

Winning Space Race with Data Science

Rui Azevedo 2022-05-18



Outline

- Executive Summary
- Introduction
- Methodology
- Results
- Conclusion

Executive Summary

- We set out to determine what factors will predict whether SpaceX will land the first stage of the Falcon 9 rocket, and to train a model that could predict successful landings
- We used Python to obtain and visualize launch data from the SpaceX API and through webscraping, and to train machine learning (ML) models
- We found that the landing success rate increased over time (consistent with introductions of improved rocket versions), and was dependent on payload mass, and target orbit
- We trained four ML models (KNN, logistic regression, decision tree, SVM), obtaining a best accuracy of 83%

Introduction

- Most rocket manufacturers have, or are working on, reusable rockets
- Companies that have this option, like SpaceX, can offer much better prices than the competition
- The demand for launch vehicles is still extremely high!
- Landing a rocket booster is hard and can only be done under certain conditions
- Can we determine what these conditions are?
- Can we predict whether a company will land a booster, under some launch conditions?



Methodology

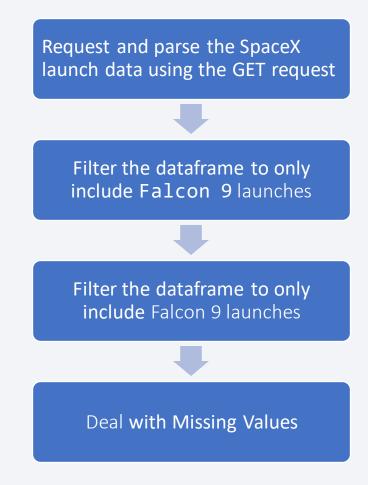
Executive Summary

- Data was collected via the SpaceX API and through webscraping, using Python
- Performed data wrangling using Pandas and Numpy, preparing data for visualization and model training
- Perform exploratory data analysis (EDA) using the Python plotting packages matplotlib and seaborn, and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models in SciKit Learn (KNN, logistic regression, decision tree, SVM)

Data Collection – SpaceX API

- The Python "Requests" package simplifies making calling API's
- Pandas was used to clean up the data, removing unwanted values and features

API Calls Notebook



Data Collection - Scraping

- Requests package Was used to get the HTML source from Wikipedia
- Beautiful Soup package was used to parse the HTML
- Pandas and custom functions ere used to convert the relevant columns into a dataframe

Webscrapping Notebook

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

 We obtained the number of launches per orbit type, of which the most common were

- GTO 27
- ISS 21
- VLEO 14

And the number of launches from each launch site:

- CCAFS SLC 40 55
- KSC LC 39A 22
- VAFB SLC 4E 13

• The several landing outcomes were into two binary categories:

- Successes: 60
- Failures: 30

Data Wrangling Notebook

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

EDA with Data Visualization

- We used the Seaborn package to plot the outcome of the landing attempts against some of the features of the data set
- We also tried to determine correlations between the features themselves, and how they affect landing outcome
- The feature with the greatest direct impact experience! SpaceX was able to learn from failure and the success rate steadily improve since the first success in 2014

EDA with Data Visualization notebook

EDA with SQL

We performed the following queries in SQL, using the package "sqlalchemy" on a IBM DB2 database:

- Displayed the names of the unique launch sites
- Displayed 5 records where launch sites begin with the string 'CCA'
- Displayed the total payload mass carried by boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1
- Listed the date when the first successful landing outcome in ground pad was achieved.
- Listed the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

- Listed the total number of successful and failure mission outcomes
- Listed the names of the booster_versions which have carried the maximum payload mass. Use a subquery
- Listed the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- Ranked 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
- EDA with SQL notebook

Build an Interactive Map with Folium

- We added the launch sites to an interactive map, as well, as clusters or markers for the launches from each one
- It is clear that launch sites are located near the coast and away from population centres, where the expended stages of rockets or debris will not fall on any populated areas.
- There are, however, highways and railways nearby, but these are relatively easy to clear for launches

Interactive map with Folium

Build a Dashboard with Plotly Dash

- We created interactive chart, in order to visualize the relations each launch site, payload mass, and the number of successful landings
- A pie chart that displays the number of successful landings from each site, or the ratio between successes and failures in each site
- An scatter plot that (for each or all sites) displays the payload mass in a selectable range and the corresponding booster version, and if the landing was successful or not.

