

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 through API
 - Data Collection with Web Scraping
 - Data Wrangling
 - Exploratory Data Analysis with SQL
 - Exploratory Data Analysis with Data Visualization
 - Interactive Visual Analytics with Folium
 - Machine Learning Prediction
- Summary of all results
 - Exploratory Data Analysis result
 - Interactive analytics in screenshots
 - Predictive Analytics result

Introduction

Project background and context

Space X 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 Space X can reuse the first stage. Therefore, 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 space X for a rocket launch. This goal of the project is to create a machine learning pipeline to predict if the first stage will land successfully.

Problems you want to find answers

- What factors determine if the rocket will land successfully?
- The interaction amongst various features that determine the success rate of a successful landing.
- What operating conditions needs to be in place to ensure a successful landing program.



Methodology

Executive Summary

- Data collection methodology:
 - Through SpaceX Rest API and Web scrapping from Wikipedia.
- Perform data wrangling
 - One-hot encoding was applied to categorical features and data fields for Machine Learning and data cleaning of null values and irrelevant columns
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - How to build, tune, evaluate classification models of Decision tree, K-nearest Neighbors, Logistic regression, Support Vector Machine to find the best classifier

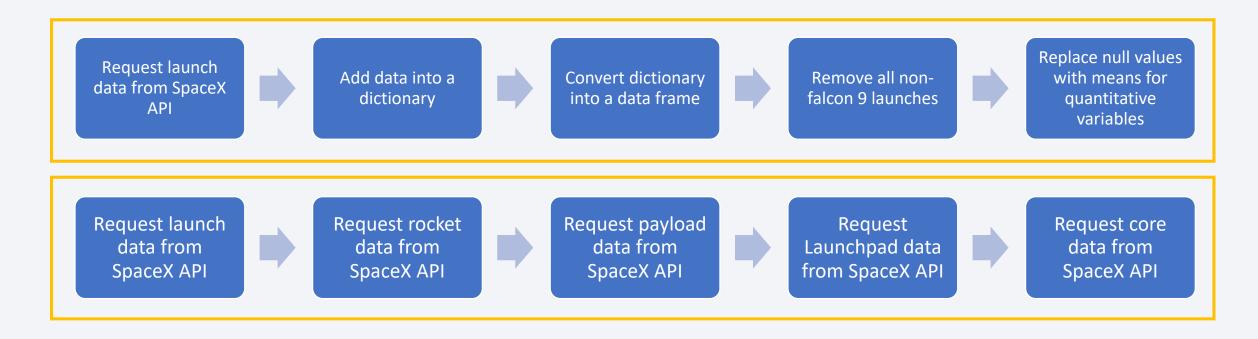
Data Collection

• The data was collected using various methods

	SpaceX API	Web Scraping
Source	Space X	Wikipedia
Step 1	Requested rocket launch data from SpaceX API	Extracted a Falcon 9 launch records HTML table from Wikipedia
Step 2	Decoded the response content as a Json using .json() and turned it into a Pandas data frame using .json_normalize()	Parsed the table and convert it into Pandas data frame using BeautifulSoup
Step 3	Performed data cleaning for missing values by replacing it with mean value of quantitative features	Extracted all column/variable names from the HTML table header

Data Collection - SpaceX API

- Present your web scraping process using key phrases and flowcharts
- Github URL: Data Collection SpaceX API



Data Collection - Scraping

- Present your web scraping process using key phrases and flowcharts
- GitHub URL: WEB SCRAPPING

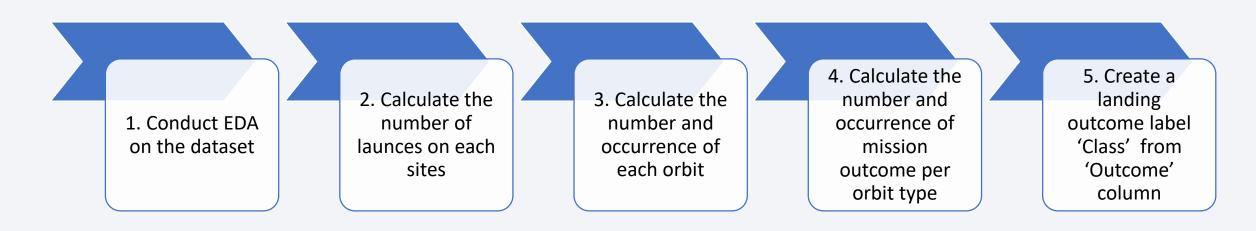
Step 1: Requested the Falcon 9 Launch data from Wikipedia URL

