

Winning Space Race with Data Science

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Outline

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

Executive Summary

- Summary of methodologies
- Summary of all results

Introduction

- Project background and context
 SpaceX's website advertises Falcon 9 rocket launches with a cost of 62 million dollars; other providers cost upwards of 165 million dollars for each launch, much of the saving is because SpaceX can reuse the first stage. The objective of the project is to determine the cost of each launch by determining if the first stage will land. This information can be used by an alternate company that wants to bid against SpaceX for a rocket launch.
- Problems you want to find answers
 - What are the factors that contributes the most to the success of a first stage landing?
 - How can machine learning and predictive modelling be used to predict the success of a first stage landing based on historical data?
 - What machine learning algorithms works best in predicting the success of a first stage landing?



Methodology

Executive Summary

- Data collection methodology:
 - The data about the launch details is collected from SpaceX REST API (https://api.spacexdata.com/v4/)
 - The data about Falcon 9 launch records is web scraped from (https://en.wikipedia.org/wiki/List of Falcon 9 and Falcon Heavy launches)
- Perform data wrangling
 - The features of the data is first summarized and analyzed, and then labelled based on findings regarding the success of the landing.

Methodology

Executive Summary

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - The data will be standardized and the split into training and testing sets. Four classification models will be trained and tuned based on the training data using different combinations of parameters.

Data Collection

- Describe how data sets were collected.
 - The datasets was collected from SpaceX API (https://api.spacexdata.com/v4/)
 - From the rocket, the data the booster name was collected. (https://api.spacexdata.com/v4/rockets/)
 - From the launchpad, the data about the launch site being used, the longitude, and the latitude was collected. (https://api.spacexdata.com/v4/launchpads/)
 - From the payload, data about the mass of the payload and the orbit that it is going to was collected. (https://api.spacexdata.com/v4/payloads/)
 - From cores, data about the outcome of the landing, the type of the landing, number of flights with that core, whether grid fins were used, whether the core is reused, whether legs were used, the landing pad used, the block of the core which is a number used to separate version of cores, the number of times this specific core has been reused, and the serial of the core was collected. (https://api.spacexdata.com/v4/cores/)
 - All of the data will complement the rocket launch data collected from past launches (https://api.spacexdata.com/v4/launches/past)
 - A new data frame with columns consist of details of each Falcon 9 launches was created using the collected data.

Data Collection

- Describe how data sets were collected.
 - Additional datasets was also collected from Wikipedia
 (https://en.wikipedia.org/wiki/List of Falcon 9 and Falcon Heavy launches)
 - A data frame consists of records of each Falcon 9 launches were also created during this process by using the data obtained by implementing web scraping techniques.

Data Collection – SpaceX API

- Data collection using SpaceX
 Public API calls were done according to the flowchart beside.
- Notebook Github link:
 https://github.com/BastiaanMatt/IBM-Applied-Data-Science-Capstone/blob/main/jupyter-labs-spacex-data-collection-api.ipynb

Request and parse the SpaceX launch data using the GET request



Filter the data frame to only include Falcon 9 launches

Data Collection - Scraping

 Data collection using web scraping techniques on Falcon 9 launches Wikipedia page were done according to the flowchart beside.

Notebook Github link:
 https://github.com/BastiaanMatt
 /IBM-Applied-Data-Science Capstone/blob/main/jupyter labs-webscraping.ipynb

Request the Falcon9
Launch Wiki page from its
URL



Extract all column/variable names from the HTML table header



Create a data frame by parsing the launch HTML tables

Data Wrangling

- Initial data analysis was done on the filtered data obtained from the SpaceX API. Missing payload mass values are replaced with the mean of the column. Then the number of launches from each site was calculated, also the number and occurrences of each orbit and of each mission outcome is also summarized. Lastly, a landing outcome label was assigned for each launch based on whether the landing was a success or not.
- Notebook Github link: <u>https://github.com/BastiaanMatt/IBM-Applied-Data-Science-Capstone/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb</u>