Plotly Dash lab

Predictive Analysis (Classification)

- We used SciKitLearn to perform classification analysis using Decision Trees, KNN, Logistic Regression and SVM
- In order to streamline the process and ensure that the data is scaled correctly, we build a pipeline that included both scaling and model training
- The grid search for hyperparameters and model training was then performed on the pipeline
 - This ensures, for example, that different scalings in train and test sets do not distort our performance metrics

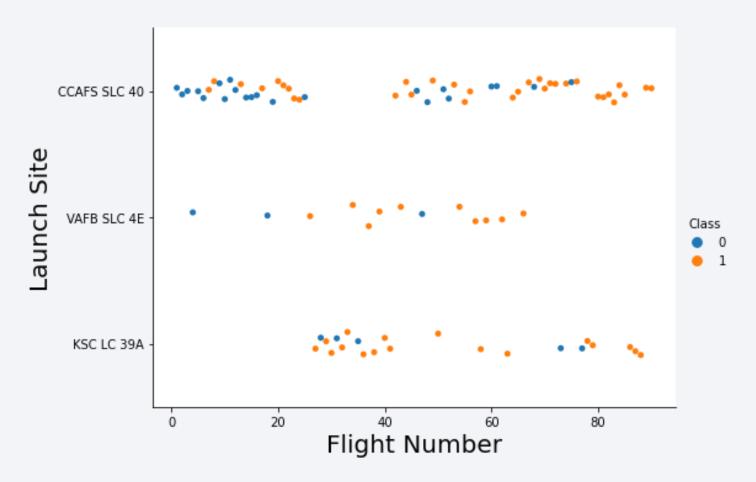
Completed predictive analysis lab

Split data into test/train set Build scaling + modeling pipeline Gridsearch crossvalidation for hyperparamenters Performance metrics



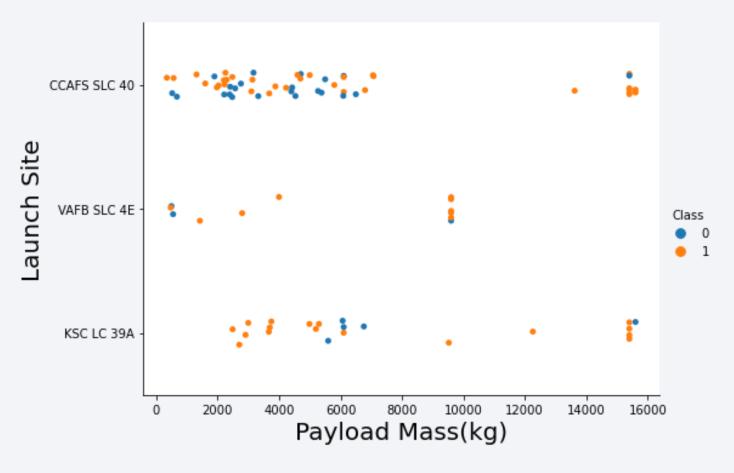
Flight Number vs. Launch Site

- SLC 40 at Cape Canaveral was the initial launch site for SpaceX, and still remains the favoured on
- It is natural that it has a lower average success rate, since it encompasses most of the early launches



