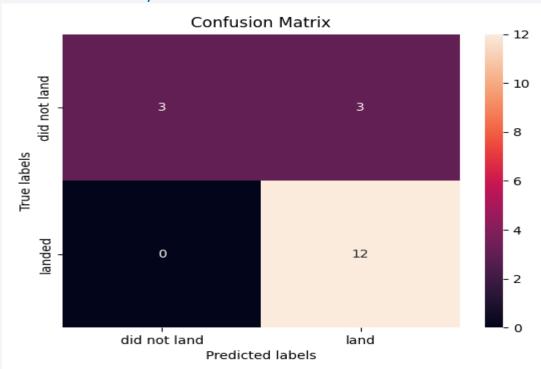
Build a Dashboard with Plotly Dash

- Dash functions were used to generate an interactive app to toggle input using dropdown menus and sliders
- Pie charts and scatterplots were used to show total launch success rate from each site and the correlation of payload mass to mission outcome per site
- <u>Applied-Data-Science-Capstone/06 Captstone Lab 6 Build and Interactive Dashboard with Ploty Dash.py at main · mberringer/Applied-Data-Science-Capstone · GitHub</u>

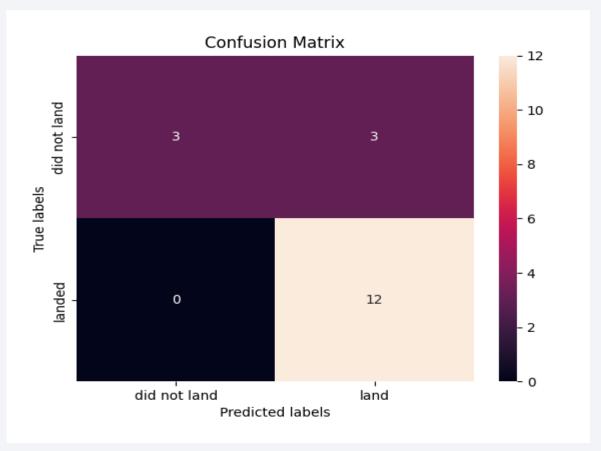
Predictive Analysis (Classification)

- Sklearn library was used to create the machine learning models
- We plotted the confusion matrix for each of the models to determine effectiveness
- Data was first standardized then split into training and testing data sets
- Models created included:
 - Logistic regression
 - Support Vector machine (SVM)
 - Decision tree
 - K nearest neighbors (KNN)
- Models were fit on the training set and the best combination of parameters for each model was determined
- Model accuracy was based on the confusion matrix and accuracy scores
- Applied-Data-Science-Capstone/07 Capsone Lab 7 Complete the Machine Learning Prediction lab.ipynb at main · mberringer/Applied-Data-Science-Capstone · GitHub

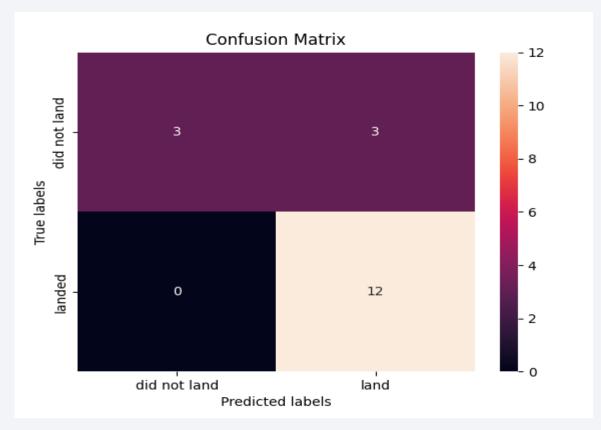
- Logistic regression
 - Major problem with false positives
 - Accuracy: 0.848