EDA with SQL

- The following list contains the SQL queries that have been performed in this project:
 - 1. Names of the unique launch sites in the space mission
 - 2. 5 records where launch sites begin with the string 'CCA'
 - 3. The total payload mass carried by boosters launched by NASA (CRS)
 - 4. Average payload mass carried by booster version F9 v1.1
 - 5. The date when the first successful landing outcome in ground pad was achieved.
 - 6. The names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000.
 - 7. Total number of successful and failure mission outcomes
 - 8. Names of the booster_versions which have carried the maximum payload mass.
 - 9. The records which will display the month names, failure landing_outcomes in drone ship ,booster versions, launch_site for the months in year 2015.
 - 10.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.
- Notebook Github link: <u>https://github.com/BastiaanMatt/IBM-Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-sql-coursera_sqllite.ipynb</u>

EDA with Data Visualization

- Exploration with data visualization was done using scatter plots, bar plots and line charts.
- Scatter plots was used to visualize the correlation between the variables (flight number, payload mass, launch site, and orbit type)
- A bar plot was used to visualize the success landing rate of each orbit type.
- A line chart was used to visualize change in the success landing rate over time.
- Notebook Github link: <u>https://github.com/BastiaanMatt/IBM-Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-dataviz.ipynb.jupyterlite.ipynb</u>

Build an Interactive Map with Folium

- The following list contains the map objects that have been created in this project:
 - 1. A set of a circle and marker of each launch sites on the map.
 - 2. A marker cluster of colored markers for each launch result.
 - 3. Lines and markers for calculating the distances between a launch site to its proximities.

Notebook Github link:

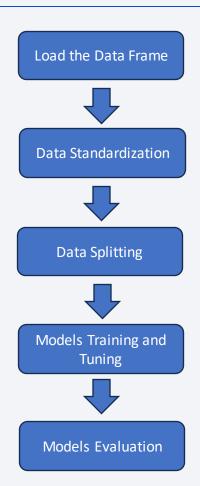
https://github.com/BastiaanMatt/IBM-Applied-Data-Science-Capstone/blob/main/lab_jupyter_launch_site_location.jupyterlite.ipynb

Build a Dashboard with Plotly Dash

- In this project, the dashboard used pie charts and scatter plots to visualize either the summary of all launch sites or for a specific site that can be selected from a dropdown.
- Pie charts were used to visualize the summary of total success launches by site and the success rate for each individual site.
- Scatter plots were used to visualize the correlation between payload and success for all sites and correlation between payload and success for each individual site.
- Notebook Github link: <u>https://github.com/BastiaanMatt/IBM-Applied-Data-Science-Capstone/blob/main/spacex_dash_app.py</u>

Predictive Analysis (Classification)

- After the data frame from the data collection step have been loaded. Features were extracted from the web scraped data frame and the response variable was extracted from 'Class' in the SpaceX API data frame.
- The features then would be standardize before split into training and testing dataset. Models will then be trained and tuned using the standardized features from the training dataset.
- Models that were built in this project includes Logistic Regression, SVM, Decision Trees and KNN.
- After the models have been trained, then the accuracy of each model predictions were tested using the testing dataset. Confusion matrix of each models were created too.
- Notebook Github link: <u>https://github.com/BastiaanMatt/IBM-Applied-Data-Science-Capstone/blob/main/SpaceX_Machine_Learning_Prediction_Part_5.jupyterlite.ipynb</u>



