

Winning Space Race with Data Science

Deogracious MUTYABA 14th January 2024



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Executive Summary

Summary of methodologies

- Data collection
- Data wrangling
- EDA with data visualization
- EDA with SQL
- Building an interactive map with Folium
- Building a Dashboard with Plotly Dash
- Predictive analysis (Classification)

Summary of all results

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

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 and then inform the commercial viability of Space X.

Questions to be answered

- 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:
 - Make requests to the SpaceX API.
 - Perform web scraping to collect Falcon 9 historical launch records on the Wikipedia page titled: <u>List of Falcon</u>
 9 and Falcon Heavy launches
- · Perform data wrangling
 - Clean the data and explore it to find patterns in the data to determine the labels for training supervised models.
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Create a machine learning pipeline to predict if the first stage will land given the data.
 - Train the best performing model to make accurate predictions.

Data Collection

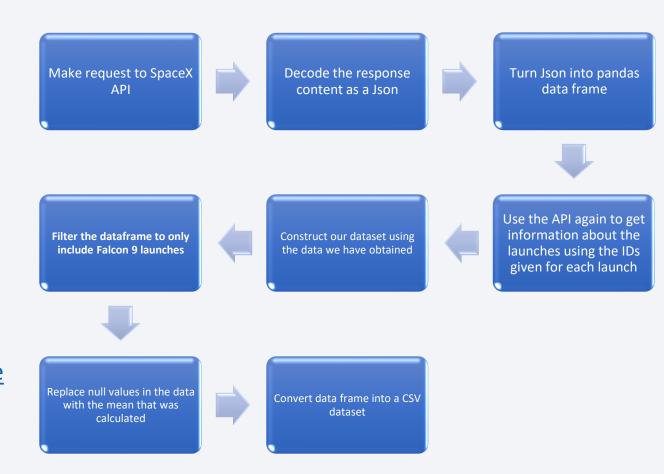
The data collection stage is a key stage in the project. This is because it provides the raw data that will be used at all other stages in this prediction study.

Two methods are used of data collection are used for this prediction study:

- Calling the SpaceX API.
- Data collection by Web Scraping

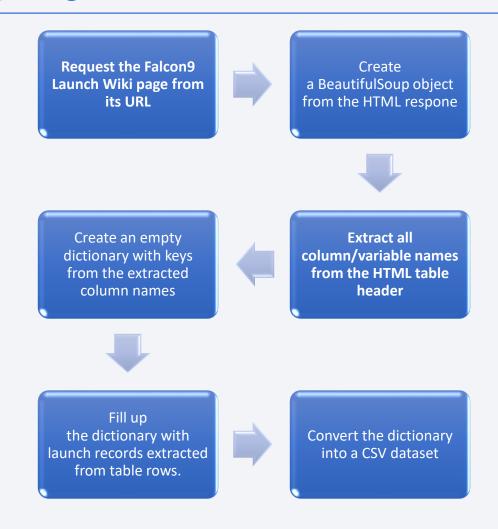
Data Collection – SpaceX API

- Make a request to SpaceX API and make sure the data is in the correct format.
- Perform some basic data wrangling and formatting in order to clean the requested data.
- Convert our data frame into a CSV dataset.
- URL link:
 https://github.com/ezramutyaba/Course
 ra-Data-Science Training/blob/main/Capstone/jupyter labs-spacex-data-collection-api.ipynb



Data Collection - Web Scraping

- Using BeautifulSoup, perform web scraping on the wikipedia page with title: <u>List of</u> <u>Falcon 9 and Falcon Heavy launches</u>
- Store the launch records in an HTML table.
- Parse the table and convert it into a CSV dataset.
- URL link:
 https://github.com/ezramutyaba/Coursera Data-Science Training/blob/main/Capstone/jupyter-labs webscraping.ipynb



Data Wrangling

The goal in this stage is to find patterns in the data and determine the label for training supervised machine learning models.

In the data set, there are several different cases where the rocket did not land successfully. For example, *True RTLS* means the rocket successfully landed on a ground pad while *False RTLS* means the rocket unsuccessfully landed on a ground pad.

Those outcomes were converted into Training Labels whereby 1 means the rocket landed successfully while 0 means it was unsuccessful.

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Determine the number of launches on each facility Calculate the number and occurrences of each orbit Calculate the number and occurences of mission outcome per orbit type Create a landing outcome training label and loop through all the landing outcomes Create a "Class" column containing the information from the outcome label Transform the data frame into a CSV dataset.