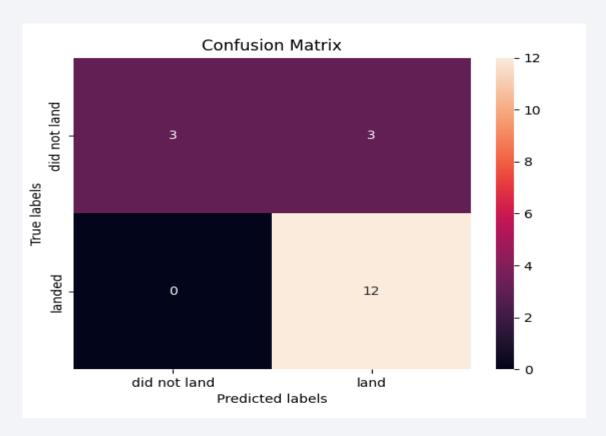
- Support Vector Machine
 - Accuracy: 0.833



- Decision Tree
 - Accuracy: 0.875



- K nearest neighbors (KNN)
 - Accuracy: 0.833

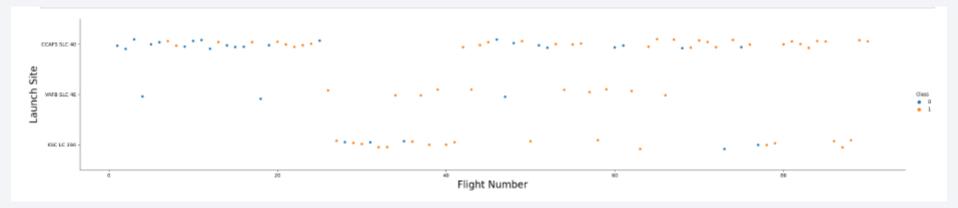


- Using each model GridSearchCV, the best scores are used to rank each model
- The models are ranked in order below, best to worst
 - Decision Tree
 - K nearest neighbors
 - Support Vector machine
 - Logistic regression