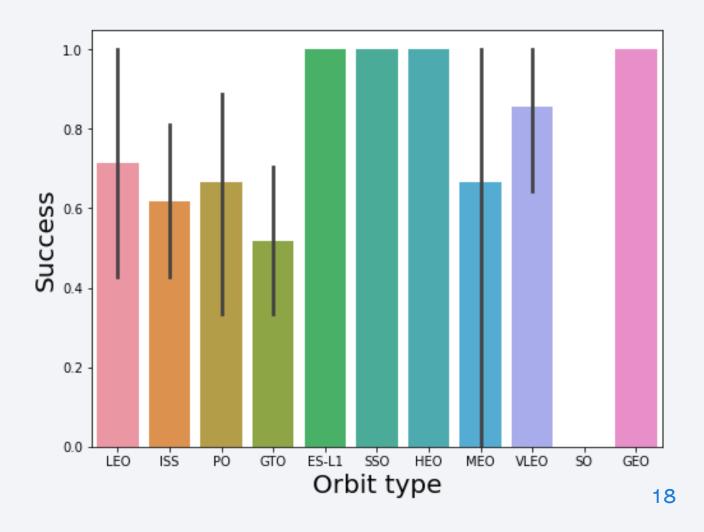
Payload vs. Launch Site

 There seems to be little correlation between payload mass and launch site beyond the launches at 9500kg from Vanderburg Air Force Base.



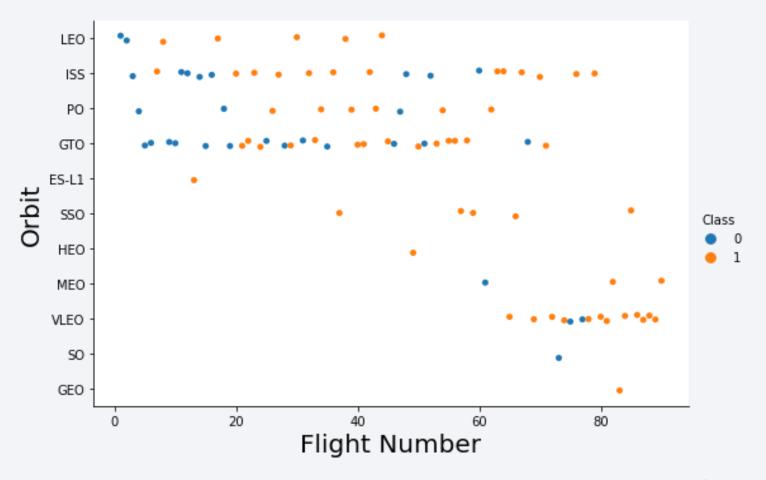
Success Rate vs. Orbit Type

- Success rate depends on the target orbit
- VLEO has higher average success and lower variance, while e.g. launches to the ISS have lower success rate and higher variance
- No meaningful conclusions for ES-L1, SSO, HEO, SO and MEO due to few or no launches into these orbits



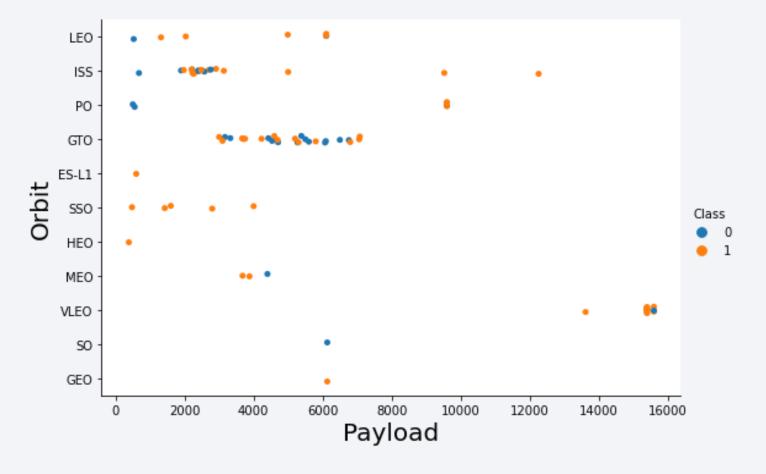
Flight Number vs. Orbit Type

- SpaceX has diversified its target orbits as they gain more experience
- The appearance of VLEO somewhat coincides with the first launches of Starlink, but the dissapearence of LEO might also suggest a reclassification of target orbits