EDA with Data Visualization

Data visualization helps us understand data by curating it into a form that's easier to understand, highlighting the trends and outliers. Several types of charts were used in the visualization of the data:

- Cat plots and scatter plots were used to view the relationships of categorical variables like *Launch Site* and *Orbit*.
- A bar chart was used to visualize the success rate of each orbit type.
- A line chart was used to visualize the launch success yearly trend.

URL link: https://github.com/ezramutyaba/Coursera-Data-Science-
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EDA with SQL

Summary of SQL queries that were used:

- Display the names of the unique launch sites in the space mission
- Compare the payload mass with boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1
- List the total number of successful and failure mission outcomes
- Determine the dates of different landing outcomes

URL link: <u>Coursera-Data-Science-Training/Capstone/jupyter-labs-eda-sql-coursera-sqllite.ipynb at main · ezramutyaba/Coursera-Data-Science-Training · GitHub</u>

Build an Interactive Map with Folium

- Folium Markers were used to show the SpaceX launch sites and their nearest important landmarks like railways, highways, cities and coastlines.
- Polylines were used to connect the launch sites to their nearest landmarks.
- Furthermore, Folium Circles were used to highlight circle area of launch sites.
- In order to mark the success/failed launches for each site, marker clusters were used on the map. Whereby Red represents rocket launch failures while Green represents the successes.
- URL link: https://github.com/ezramutyaba/Coursera-Data-Science-
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Build a Dashboard with Plotly Dash

- Pie charts and scatter charts were used to visualize the launch records of SpaceX.
- These charts displayed the rocket launch success rate per launch site. We are
 were able to get an understanding of the factors that may have been
 influencing the success rate at each site. Such as the payload mass and
 booster versions.
- Successful launches were represented by 1 while failures were represented by
 O.
- URL link: https://github.com/ezramutyaba/Coursera-Data-Science-Training/blob/main/Capstone/spacex dash app.py

Predictive Analysis (Classification)

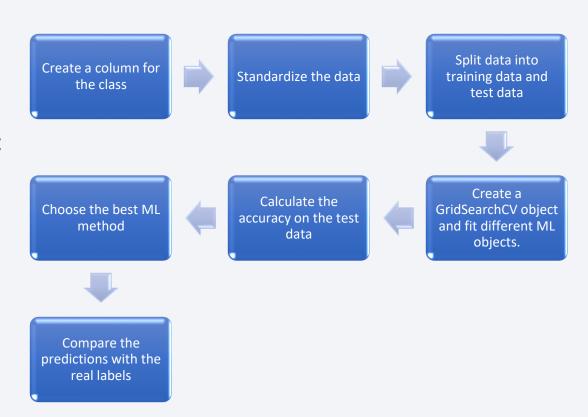
Scikit-learn is the primary ML(machine learning) library that was used for predictive analysis. The following took place:

- Created a machine learning pipeline to predict if the first stage will land given the data.
- Using GridSearchCV, found the best ML method for predictions.
- Compared the predictions with the real labels.
- The ML model scored an accuracy of 83.33%

URL link:

https://github.com/ezramutyaba/Coursera-Data-Science-

<u>Training/blob/main/Capstone/SpaceX Machine</u> <u>Learning Prediction Part 5.jupyterlite.ipynb</u>



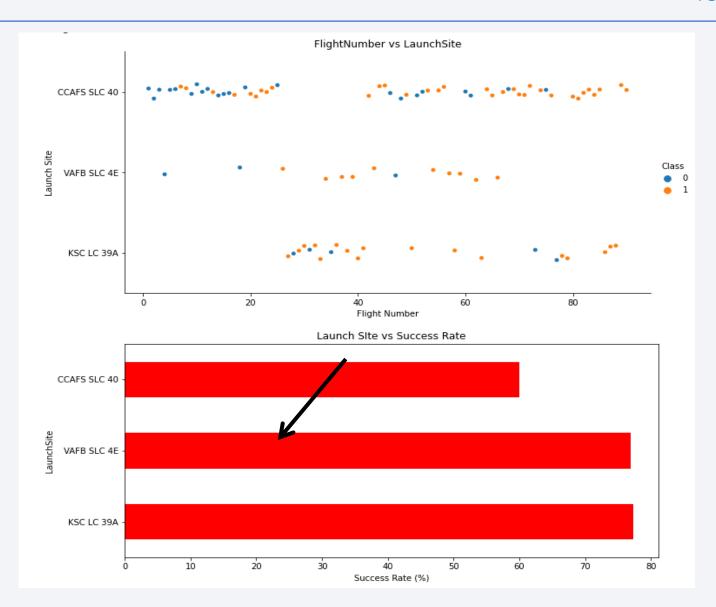
Results

- The exploratory data analysis has shown us that successful landing outcomes are somewhat correlated with flight number. It was also apparent that successful landing outcomes have had a significant increase since the year 2015.
- All launch sites are located near the coastline. Perhaps, this makes it easier to test rocket landings in the water.
- Furthermore, the sites are also located near highways and railways. This may facilitate transportation of equipment and research material.
- The machine learning models that were built, were able to predict the landing success of rockets with an accuracy score of 83.33%. This accuracy can be increased in future projects with more data.

