

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
- Data wrangling
- Exploratory Data Analysis with Data Visualization
- Exploratory Data Analysis 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

• SpaceX is the most successful company of the commercial space age, making space travel affordable. The company 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 SpaceX can reuse the first stage. Therefore, if we can determine if the first stage will land, we can determine the cost of a launch. Based on public information and machine learning models, we are going to predict if SpaceX will reuse the first stage.

Problems we want to find answers

- How do variables such as payload mass, launch site, number of flights, and orbits affect the success of the first stage landing?
- Does the rate of successful landings increase over the years?
- What is the best algorithm that can be used for binary classification in this case?



Methodology

Data collection methodology:

- Using SpaceX Rest API
- Using Web Scrapping from Wikipedia

Performed data wrangling

- Filtering the data
- Dealing with missing values
- Using One Hot Encoding to prepare the data to a binary classification

Performed exploratory data analysis (EDA) using visualization and SQL

Performed interactive visual analytics using Folium and Plotly Dash

Performed predictive analysis using classification models

- Building, tuning and evaluation of classification models to ensure the best results

Data Collection

Data collection

Data collection process involved a combination of API requests from SpaceX REST API and Web Scraping data from a table in SpaceX's Wikipedia entry. We had to use both of these data collection methods in order to get complete information about the launches for a more detailed analysis.

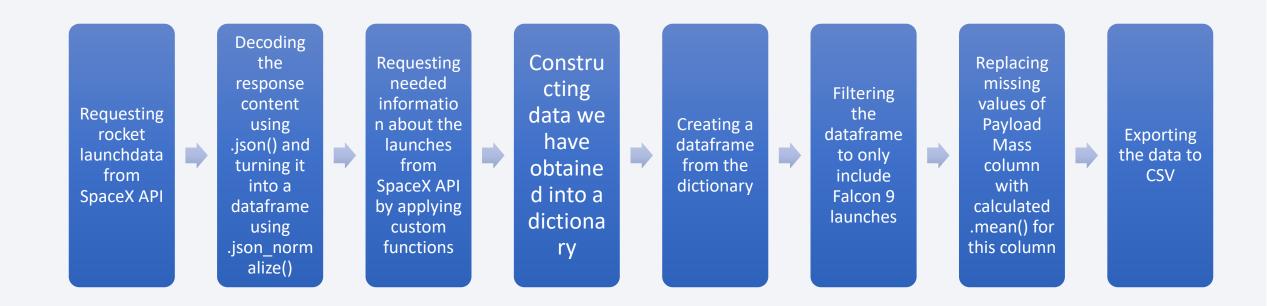
Data Columns obtained by using SpaceX REST API:

FlightNumber, Date, BoosterVersion, PayloadMass, Orbit, LaunchSite, Outcome, Flights, GridFins, Reused, Legs, LandingPad, Block, ReusedCount, Serial, Longitude, Latitude

Data Columns obtained by using Wikipedia Web Scraping:

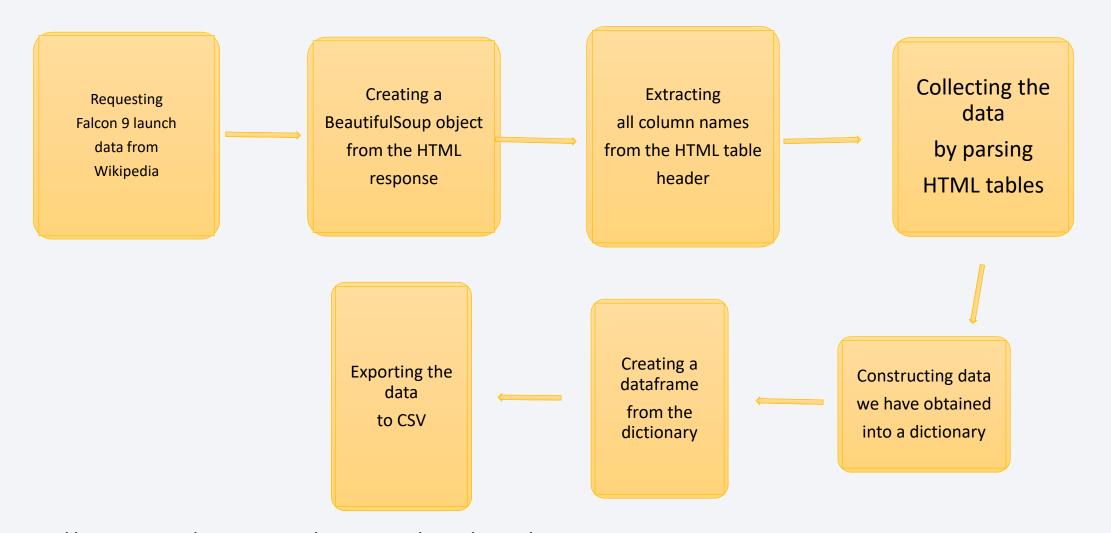
Flight No., Launch site, Payload, PayloadMass, Orbit, Customer, Launch outcome, Version Booster, Booster landing, Date, Time