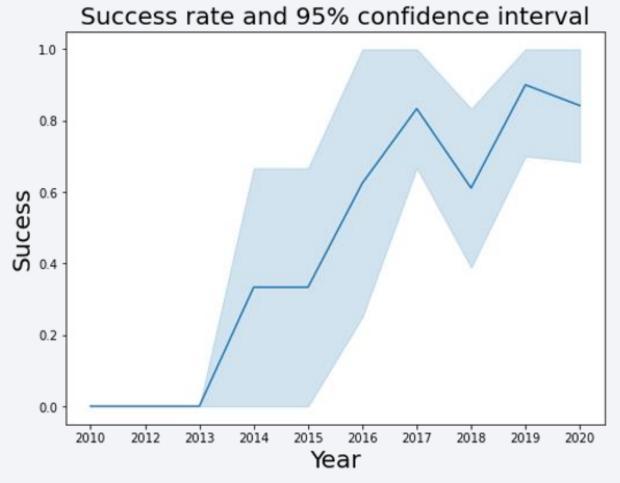
Payload vs. Orbit Type

Orbit type influences
 maximum payload: orbits
 with higher altitude or
 inclination will require more
 energy, and therefor allow a
 lower payload mass.



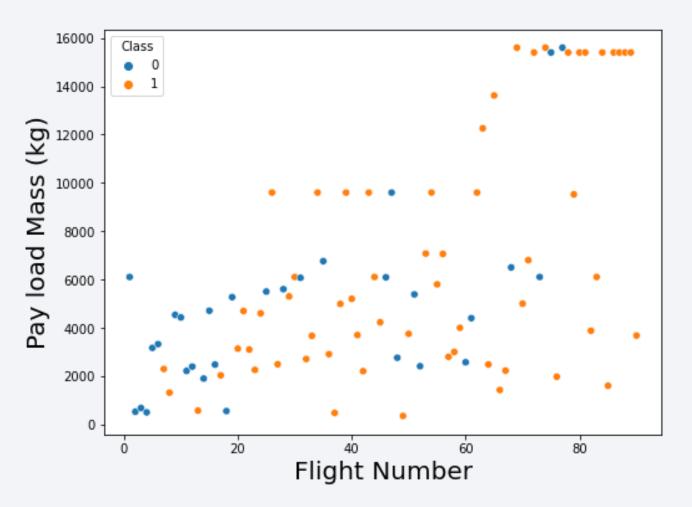
Launch Success Yearly Trend

- •The feature with the greatest direct impact experience! SpaceX was able to learn from failure and the success rate steadily improve since the first success in 2014.
- •The variance of the success rate also shrinks with time
- •This is probably due to improvements in both software and hardware



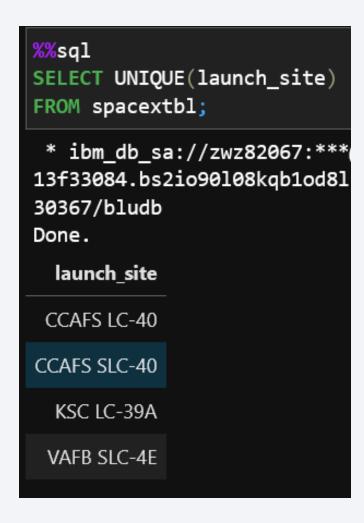
Flight number & Payload mass

- It is also possible to see that with experience came a higher average payload mass...
-but also a higher success rate for the same payload mass