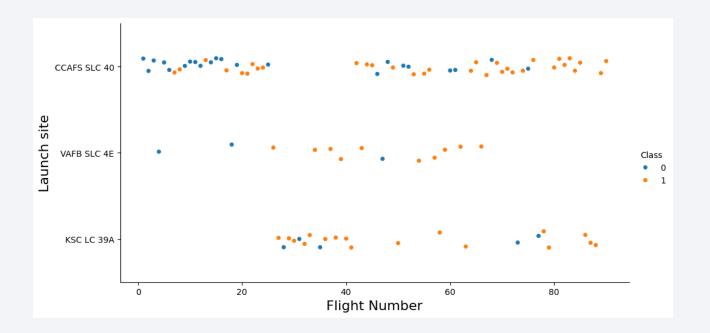
Results

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



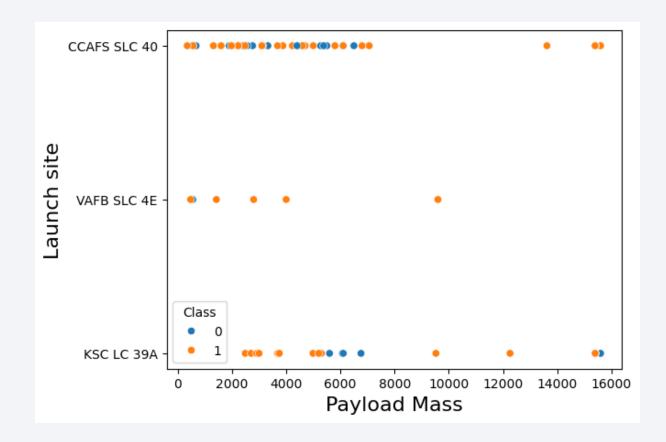
Flight Number vs. Launch Site

- Scatter plot of Flight Number vs. Launch Site.
- Most of the launch were from CCAFS SLC-40.
- Launches with higher flight number have a higher chance of success.



Payload vs. Launch Site

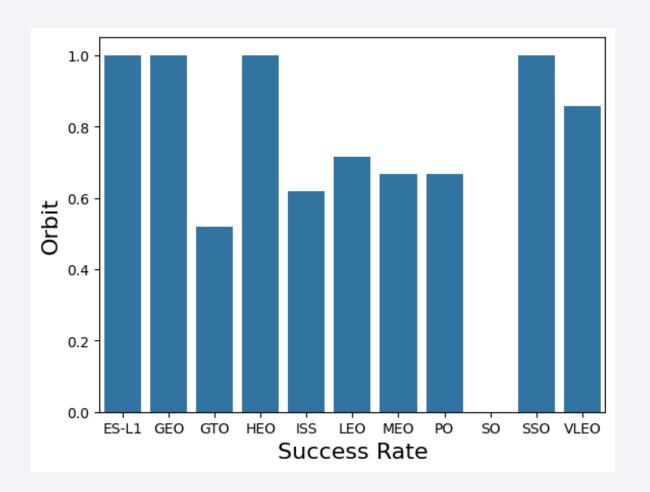
- Scatter plot of Payload vs.
 Launch Site
- There are no rockets launched with a payload mass greater than 10000 for the VAFB-SLC launch site.



Success Rate vs. Orbit Type

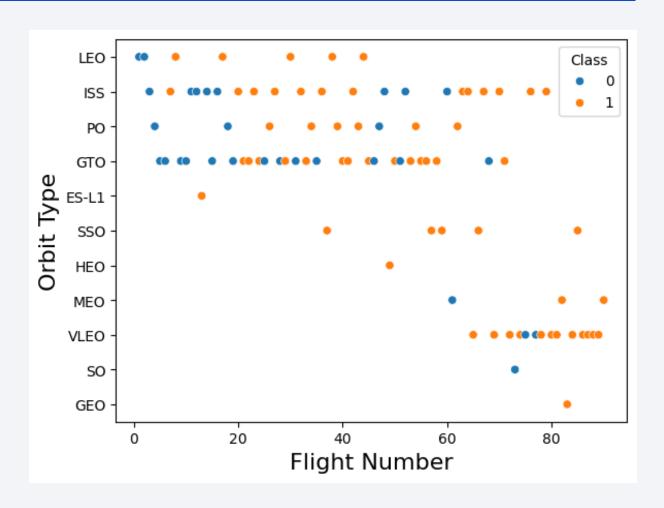
 Bar chart for the success rate of each orbit type

 Launches with the highest success rate were launches with orbit type ES-L1, GEO, ISS, and SSO.