Step 2: Parsed the table and convert it into a Pandas data frames using Beautiful Soup

Step 3: Extract all column/variable names from the HTML table header

Data Wrangling

- Describe how data were processed
- You need to present your data wrangling process using key phrases and flowcharts
- Github URL : <u>Data Wrangling</u>



EDA with Data Visualization

- Summarize what charts were plotted and why you used those charts
- Github URL: EDA with Data Visualization

No	Features	Type of Chart	Reason
1	Flight Number vs Launch Site		To find relationship between these two numeric variables
2	Payload vs Launch Site	Coattorplat	
3	FlightNumber vs Orbit type	Scatterplot	
4	Payload vs Orbit type		
5	Success rate of each orbit type	Bar chart	To find the probability of success rate for each orbit type
6	Year vs Success rate	Line chart	To observe the trend of success rate across a period of time (years)

EDA with SQL

- Using bullet point format, summarize the SQL queries you performed
 - Launch site unique, string 'CCA'
 - Payload Mass (kg) total, average
 - Mission Outcome total number of success and fail mission outcome
 - Booster version carried the maximum payload mass
 - Date 1st successful landing outcome in ground pad

Github URL : EDA with SQL

Build an Interactive Map with Folium

- Summarize what map objects such as markers, circles, lines, etc. you created and added to a folium map
- Explain why you added those objects
- GitHub URL: Interactive Map with Folium

Add each site's location, label by using site's latitude & longitude coordinates and circle marker with name of launch sites, respectively

Visualize all launch sites into an interactive map

Mark the success/failed launches for each site on the map

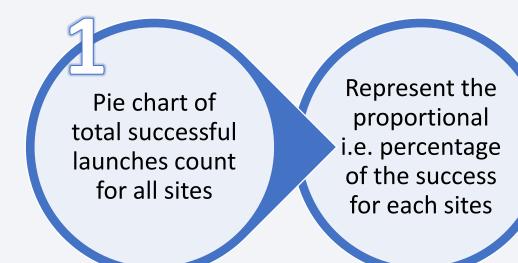
Assign feature of launch_outcomes with green marker for success launch site whereas red or failed launch site

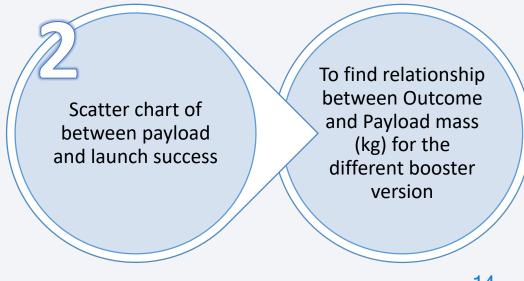
Using Haversine's formula

Calculated distanced to the closest coastline, city, railway and highway, represented by a blue line on the map

Build a Dashboard with Plotly Dash

- Summarize what plots/graphs and interactions you have added to a dashboard
- Explain why you added those plots and interactions
- GitHub URL: Plotly Dash





Predictive Analysis (Classification)

- Summarize how you built, evaluated, improved, and found the best performing classification model
- You need present your model development process using key phrases and flowchart
- GitHub URL: <u>Machine Learning Predictions</u>

Building model



Evaluating model



Improving model



Find the best model

- Loaded the data using numpy and pandas
- 2. Transformed the data
- Split our data into training and testing
- 4. Decide which ML models to be deployed

- 1. Build ML model
- Check accuracy of each ML model
- Tune different hyperparametes using GridSearchCV