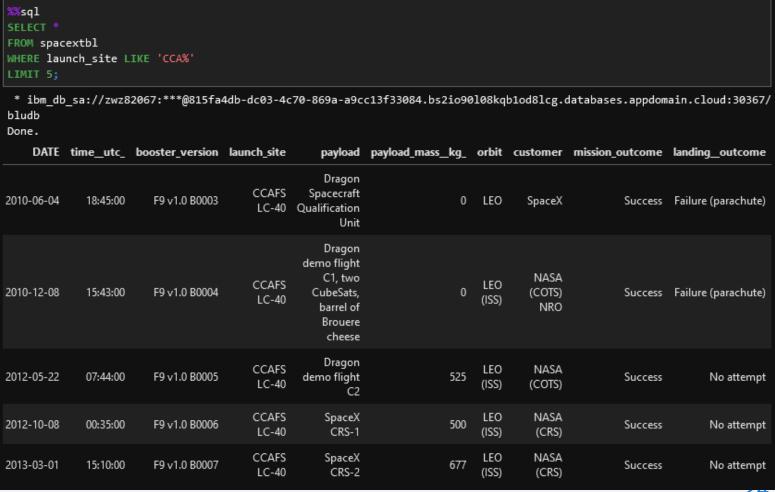
All Launch Site Names

 Find the names of the unique launch sites



Launch Site Names Begin with 'CCA'

Find 5 records
 where launch sites
 begin with `CCA`



Total Payload Mass

 Calculate the total payload carried by boosters from NASA

```
8%sq1
SELECT SUM(payload_mass__kg_) AS
FROM spacextbl
WHERE customer LIKE 'NASA (CRS)%';
 * ibm_db_sa://zwz82067:***@815fa4d
bludb
Done.
48213
```

Average Payload Mass by F9 v1.1

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

```
%%sq1
SELECT AVG(payload_mass_kg_) as avg_payload_mass
FROM spacextbl
WHERE booster_version LIKE 'F9 v1.1%';
 * ibm_db_sa://zwz82067:***@815fa4db-dc03-4c70-869
bludb
Done.
avg_payload_mass
           2534
```

First Successful Ground Landing Date

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

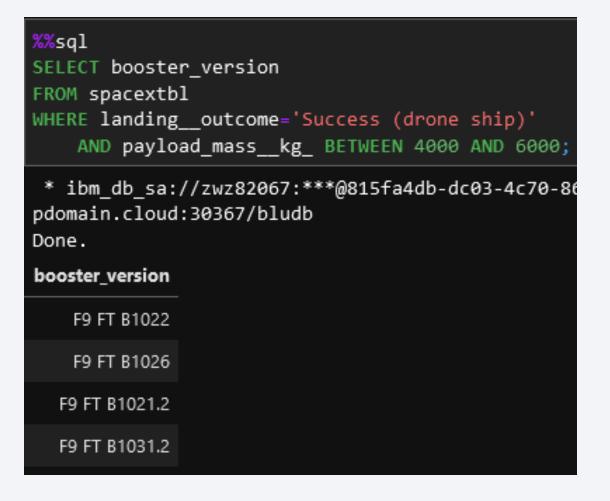
```
%%sql
SELECT MIN(date) as date_first_landing
FROM spacextbl
WHERE landing__outcome='Success (ground pad)';

* ibm_db_sa://zwz82067:***@815fa4db-dc03-4c70
pdomain.cloud:30367/bludb
Done.
date_first_landing

2015-12-22
```

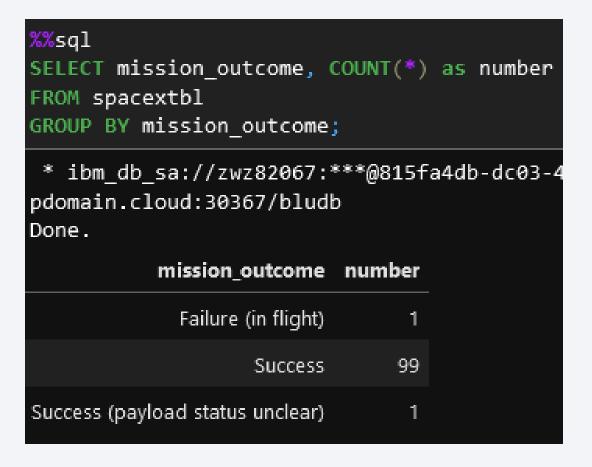
Successful Drone Ship Landing with Payload between 4000 and 6000