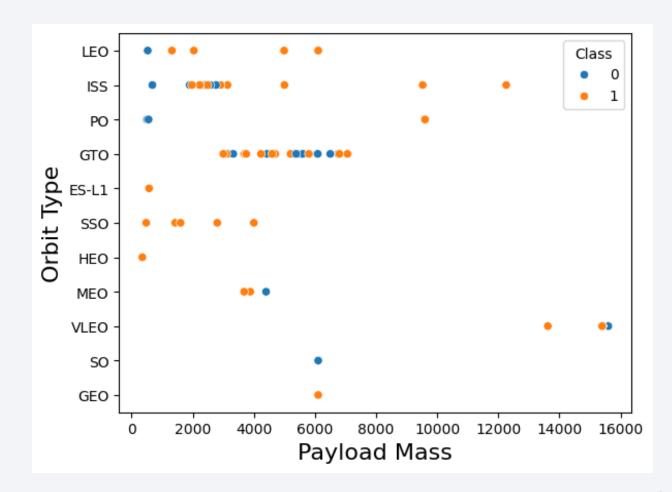
Flight Number vs. Orbit Type

- Scatter point of Flight number vs. Orbit type
- LEO orbit launches successes seems to be related with the flight numbers.
- However GTO orbit launches successes seems to be unrelated to the flight numbers.



Payload vs. Orbit Type

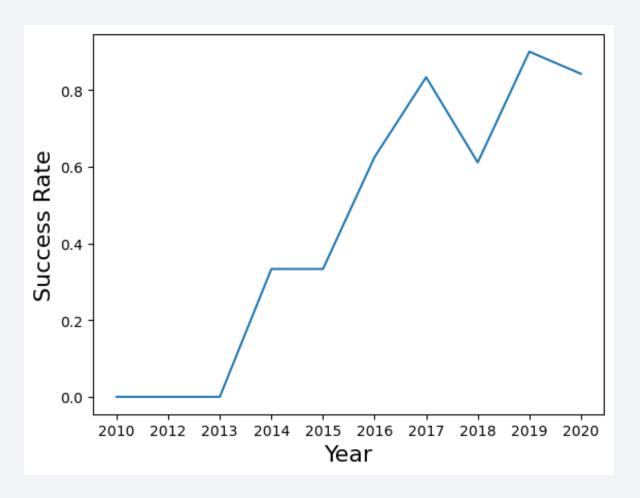
- Show a scatter point of payload vs. orbit type
- Orbit type PO, LEO and ISS seems to have more successful landings for launches with heavy payloads.
- However GTO orbit launches successes seems to have no relationship with payload mass.



Launch Success Yearly Trend

• Line chart of yearly average success rate

• The success landing rate for rocket launches increases from 2010 until 2020.

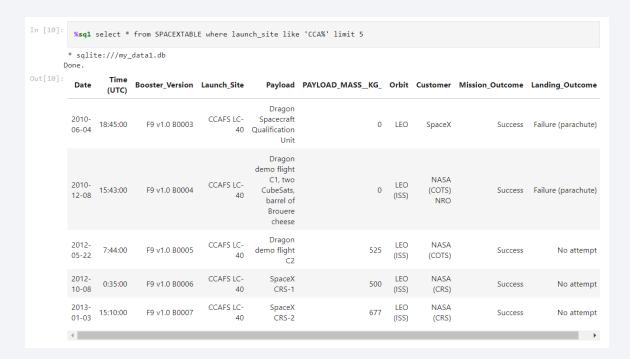


All Launch Site Names

- The names of the unique launch sites are as follows:
 - CCAFS LC-40
 - VAFB SLC-4E
 - KSC LC-39A
 - CCAFS SLC-40
- The query used the **distinct** command line to avoid double entries and provide a list of unique launch site names.

Launch Site Names Begin with 'CCA'

- 5 records where launch sites begin with 'CCA' are displayed beside.
- The first customers of the launches were from SpaceX itself and NASA.
- The query retrieve 5 records that have launch sites begins with 'CCA'.



Total Payload Mass

- The total payload carried by boosters from NASA (CRS) was 45596 kgs.
- The query used an aggregate function **sum** to sum all of the payload carried by the boosters with NASA (CRS) as the customers.