- 1. Feature engineering
- 2. Algorithm tuning

 The best model will be chosen based on the best (highest) accuracy score

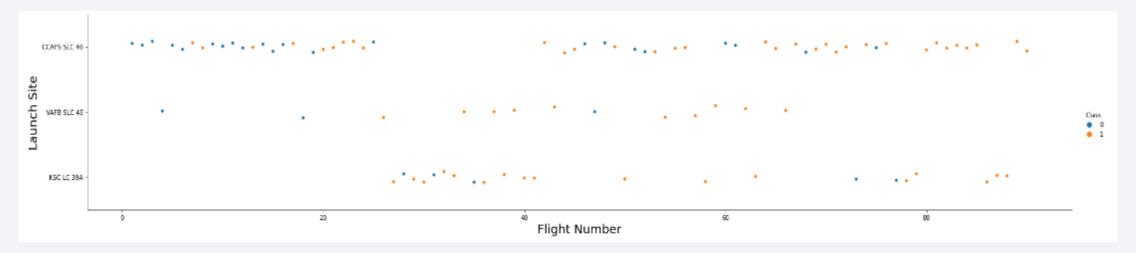
Results

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



Flight Number vs. Launch Site

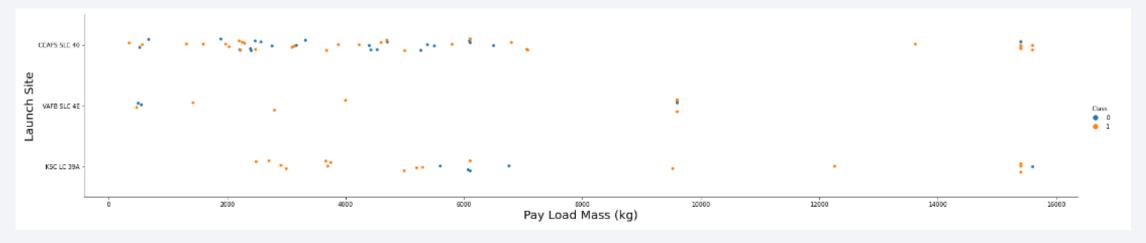
- Show a scatter plot of Flight Number vs. Launch Site
- Show the screenshot of the scatter plot with explanations



• The larger amount of the flight number at the launch site, the greater the success rate at a launch site will be.

Payload vs. Launch Site

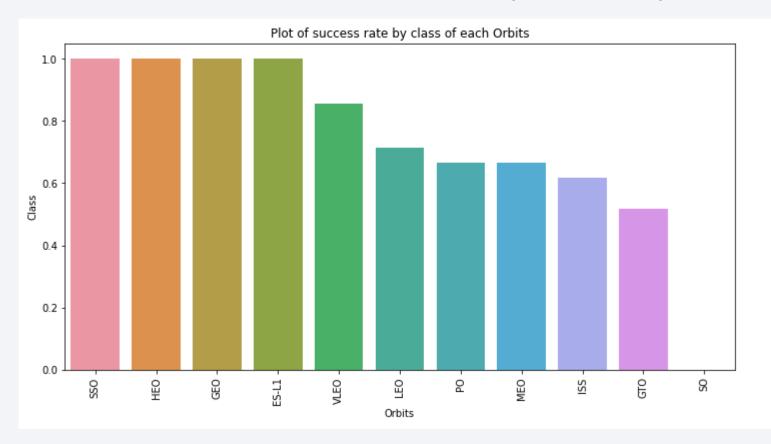
- Show a scatter plot of Payload vs. Launch Site
- Show the screenshot of the scatter plot with explanations



- CCAFS SLC 40: The greater the payload mass at the launch site, the higher the success rate for the Rocket.
- VAFB-SLC: No rockets launched for heavy payload mass (greater than 10,000 kg)

Success Rate vs. Orbit Type

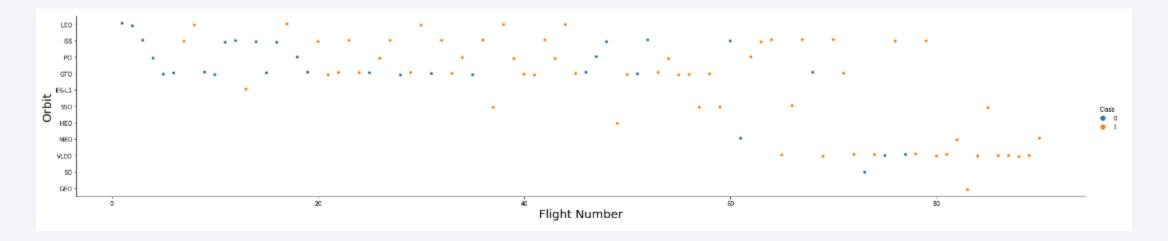
- Show a bar chart for the success rate of each orbit type
- Show the screenshot of the scatter plot with explanations



 It shows that Orbit ES-L1, GEO, HEO and SSO have 100% success rate at the landing outcomes whereas Orbit SO recorded 0% of success rate at the landing outcome