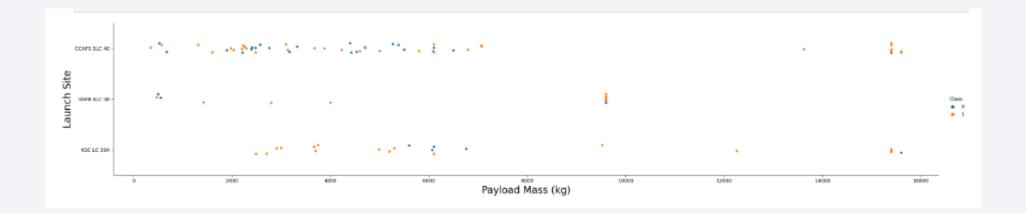
Flight Number vs. Launch Site

• Show a scatter plot of Flight Number vs. Launch Site



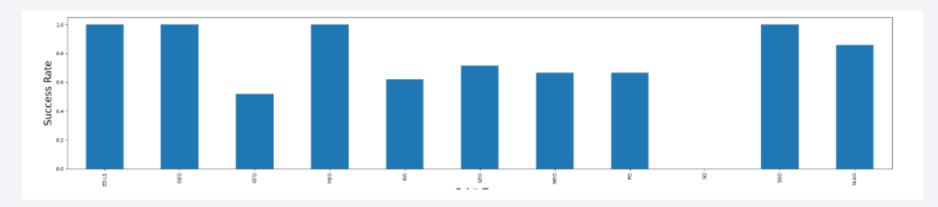
Payload vs. Launch Site

• Show a scatter plot of Payload vs. Launch Site

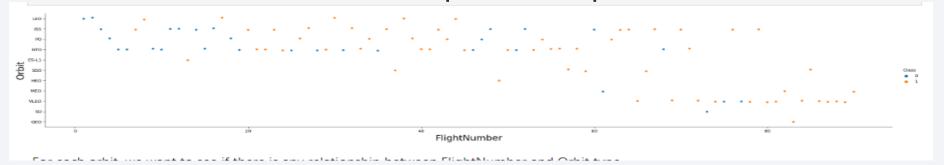


Success Rate vs. Orbit Type

Show a bar chart for the success rate of each orbit type

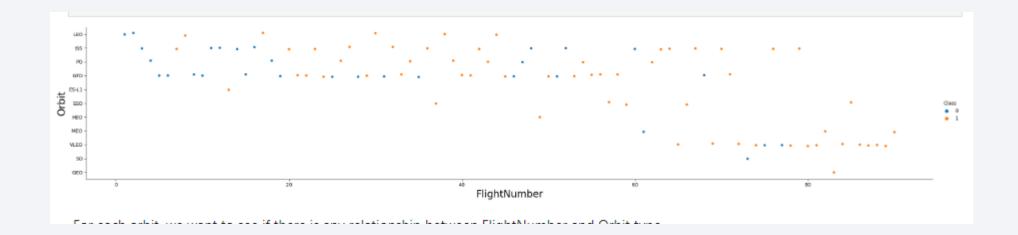


• Show the screenshot of the scatter plot with explanations



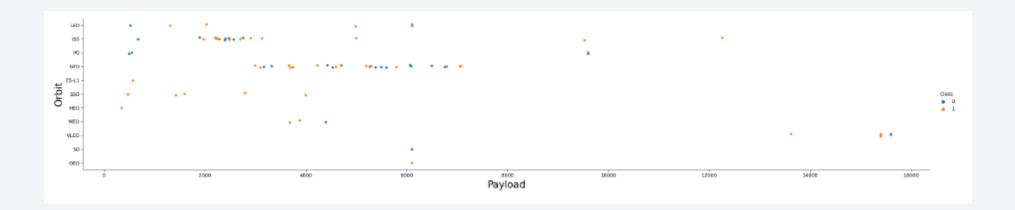
Flight Number vs. Orbit Type

• Show a scatter point of Flight number vs. Orbit type



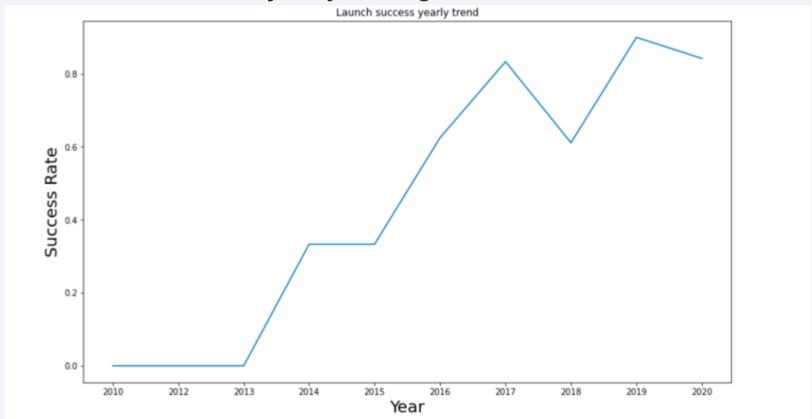
Payload vs. Orbit Type

• Show a scatter point of payload vs. orbit type



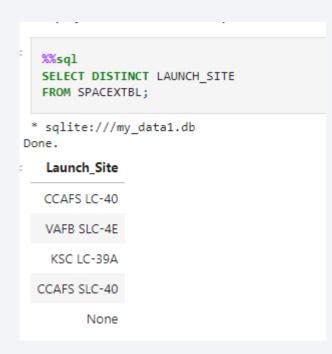
Launch Success Yearly Trend

• Show a line chart of yearly average success rate



All Launch Site Names

• Find the names of the unique launch site



Launch Site Names Begin with 'CCA'

• Find 5 records where launch sites begin with `CCA`

```
Display 5 records where launch sites begin with
  %%sql
  SELECT LAUNCH SITE
  FROM SPACEXTBL
  WHERE LAUNCH SITE LIKE 'CCA%'
  LIMIT 5;
 * sqlite:///my_data1.db
Done.
 Launch_Site
 CCAFS LC-40
 CCAFS LC-40
 CCAFS LC-40
 CCAFS LC-40
 CCAFS LC-40
```