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

**sql select sum(PAYLOAD_MASS__KG_) as sum_payload_mass from SPACEXTABLE where customer ='NASA (CRS)'

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

Out[11]: sum_payload_mass

45596
```

Average Payload Mass by F9 v1.1

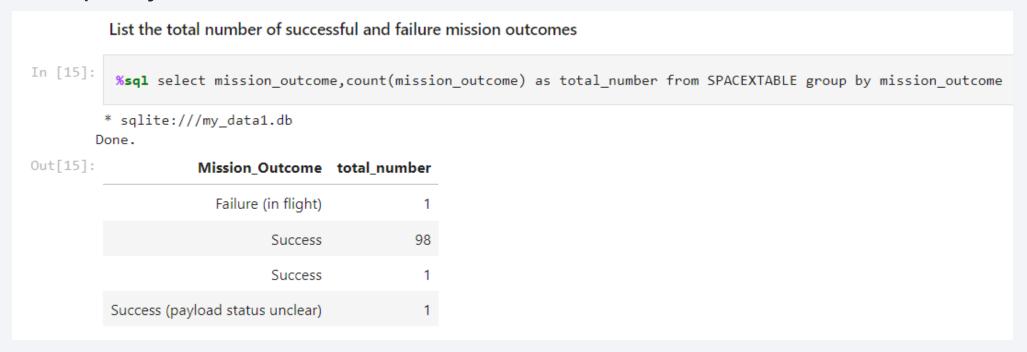
- The average payload mass carried by booster version F9 v1.1 was 2534.67 kgs.
- The query used the aggregate function **avg** to retrieve the mean of the payload mass on booster versions that starts with 'F9 v1.1'.

Successful Drone Ship Landing with Payload between 4000 and 6000

- The names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000 were all starts with 'F9 FT'.
- The query retrieve the booster versions of all landing outcome that contains the words 'success' and 'drone' that have a payload mass between 4000 and 6000.

Total Number of Successful and Failure Mission Outcomes

- There were 100 successful outcomes and 1 failure mission outcome.
- The query groups the records by mission outcomes and retrieves the frequency of each mission outcomes.

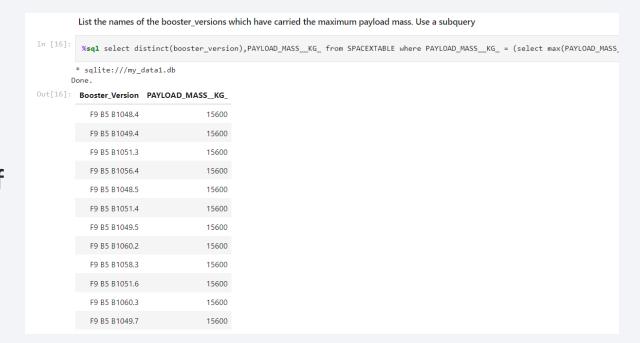


First Successful Ground Landing Date

- The date of the first successful landing outcome on ground pad was on December 22 2015
- The query use the aggregate function min to retrieve the earliest record where the landing outcome contains the word 'ground'.

Boosters Carried Maximum Payload