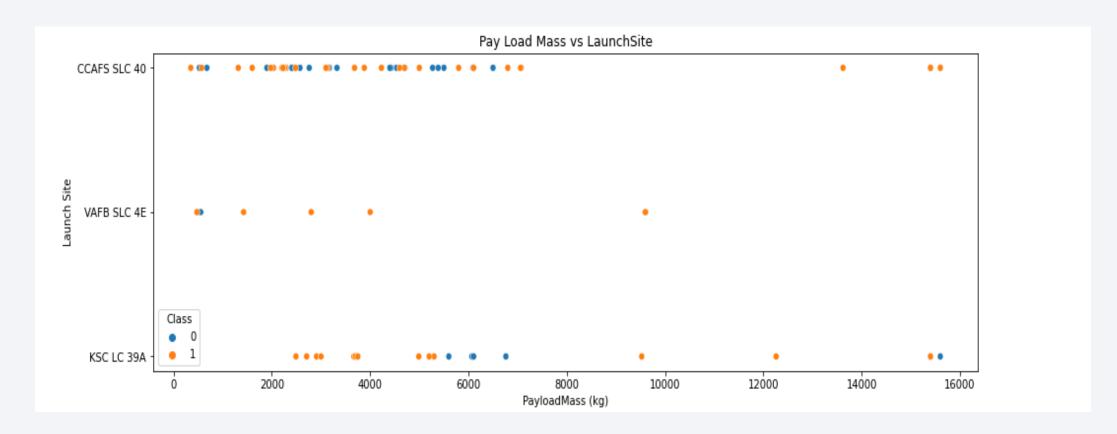


Flight Number vs. Launch Site

- It appears that there were more successful landings as the flight numbers increased. It also seems that launch site CCAFS SLC 40 had the greatest number of landing attempts while the site VAFB SLC 4E had the least number of attempts.
- Looking at the second chart, we can see that there is no Launch Site with a success rate below 60%.