Flight Number vs. Orbit Type

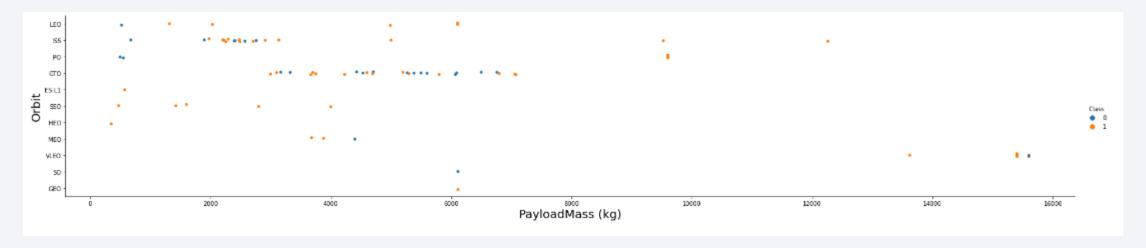
- Show a scatter point of Flight number vs. Orbit type
- Show the screenshot of the scatter plot with explanations



- In LEO orbit, the Success appears related to the number of flights
- However, there exists no relationship between flight number when in GTO orbit

Payload vs. Orbit Type

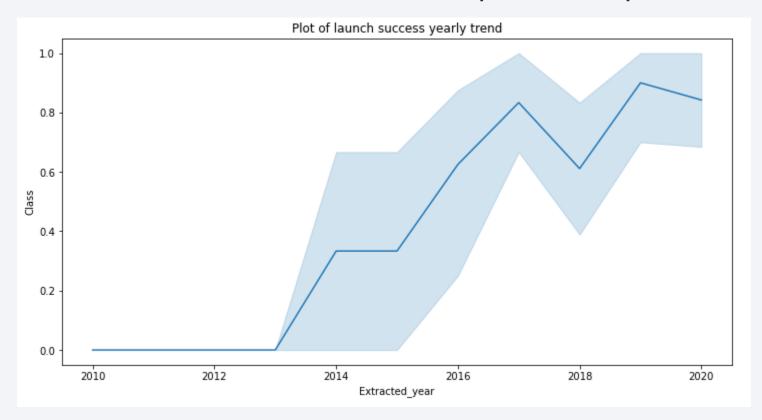
- Show a scatter point of payload vs. orbit type
- Show the screenshot of the scatter plot with explanations



- Polar, LEO and ISS Orbit: Heavy payloads contributed to the successful landing
- GTO orbit: It seems like there exists no relationship between payloadmass and landing outcomes at the site

Launch Success Yearly Trend

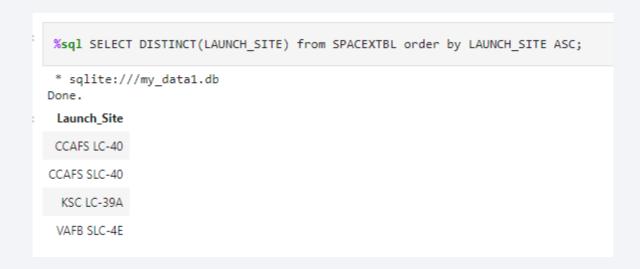
- Show a line chart of yearly average success rate
- Show the screenshot of the scatter plot with explanations



It is observed that the success rate is on an increasing trend, from 2013 to 2020

All Launch Site Names

- Find the names of the unique launch sites
- Present your query result with a short explanation here



We used the key word DISTINCT to show only unique launch sites from the SpaceX data.

Launch Site Names Begin with 'CCA'

- Find 5 records where launch sites begin with `CCA`
- Present your query result with a short explanation here
 - We used the word like 'CCA%' to find the launch sites begin with the string 'CCA'
 * We used the word like 'CCA%' to find the launch site begin with CCA
 * ** sqlite://my_datal.db
 * Done.