- The names of the booster which have carried the maximum payload mass all starts with 'F9 B5', the maximum payload mass was 15600 kgs.
- The query retrieves the unique names of booster versions that have carried payloads of 15600 kgs.

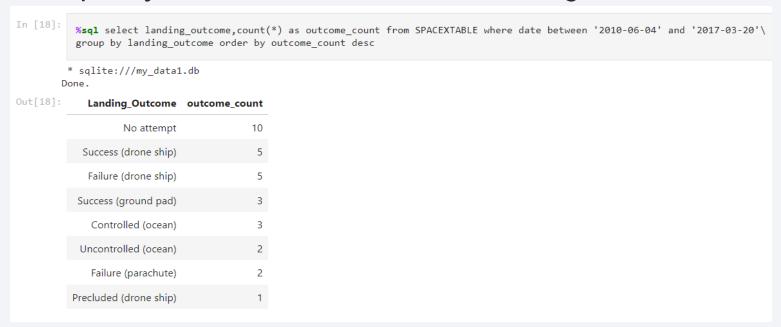


2015 Launch Records

- There are 2 records in 2015 with a landing outcome of 'Failure (drone ship)'. Both have were launched from the same site with 2 different booster versions at 2 different months.
- The query used **substr** to retrieves months and year of the records.

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

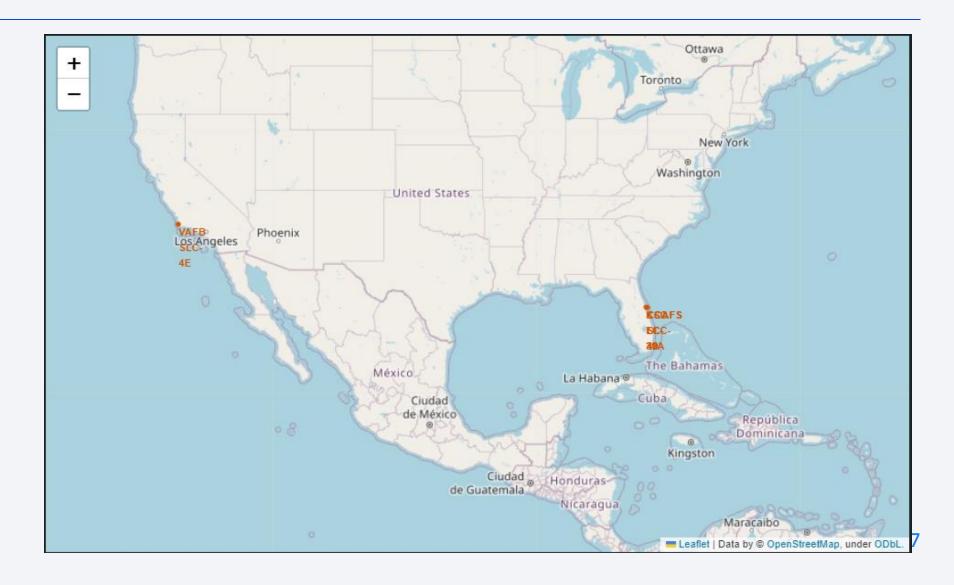
- Between the date 2010-06-04 and 2017-03-20, the landing outcome with the highest frequency were records that have not attempt to land.
- The query groups the records by landing outcomes and orders them by the frequency of each outcome in a descending manner.





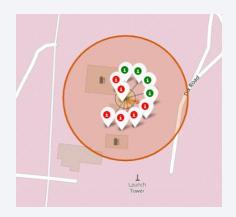
All Launch Sites Location Markers

All of SpaceX launch sites are located on USA coasts near Los Angeles and Florida.

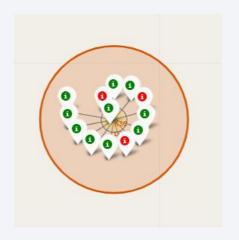


Success/Failed Launches for Each Site Markers

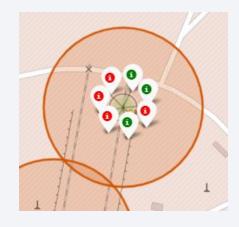
VAFB SLC-4E



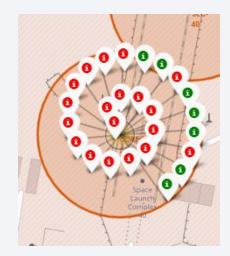
KSC LC-39A



CCAFS SLC-40



CCAFS LC-40



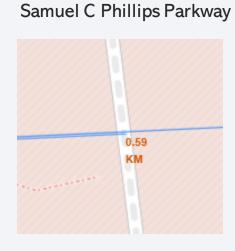
Where **green** markers indicate successful landings and **red** markers indicate unsuccessful landings.

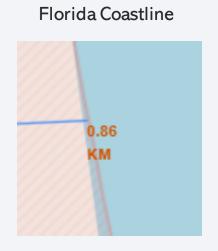