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



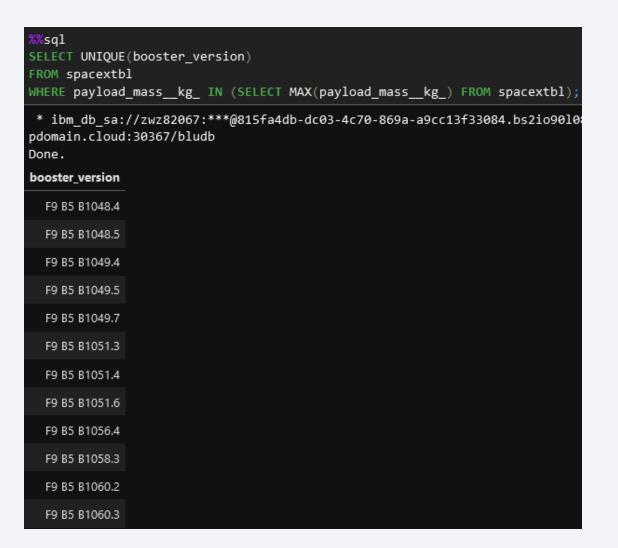
Total Number of Successful and Failure Mission Outcomes

 Calculate the total number of successful and failure mission outcomes



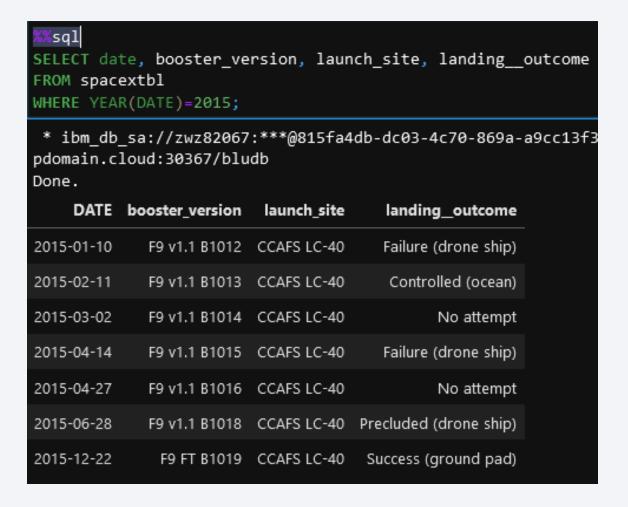
Boosters Carried Maximum Payload

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



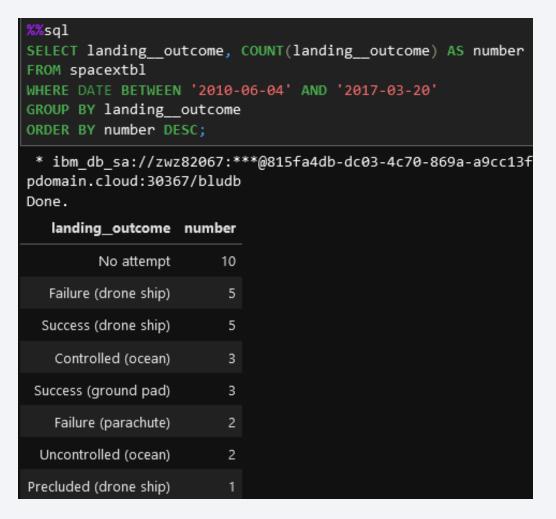
2015 Launch Records

 List the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015



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

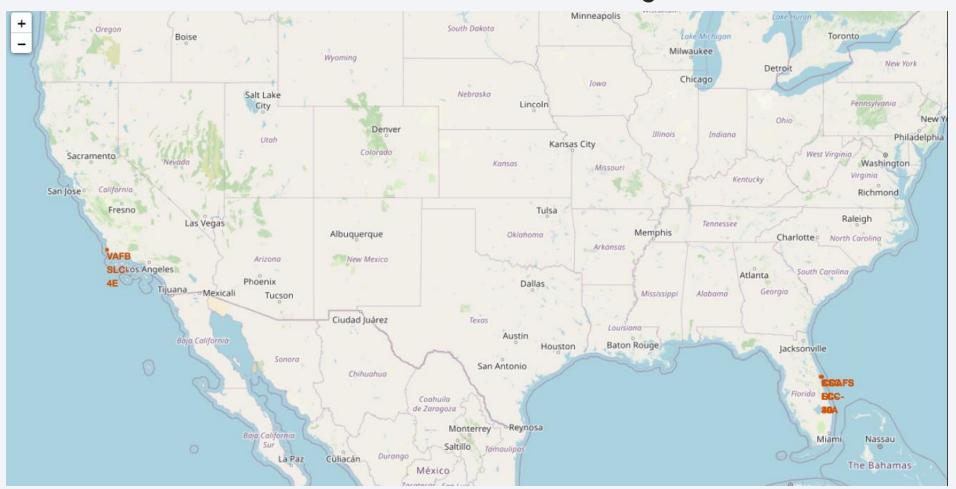
 Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order





SpaceX launch site

• SpaceX launch site are all near the coast and at low longitudes.



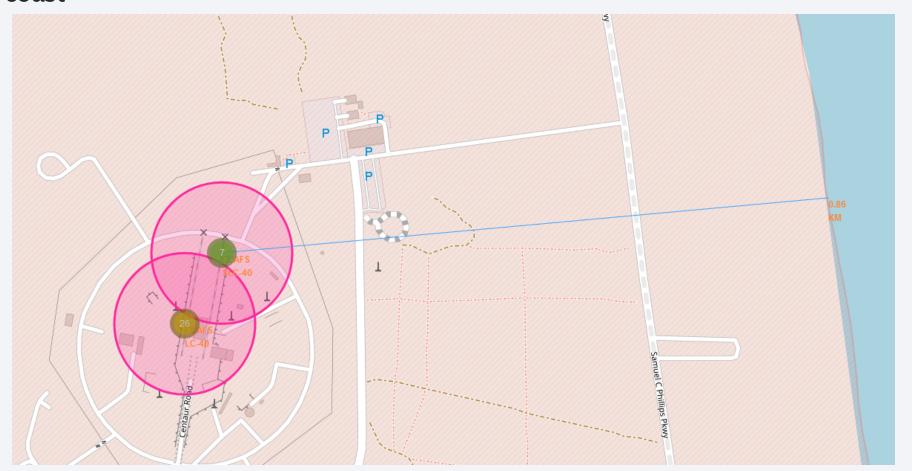
Launches from Vandenberg Space Force Base

- We can see the launches from VSFB (VAFB in the data), green for successful landings, and red for unsuccessful landings.
- VSFB accounts for the lowest success/fail ratio for landings



Distance to coastline

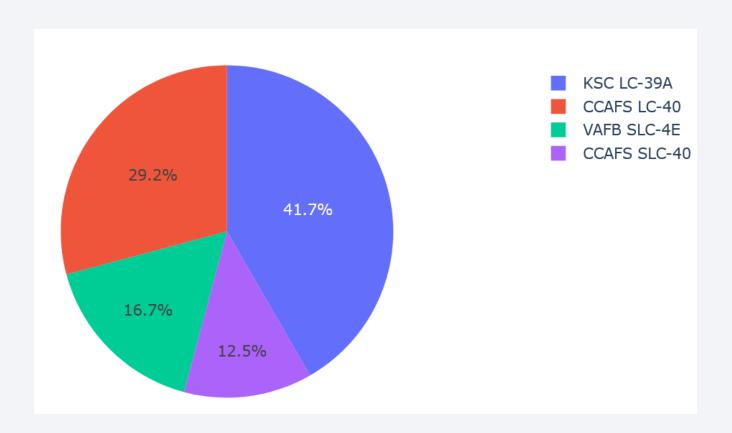
 We can see that the launch sites in Cape Canaveral are less than 1km from the coast