 Time

 Landing

X Success , Failure
(parachute)
Success
Success No attempt
Success No attempt
Success No attempt
TS IRO TS

Total Payload Mass

- Calculate the total payload carried by boosters from NASA
- Present your query result with a short explanation here

```
Display the total payload mass carried by boosters launched by NASA (CRS)

**sql select sum (PAYLOAD_MASS__KG_) from SPACEXTBL where Customer ='NASA (CRS)';

**sqlite:///my_data1.db
Done.

**um (PAYLOAD_MASS__KG_)

**We use function of SUM to find the total in column
PAYLOAD_MASS__KG__

**We use the WHERE clause to perform calculation for
Customer NASA (CRS) from the SpaceX data

**The total payload carried by boosters from NASA is
45596kg as per query above
```

Average Payload Mass by F9 v1.1

- Calculate the average payload mass carried by booster version F9 v1.1
- Present your query result with a short explanation here

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

**sql select avg(PAYLOAD_MASS__KG_) from SPACEXTBL where Booster_Version ='F9 v1.1';

* sqlite:///my_data1.db
Done.

avg(PAYLOAD_MASS__KG_)

• We use function of AVG to find the average in column
PAYLOAD_MASS__KG__

• We use the WHERE clause to perform calculation for
Booster Version F9 v1.1 from the SpaceX data