Distance between a launch site to its Proximities

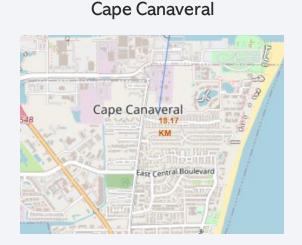
• Launch site CCAFS SLC-40 to its proximities:

- Are launch sites in close proximity to railways? Yes, the closest railway is at 1.30 km away from the launch site.
- Are launch sites in close proximity to highways? Yes, the closest highway is at 0.59 km away from the launch site.
- Are launch sites in close proximity to coastline? Yes, the closest coastline is at 0.86 km away from the launch site.
- Do launch sites keep certain distance away from cities? Yes, the closest city is at 18.17 km away from the launch site.

Nasa Railroad

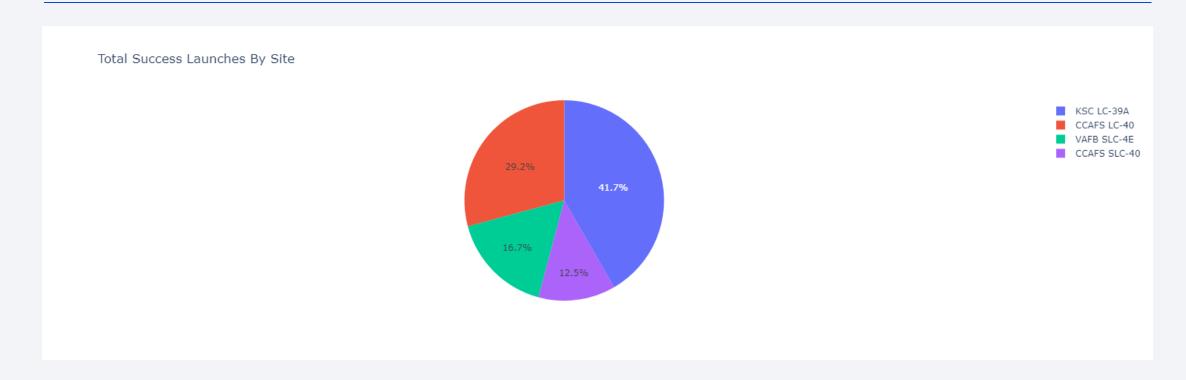






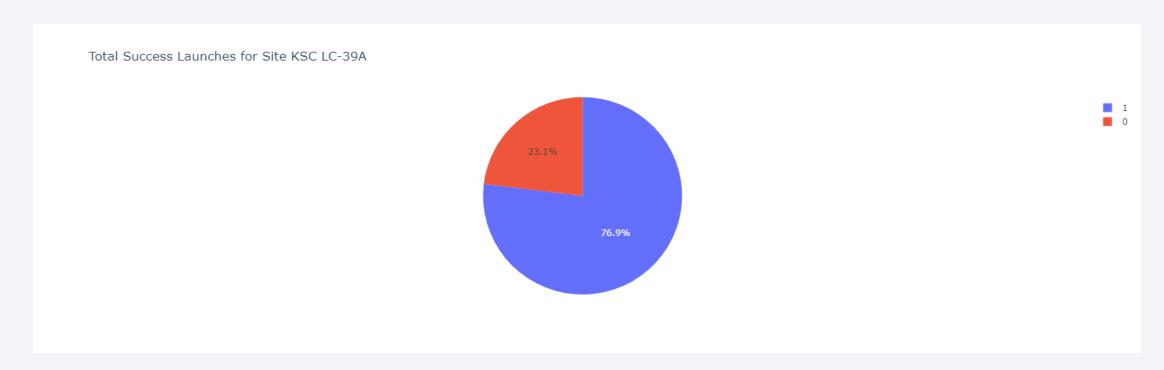


Total Success Launches By Site



• The launch site KSC LC-39A have the highest number of successful launches at 41.7% out of total successful launches from all sites.

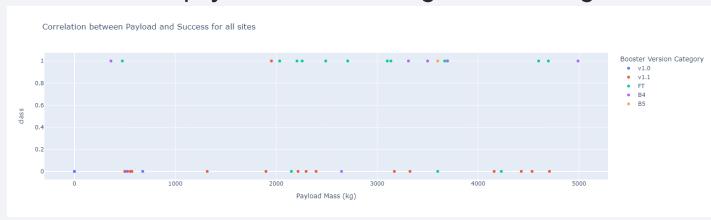
Launch Site with Highest Launch Success Ratio



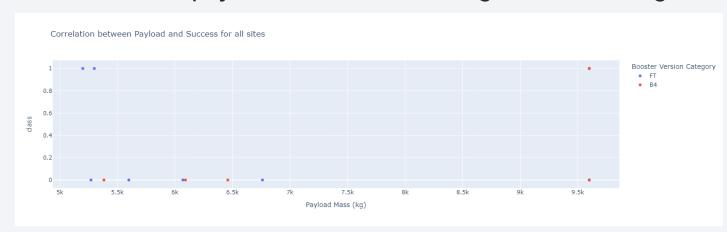
• Not only the launch site KSC LC-39A have the highest number of successful launches, it also have the highest launch success ratio of 76.9%.

Payload vs. Launch Outcome

Launches with payload between 0 kg and 5000 kgs.



• Launches with payload between 5000 kg and 10000 kgs.

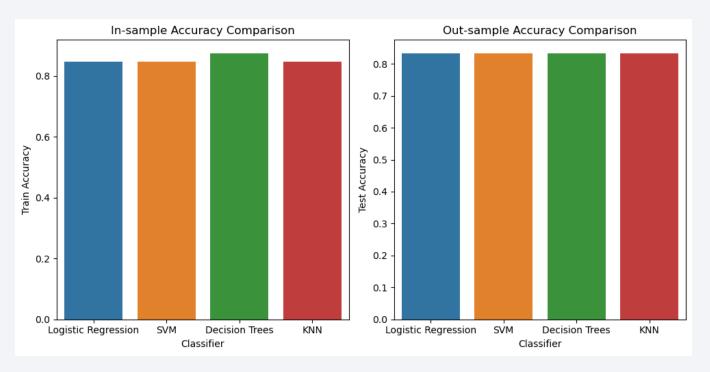


Launches with heavier
 payloads have lower launch
 success ratio compared to the
 ones with lighter payload