Total Payload Mass

Calculate the total payload carried by boosters from NASA

Average Payload Mass by F9 v1.1

Calculate the average payload mass carried by booster version F9 v1.1

```
Display average payload mass carried by booster version F9 v1.1

**Select Avg(PAYLOAD_MASS__KG_) as AvgPayloadMass
FROM SPACEXTBL
WHERE Booster_Version LIKE 'F9 v1.0%';

* sqlite:///my_data1.db
Done.

AvgPayloadMass

340.4
```

First Successful Ground Landing Date

• Find the dates of the first successful landing outcome on ground pad

```
List the date when the first successful landing outcome in ground pad was acheived.

Hint:Use min function

***sql
SELECT MIN(Date)
FROM SPACEXTBL
WHERE Landing_Outcome = 'Success (ground pad)';

* sqlite:///my_data1.db
Done.

MIN(Date)

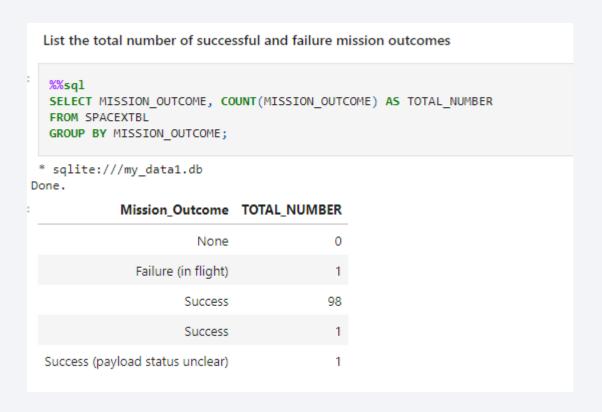
01/08/2018
```

Successful Drone Ship Landing with Payload between 4000 and 6000

 List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

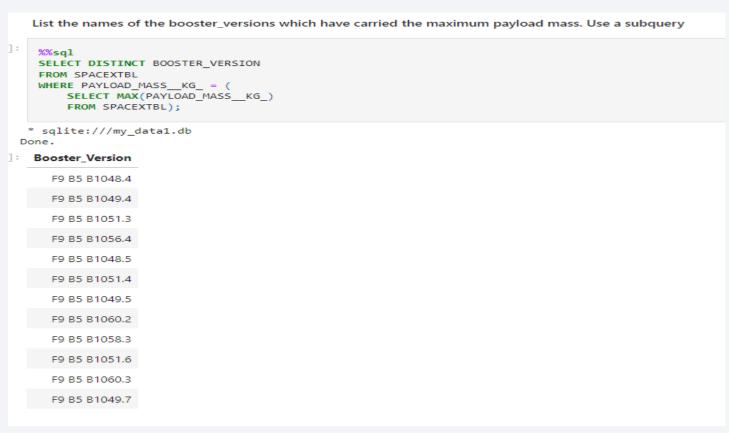
Total Number of Successful and Failure Mission Outcomes

Calculate the total number of successful and failure mission outcomes



Boosters Carried Maximum Payload

• List the names of the booster which have carried the maximum payload mass



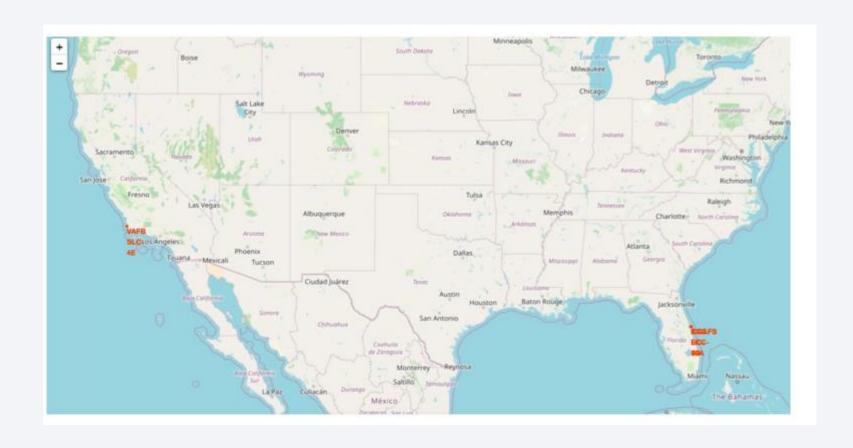
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