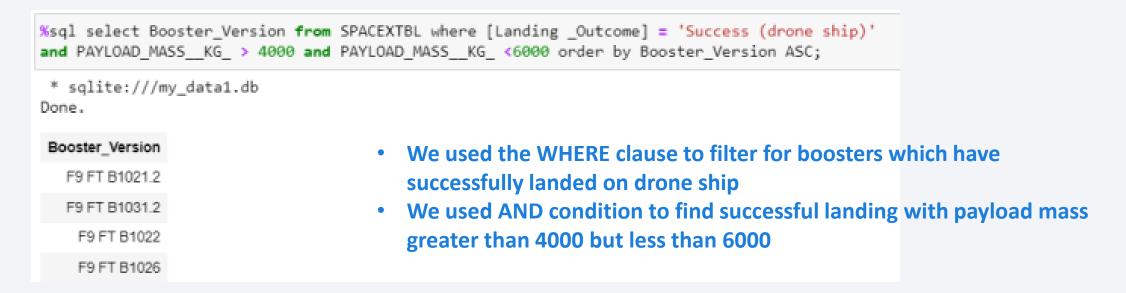
• The average payload mass carried by booster version F9
v1.1 is 2928.4 kg
```

First Successful Ground Landing Date

- Find the dates of the first successful landing outcome on ground pad
- Present your query result with a short explanation here

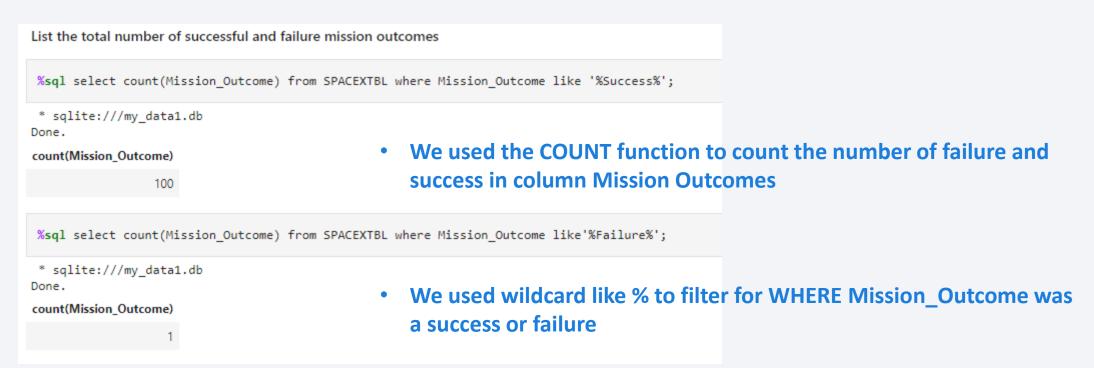
Successful Drone Ship Landing with Payload between 4000 and 6000

- List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000
- Present your query result with a short explanation here



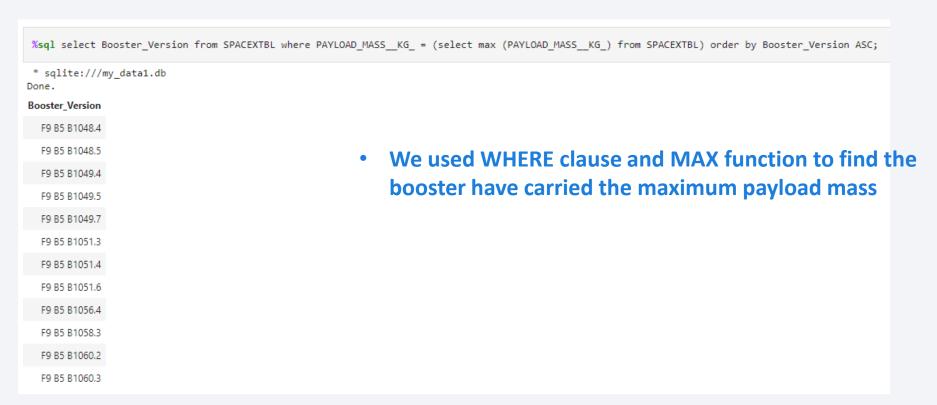
Total Number of Successful and Failure Mission Outcomes

- Calculate the total number of successful and failure mission outcomes
- Present your query result with a short explanation here



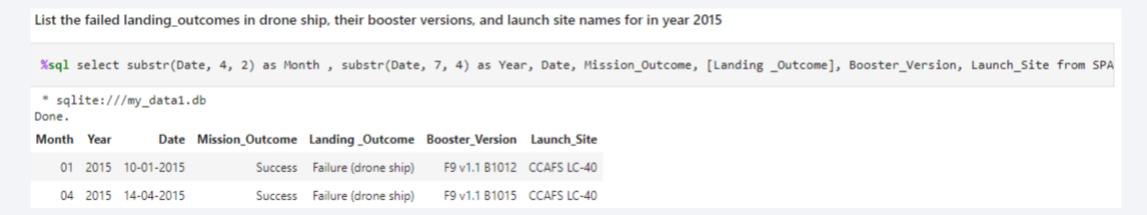
Boosters Carried Maximum Payload

- List the names of the booster which have carried the maximum payload mass
- Present your query result with a short explanation here



2015 Launch Records

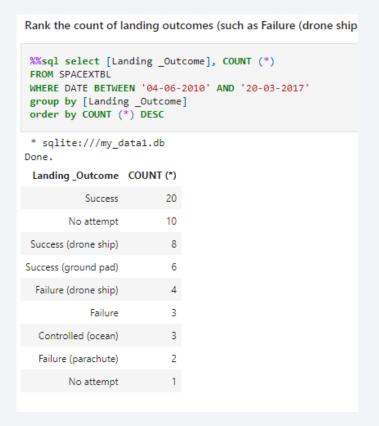
- List the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- Present your query result with a short explanation here



We used WHERE clause to find the failed landing outcomes in drone ship in year 2015

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
- Present your query result with a short explanation here

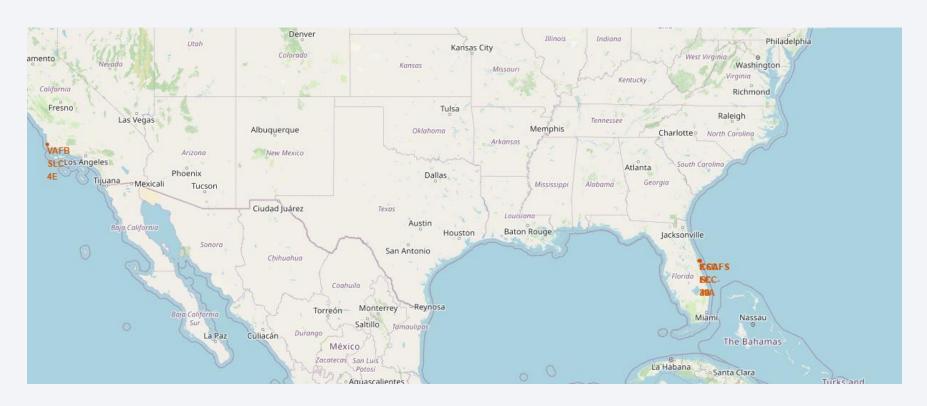


- We used COUNT function to count the frequency for each landing outcomes
- We used WHERE clause to find the data between the mentioned data
- We used ORDER BY DESC function to sort the number of outcomes in descending order



All launch sites global map markers

- Explore the generated folium map and make a proper screenshot to include all launch sites' location markers on a global map
- Explain the important elements and findings on the screenshot



All SpaceX

 launch sites are
 in United States
 of America
 coasts, Florida
 and California

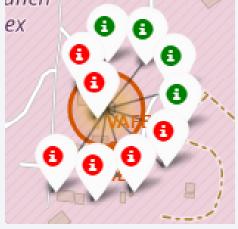
Status of each launch sites by markers

- Explore the folium map and make a proper screenshot to show the colorlabeled launch outcomes on the map
- Explain the important elements and findings on the screenshot