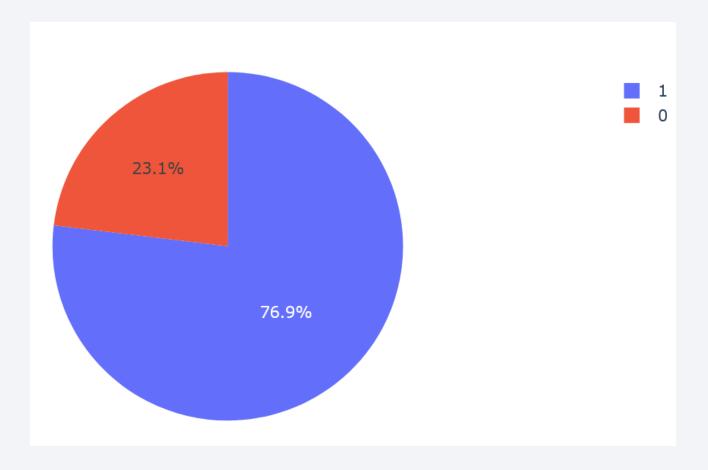
Launch success count for all sites

 We can see that almost half of the successful Landings come from KSC LC-39A



Successful Launches for site KSC LC-39A

• Site KSC LC-39A accounts for 10 successes and 3 failures



Payload vs. Launch Outcome

- We can see that heavy launches are exclusively done with the FT or B4 boosters, while all of them are used for lighter payloads.
- Heavier payloads seem to have a smaller success rate

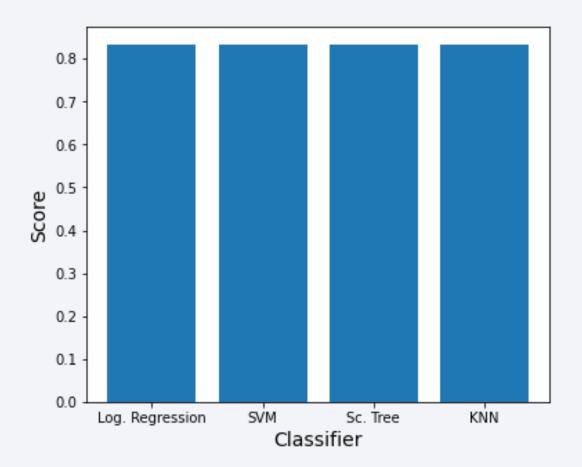






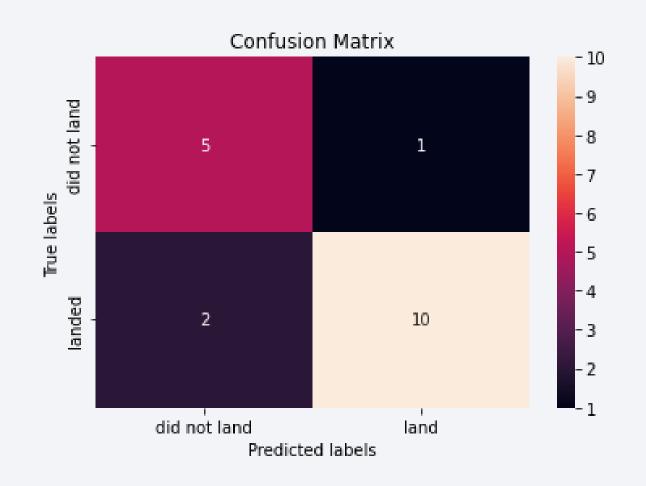
Classification Accuracy

 All classifiers achieved the same accuracy, at around 83.3%



Confusion Matrix – Decision Tree

- The classifier using logistic regression was able to correctly predict 10 successful landings and 5 failures
- However, it misclassified one failure as a success as 2 successes as failures



Conclusions

- Reusable rocket boosters are the future of the space industry
- Being able to predicted when other companies (like SpaceX) land their boosters can help a newcomer to the business decide which markets to target
- For example, we could develop a better booster than SpaceX to outperform them in a certain sector (payload mass range, orbit, launch site, etc.)
- Our models can predict whether SpaceX will land a booster with 83% accuracy
- Future models could improve this performance by taking other factors into account, such as weather conditions