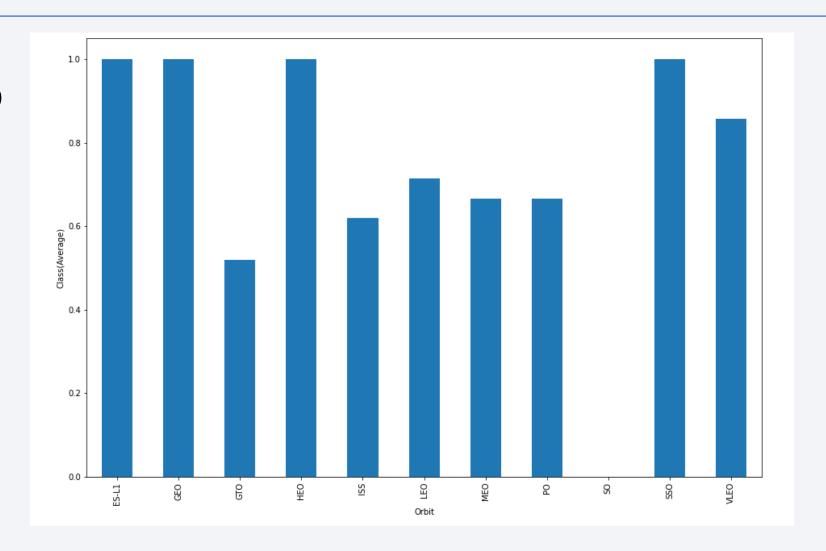
Payload vs. Launch Site



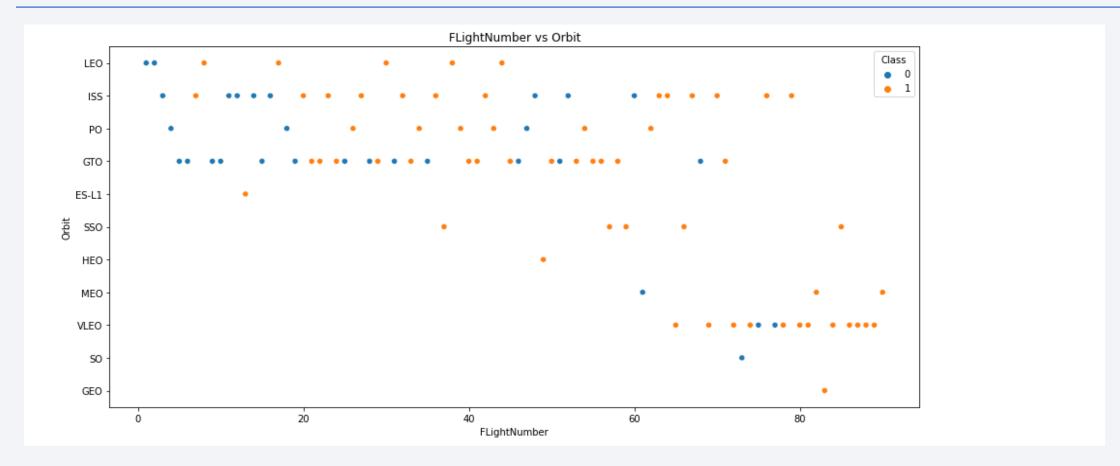
Now if you observe the scatter point chart, you will find for the VAFB-SLC launch site there
are no rockets launched for heavy payload mass(greater than 10000).

Success Rate vs. Orbit Type

The orbit types **SSO**, **HEO**, **GEO** and **ES-L1** had the highest success rate.

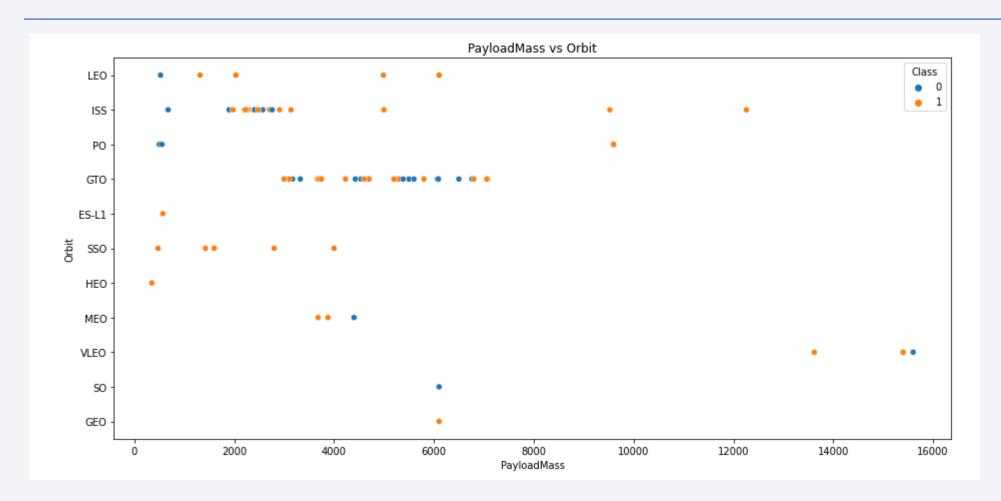


Flight Number vs. Orbit Type



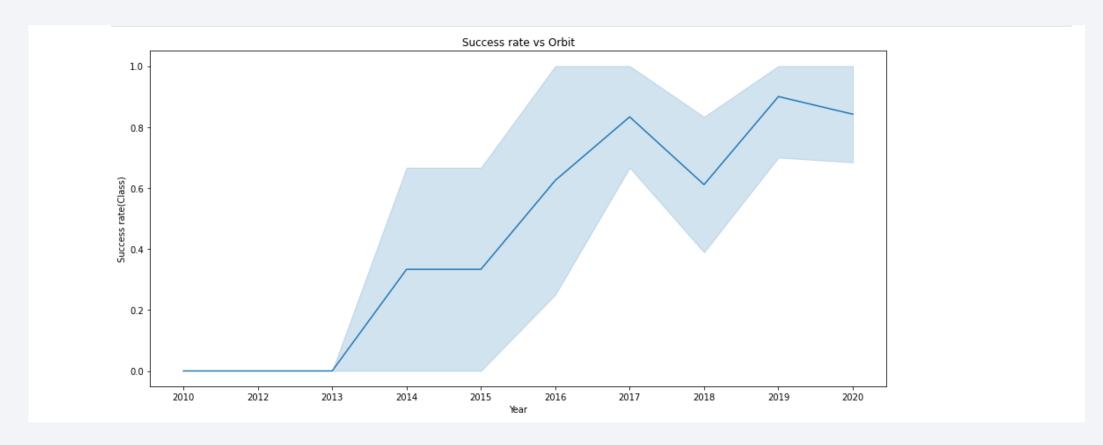
You can see that in the LEO orbit the Success appears related to the number of flights; on the other hand, there seems to be no relationship between flight number when in GTO orbit.