Green marker – Successful launches

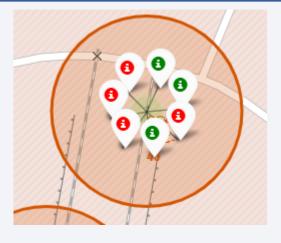
Red marker – Failed launched

FLORIDA LAUNCH SITES

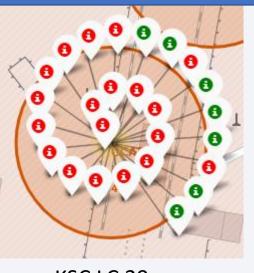


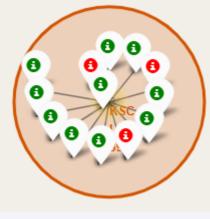
CCAFS LC 40

CALIFORNIA LAUNCH SITES





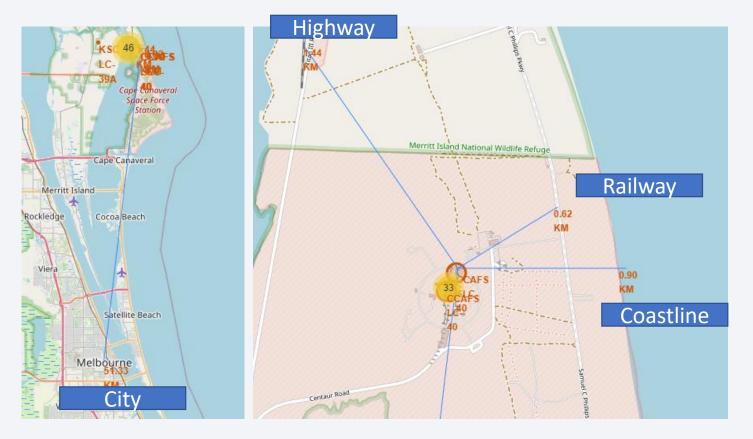




VAFB SLC 4E

Launch Site distance to its proximities

- Explore the generated folium map and show the screenshot of a selected launch site to its proximities such as railway, highway, coastline, with distance calculated and displayed
- Explain the important elements and findings on the screenshot

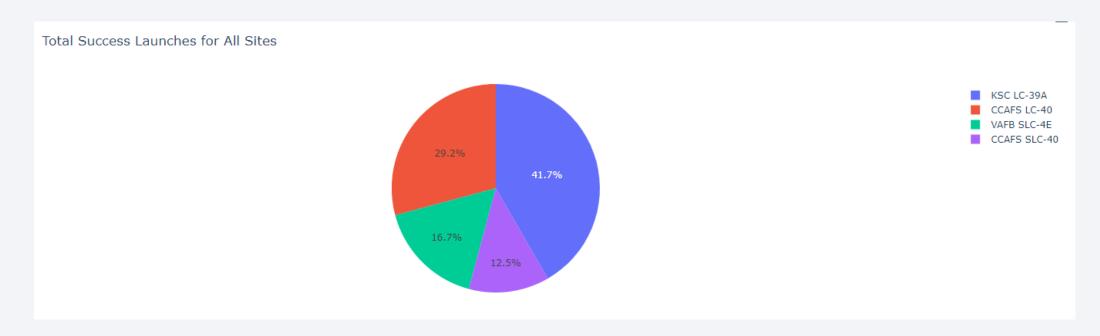


Questions	Answer
Are launch sites in close proximity to railways?	Yes
Are launch site in close proximity to highways?	No
Are launch sites in close proximity to coastline	Yes
Do launch sites keep certain distance away from cities?	Yes



Launch success count for all sites

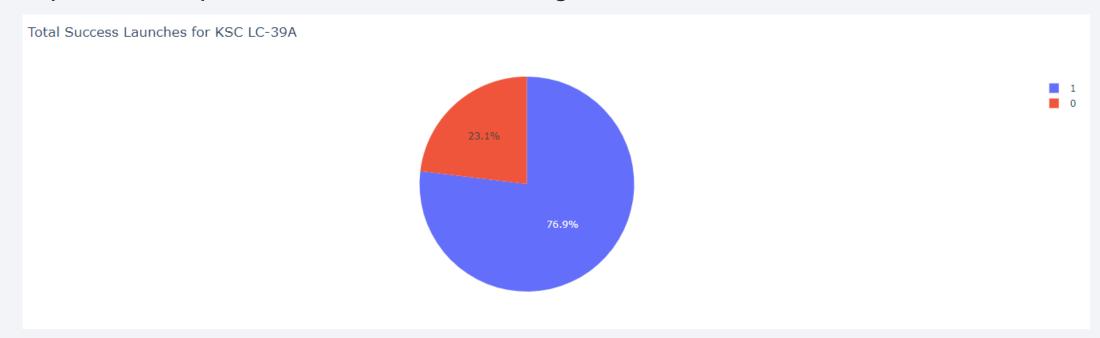
- Show the screenshot of launch success count for all sites, in a piechart
- Explain the important elements and findings on the screenshot



- KSC LC-39A was the site with highest success rate (41.7%)
- The remaining 3 launch sites success rate were below 30%

Launch site with highest success rate: KSC LC-39A

- Show the screenshot of the piechart for the launch site with highest launch success ratio
- Explain the important elements and findings on the screenshot



KSC LC-39A has a record of 76.9% of success rate whereas failure rate at 23.1%

Payload Mass (kg) vs Success rate for all sites

- Show screenshots of Payload vs. Launch Outcome scatter plot for all sites, with different payload selected in the range slider
- Explain the important elements and findings on the screenshot, such as which payload range or booster version have the largest success rate, etc.



It is observed that low weighted payload mass have largest success rate as compared to high weighted payload mass.



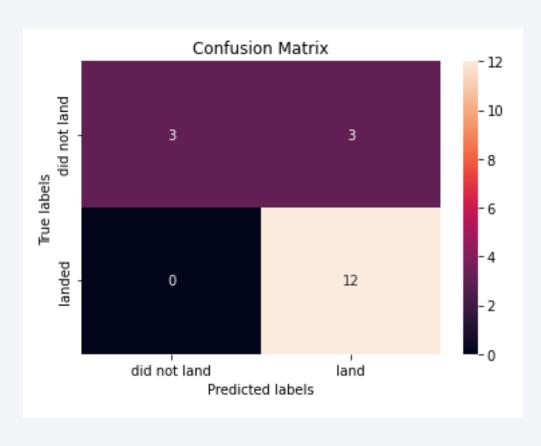
Classification Accuracy

- Visualize the built model accuracy for all built classification models, in a bar chart
- Find which model has the highest classification accuracy