 mass.



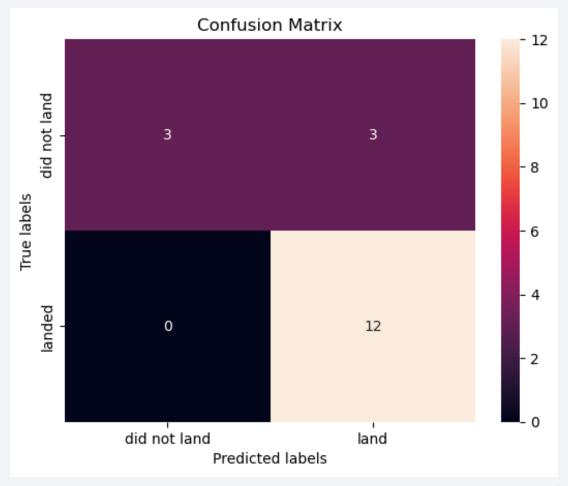
Classification Accuracy

- The Decision Trees model have the best performance for insample (training) accuracy at 88.75%.
- All models have the same performance for out-sample (testing) accuracy at 83.33%.



Confusion Matrix

- As all of the models have similar performance in out-sample accuracy score. All of the models also produced a similar Confusion Matrix.
- One thing to consider is the false positive error where there are 3 launch records where the landing was predicted as a successful land when in reality it was an unsuccessful land.



Conclusions

- Launches with higher flight number have a higher chance of success.
- The success landing rate for rocket launches increases from 2010 until 2020.
- Launches with the highest success rate were launches with orbit type ES-L1, GEO, ISS, and SSO.
- Launches with **heavier payloads** have **lower launch success ratio** compared to the ones with lighter payload mass.
- Not only the launch site KSC LC-39A have the highest number of successful launches at 41.7% out of total successful launches from all sites, it also have the highest launch success ratio of 76.9%.
- Out of the 4 classification models tested in this project, the were **no best** model that results in the best out-sample accuracy compared to others at **83.33%**. However, the decision trees classification models performed the best in the in-sample accuracy test at **88.75%**.