Payload vs. Orbit Type



- With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS.
- However for GTO we cannot distinguish this well as both positive landing rate and negative landing (unsuccessful mission) are both there.

Launch Success Yearly Trend



It is apparent that the success rate has significantly increased from 2013 to 2020.

All Launch Site Names

Given the data, these are the names of the launch sites where different rocket landings where attempted:

- CCAFS LC-40
- CCAFS SLC-40
- KSC LC-39A
- VAFB SLC-4E

Launch Site Names Beginning with 'CCA'

Date	Launch_Site	Orbit	Customer	Mission_Outcome	Landing_Outcome
04-06-2010	CCAFS LC-40	LEO	SpaceX	Success	Failure (parachute)
08-12-2010	CCAFS LC-40	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
22-05-2012	CCAFS LC-40	LEO (ISS)	NASA (COTS)	Success	No attempt
08-10-2012	CCAFS LC-40	LEO (ISS)	NASA (CRS)	Success	No attempt
01-03-2013	CCAFS LC-40	LEO (ISS)	NASA (CRS)	Success	No attempt

These are 5 records where launch sites begin with the letters 'CCA'. As we can see, there are other organizations besides SpaceX that were testing their rockets.

Total Payload Mass

- The information in the table displays the total payload mass carried by boosters launched by NASA.
- It seems that *NASA (CRS)* had a significantly higher total payload mass compared to the rest.

Customer	Total_Payload_Mass
NASA (CRS)	45596
NASA (CCDev)	12530
NASA (CCP)	12500
NASA (CCD)	12055
NASA (CTS)	12050
NASA (CRS), Kacific 1	2617
NASA / NOAA / ESA / EUMETSAT	1192
NASA (LSP) NOAA CNES	553
NASA (COTS)	525
NASA (LSP)	362
NASA (COTS) NRO	0

Average Payload Mass by F9 v1.1

Average_Payload_Mass (kg)	Booster_Version
2928.4	F9 v1.1

• The average payload mass carried by F9 v1.1 was 2928.4 kg.

First Successful Ground Landing Date

Landing_Outcome	Date
Success (ground pad)	22-12-2015

- The first successful ground pad landing took place in December 2015. This was a historic reusable-rocket milestone for both SpaceX and the world.
- Prior to this, no one had ever brought an orbital class booster back intact.

Booster_Version	PAYLOAD_MASSKG_	Landing_Outcome
F9 FT B1022	4696	Success (drone ship)
F9 FT B1026	4600	Success (drone ship)
F9 FT B1021.2	5300	Success (drone ship)
F9 FT B1031.2	5200	Success (drone ship)

- It appears that there only 4 Boosters with a payload mass between 4000 and 6000.
- It is interesting to see that they all had successful landing outcomes.

Total Number of Successful and Failure Mission Outcomes

Mission_Outcome	Outcomes
Failure (in flight)	1
Success	99
Success (payload status unclear)	1

• It appears that missions generally tend to be successful with the exception of one failure.

Boosters That Carried the Maximum Payload Mass

- 12 boosters have carried the maximum payload mass of 15600 kg.
- Since the version names are similar, they might be from the same manufactures.

Booster_Version	PAYLOAD_MASSKG_
F9 B5 B1048.4	15600
F9 B5 B1049.4	15600
F9 B5 B1051.3	15600
F9 B5 B1056.4	15600
F9 B5 B1048.5	15600
F9 B5 B1051.4	15600
F9 B5 B1049.5	15600
F9 B5 B1060.2	15600
F9 B5 B1058.3	15600
F9 B5 B1051.6	15600
F9 B5 B1060.3	15600
F9 B5 B1049.7	15600

2015 Launch Records - Failed Landing Outcomes

Date	Launch_Site	Booster_Version	Landing_Outcome
10-01-2015	CCAFS LC-40	F9 v1.1 B1012	Failure (drone ship)
14-04-2015	CCAFS LC-40	F9 v1.1 B1015	Failure (drone ship)

- It appears that 2 boosters failed to land at the beginning of the year...
- The first successful landing took place later that year in December as we saw earlier.