```
Accuracy
                                                                                                                                                    0.87
models = { 'KNeighbors':knn cv.best score ,
               'DecisionTree':tree cv.best score ,
               'LogisticRegression':logreg cv.best score ,
               'SupportVector': svm cv.best score }
                                                                                                                  0.85
                                                                                                                                   0.85
                                                                                                 0.85
bestalgorithm = max(models, key=models.get)
 print('Best model is', bestalgorithm,'with a score of', models[bestalgorithm])
if bestalgorithm == 'DecisionTree':
    print('Best params is :', tree cv.best params )
if bestalgorithm == 'KNeighbors':
     print('Best params is :', knn cv.best params )
                                                                                                 LR
                                                                                                                  kNN
                                                                                                                                   SVM
                                                                                                                                                Decision Tree
if bestalgorithm == 'LogisticRegression':
    print('Best params is :', logreg cv.best params )
if bestalgorithm == 'SupportVector':
    print('Best params is :', svm cv.best params )
Best model is DecisionTree with a score of 0.8732142857142856
Best params is : {'criterion': 'gini', 'max_depth': 6, 'max_features': 'auto', 'min_samples_leaf': 2, 'min_samples_split': 5, 'splitter': 'random'}
```

Confusion Matrix

Show the confusion matrix of the best performing model with an explanation



 The confusion matrix for the decision tree classifier shows that the classifier can distinguish between the different classes.

• Problem:

- False positive i.e. unsuccessful landing marked as successful landing by the classifier.
- False negative i.e. successful landing marked as unsuccessful landing by classifier

Conclusions

- The larger amount of the flight number at the launch site, the greater the success rate at a launch site will be.
- Orbit ES-L1, GEO, HEO and SSO have 100% success rate at the landing outcomes.
- Success rate is on an increasing trend from 2013 till 2020.
- Low weighted payload mass have largest success rate as compared to high weighted payload mass.
- KSC LC-39A had the most successful launches of any sites.
- The Decision tree classifier is the best machine learning algorithm for this project.

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