 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

%%sql SELECT LANDING_OUTCOME, COUNT(LANDING_OUTCOME) AS TOTAL_NUM FROM SPACEXTBL GROUP BY LANDING_OUTCOME ORDER BY TOTAL_NUMBER DESC		
sqlite:///my_data1 ne.	.db	
Landing_Outcome	TOTAL_NUMBER	
Success	38	
No attempt	21	
Success (drone ship)	14	
Success (ground pad)	9	
Failure (drone ship)	5	
Controlled (ocean)	5	
Failure	3	
Uncontrolled (ocean)	2	
Failure (parachute)	2	
Precluded (drone ship)	1	
No attempt	1	
None	0	



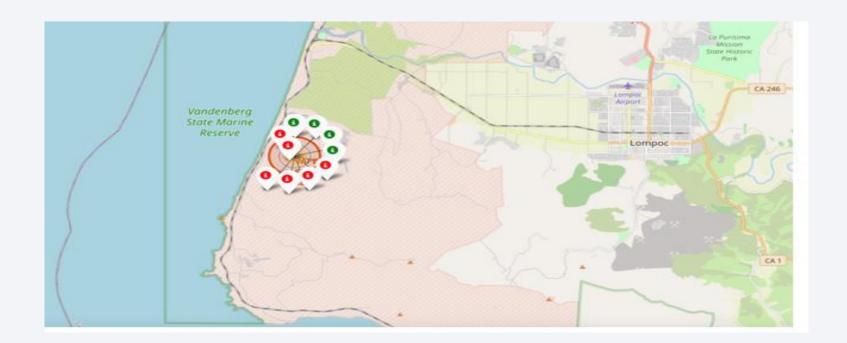
All Launch sites map



Success Rate for each launch site

Successful launch's per site

 Zooming in shows green or red with green indicating a success and red indicating a failure



Distance between launch sites and proximities

- The distance between a launch site and the proximity including the nearest city, highway or railway.
- Proximity of Site VFAB SLC 4E to nearest coastline





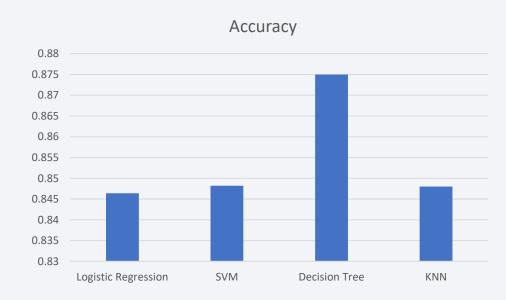
Payload mass and success rate





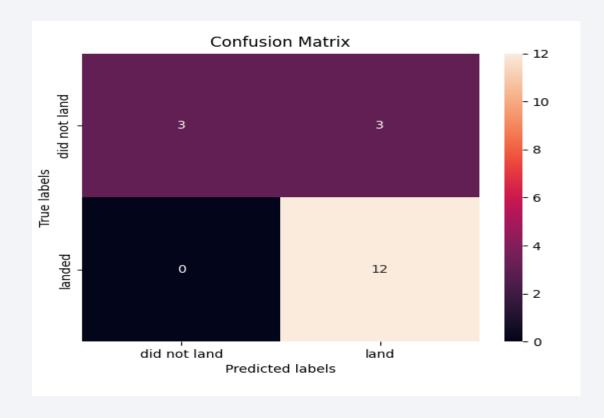
Classification Accuracy

Decision Tree is the most accurate



Confusion Matrix

Decision tree confusion matrix



Conclusions

- We attempted to predict the success rate of the Falcon 9 booster given a set of parameters to determine the cost of each launch
- Payload Mass, Orbit Type and launch site may impact the success rate
- Using different machine learning models to make the prediction, the decision tree model was the most accurate