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

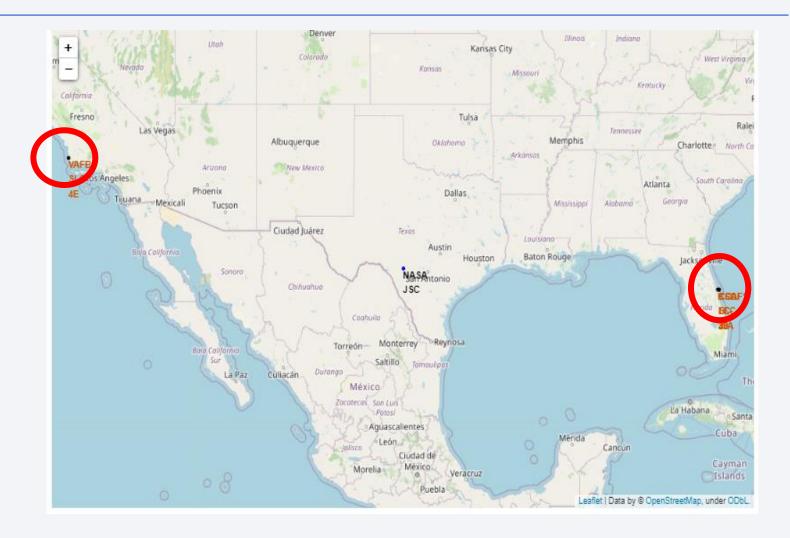
- If we observe the table, it is apparent that the number of successful landings have increased since 2015.
- Before 2013, it seems that there were no attempts to land the boosters.

date	Landing_Outcome	Outcomes
2016-04-08	Success (drone ship)	14
2015-12-22	Success (ground pad)	9
2015-06-28	Precluded (drone ship)	1
2015-01-10	Failure (drone ship)	5
2014-04-18	Controlled (ocean)	5
2013-09-29	Uncontrolled (ocean)	2
2012-05-22	No attempt	22



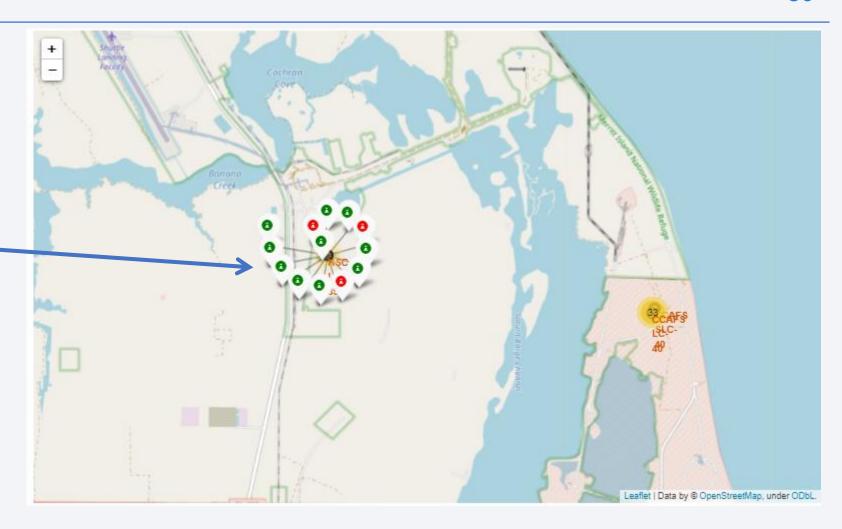
Launch Site Locations

- We can see that all launch sites are in very close proximity to the coast and they are also a couple thousand kilometers away from the equator line.
- It is interesting to see that most launch sites are concentrated near Miami.



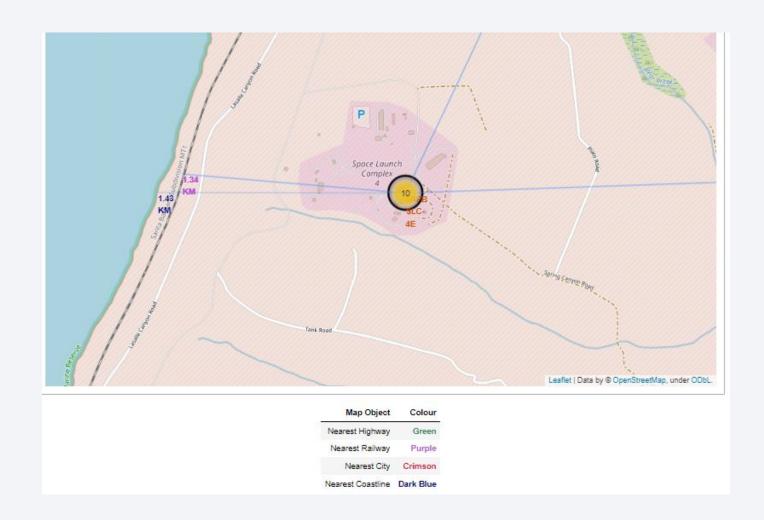
Success Rate of Rocket Launches

- The successful launches are represented by a green marker while the red marker represents failed rocket launches.
- It appears that KSC LC-39A had the highest success rate of rocket launches compared to other launch sites.