Data Collection – SpaceX API



https://github.com/karthik-bha/IBM_data/blob/main/Data%20Science%20Capstone%20us ing%20API%20Data%20collection%20%5BPrediction%20of%20 Falcon%209%20launches%5D.ipynb

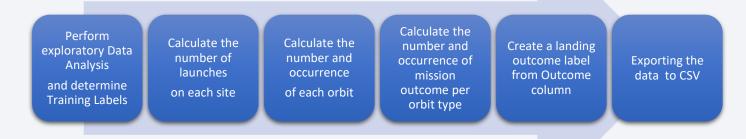
Data Collection - Scraping



Data Wrangling

In the data set, there are several different cases where the booster did not land successfully. Sometimes a landing was attempted but failed due to an accident; for example, True Ocean means the mission outcome was successfully landed to a specific region of the ocean while False Ocean means the mission outcome was unsuccessfully landed to a ground pad False RTLS means the mission outcome was unsuccessfully landed to a ground pad. True ASDS means the mission outcome was successfully landed on a drone ship False ASDS means the mission outcome was unsuccessfully landed on a drone ship False ASDS means the mission outcome was unsuccessfully landed on a drone ship.

We mainly convert those outcomes into Training Labels with "1" means the booster successfully landed, "0" means it was unsuccessful.



https://github.com/karthik-

bha/IBM_data/blob/main/Data%20Science%20Capstone%20using%20API%20Data%20collection%20%5BPrediction%20of%20Falcon%209%20launches%5D.ipynb

EDA with Data Visualization

Charts that were plotted:

Flight Number vs. Payload Mass, Flight Number vs. Launch Site, Payload Mass vs. Launch Site, Orbit Type vs. Success Rate, Flight Number vs. Orbit Type, Payload Mass vs Orbit Type and Success Rate Yearly Trend

Scatter plots show the relationship between variables. If a relationship exists, they could be used in machine learning model.

Bar charts show comparisons among discrete categories. The goal is to show the relationship between the specific categories being compared and a measured value.

Line charts show trends in data over time (time series).

https://github.com/karthik-bha/IBM_data/blob/main/Data_visualization_of_SpaceX_data.ipynb

EDA with SQL

Performed SQL queries:

- Displaying the names of the unique launch sites in the space mission
- Displaying 5 records where launch sites begin with the string 'CCA'
- Displaying the total payload mass carried by boosters launched by NASA (CRS)
- Displaying average payload mass carried by booster version F9 v1.1
- Listing the date when the first successful landing outcome in ground pad was achieved
- Listing the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- Listing the total number of successful and failure mission outcomes
- Listing the names of the booster versions which have carried the maximum payload mass
- Listing the failed landing outcomes in drone ship, their booster versions and launch site names for the months in year 2015
- Ranking 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

https://github.com/karthik-bha/IBM data/blob/main/SQL Data science Capstone.ipynb

Build an Interactive Map with Folium

Build an interactive map with Folium

Markers of all Launch Sites:

- Added Marker with Circle, Popup Label and Text Label of NASA Johnson Space Center using its latitude and longitude coordinates as a start location.
- Added Markers with Circle, Popup Label and Text Label of all Launch Sites using their latitude and longitude coordinates to show their geographical locations and proximity to Equator and coasts.

Coloured Markers of the launch outcomes for each Launch Site:

- Added coloured Markers of success (Green) and failed (Red) launches using Marker Cluster to identify which launch sites have relatively high success rates.