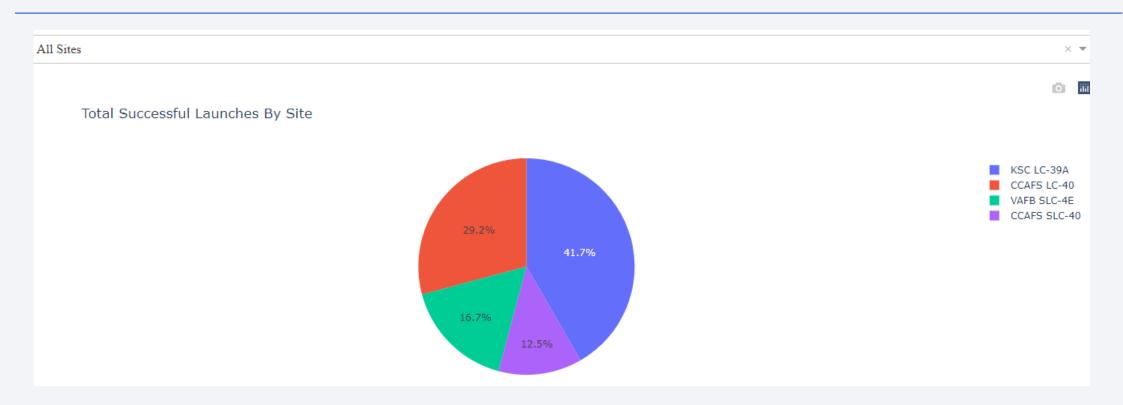
Surrounding Landmarks

- It appears that launch sites are usually set up at least 18 km away from cities. This may be because of the desire to prevent any crashes near populated areas.
- It is also apparent that launch sites are in very close proximity to railways and highways. Perhaps, due to the necessary transportation requirements for rocket parts.
- The sites are close the coastline. This is evident with the many rocket landing tests on water bodies like the ocean.



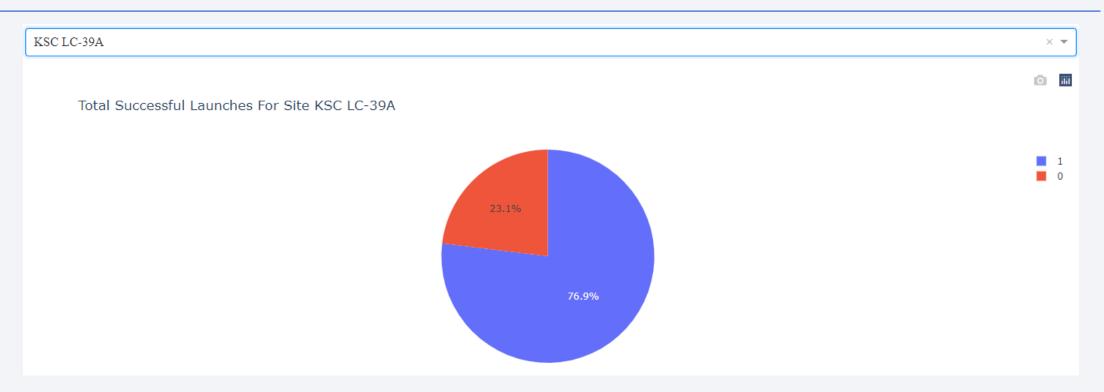


Successful Launches by Site



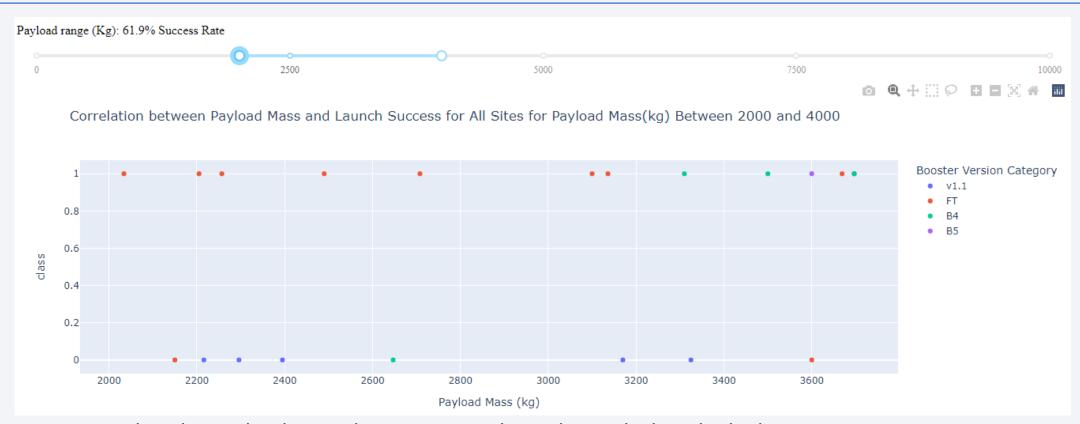
- Site KSC LC-39A has the largest successful launches as well the highest launch success rate.
- More investigation may be needed to determine why KSC LC-39A is the preferred launch site.

Total Successful Launches for Site KSC LC-39A



- As we can see, 76.9% of the total launches at site KSC LC-39A were successful. This is a the highest success rate of all the different launch sites.
- However, this success rate was only around 3% higher than the runner up; site CCAFS LC-40.

Payload Mass vs. Launch Success for All Sites

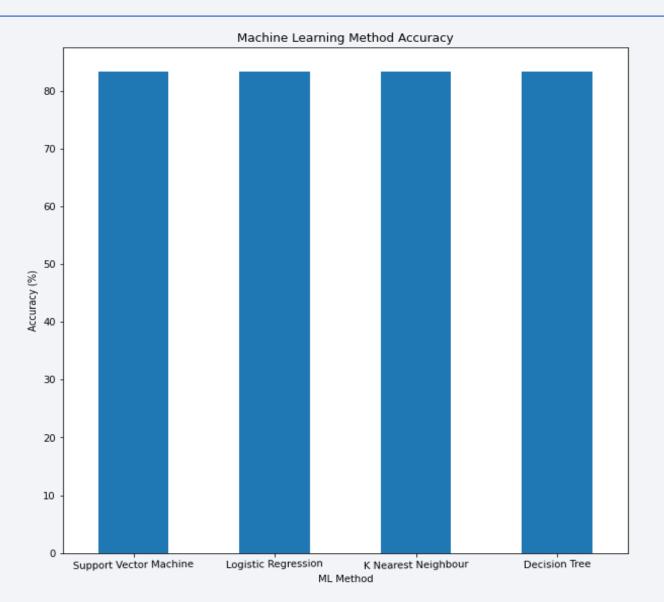


- It appears that the payload range between 2000 kg and 4000 kg has the highest success rate.
- The launch success rate was also dramatically low between the payload range of 0kg and 2500kg. Perhaps very low masses decrease launch success.
- The booster version **FT**, seems to have a higher success rate than other booster versions



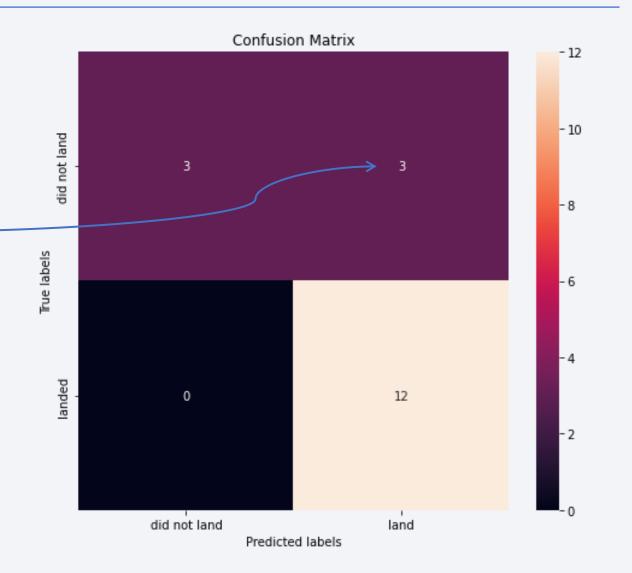
Classification Accuracy

 Since all the methods have an identical accuracy score of 83.33%, we decided to use Logistic Regression for the classification



Confusion Matrix

- The chart shows the confusion matrix of the Logistic Regression model that was chosen.
- The model only failed to accurately predict 3 labels.



Conclusions

In order to compete with SpaceX, it was crucial to analyze their data. Through this process, a general picture of their success methods was produced.

- All their launch sites are located near the coast, away from nearby cities. This enabled to them to test their rocket landings without much interference.
- Site KSC LC-39A had the highest launch success rate out of all the launch sites.
- From 2015 onwards, the success rate of rocket landings significantly increased. It was also apparent that landing success increased with flight number

All this data was used to train a machine learning model that is able to predict the landing outcome of rocket launches with 83.33% accuracy.

This will allow our company to make more attractive offers than SpaceX and increase the confidence of our investors and customers. Can anyone say "No" to a company that can predict the success of their product?