Distances between a Launch Site to its proximities:

- Added coloured Lines to show distances between the Launch Site KSC LC-39A (as an example) and its proximities like Railway, Highway, Coastline and Closest City.

https://github.com/karthik-bha/IBM_data/blob/main/DV_with_Folium_and_Dashboard_with_Plotly_and_Dash.ipynb

Build a Dashboard with Plotly Dash

Launch Sites Dropdown List:

- Added a dropdown list to enable Launch Site selection.

Pie Chart showing Success Launches (All Sites/Certain Site):

- Added a pie chart to show the total successful launches count for all sites and the Success vs. Failed counts for the site, if a specific Launch Site was selected.

Slider of Payload Mass Range:

- Added a slider to select Payload range.

Scatter Chart of Payload Mass vs. Success Rate for the different Booster Versions:

- Added a scatter chart to show the correlation between Payload and Launch Success.

```
https://github.com/karthik-bha/IBM_data/blob/main/DV_with_Folium_and_Dashboard_with_Plotly_and_Dash.ipy nb
```

Predictive Analysis (Classification)

Creating a NumPy array from the column "Class" in data Standardizing
the
data with
StandardScaler,
then
fitting and
transforming it

Splitting the data into training and testing sets with train_test_split Function

Splitting the data into training and testing sets with train_test_split function

Creating a
GridSearchCV
object with cv =
10 to find the
best
parameters

Applying GridSearchCV on LogReg, SVM, Decision Tree, and KNN models Calculating the accuracy on the test data using the method .score() for all models

Examining the confusion matrix for all models

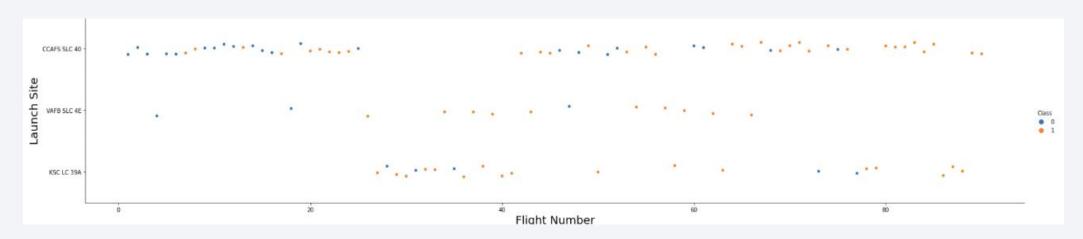
Finding the method performs best by examining the Jaccard_score and F1_score metrics

Results

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

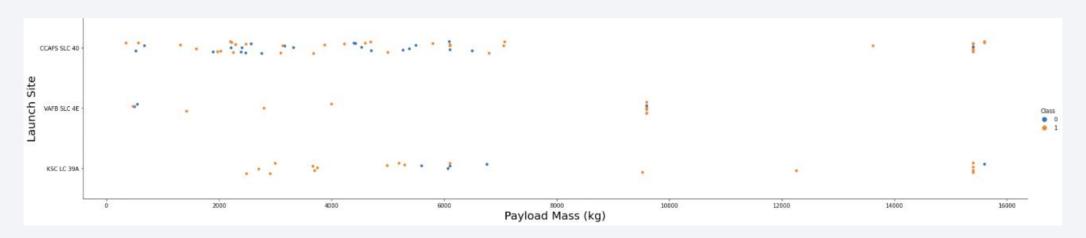


Flight Number vs. Launch Site



- The earliest flights all failed while the latest flights all succeeded.
- The CCAFS SLC 40 launch site has about a half of all launches.
- VAFB SLC 4E and KSC LC 39A have higher success rates.
- It can be assumed that each new launch has a higher rate of success

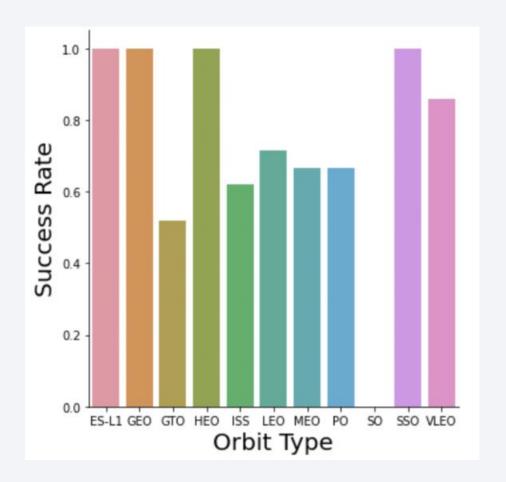
Payload vs. Launch Site



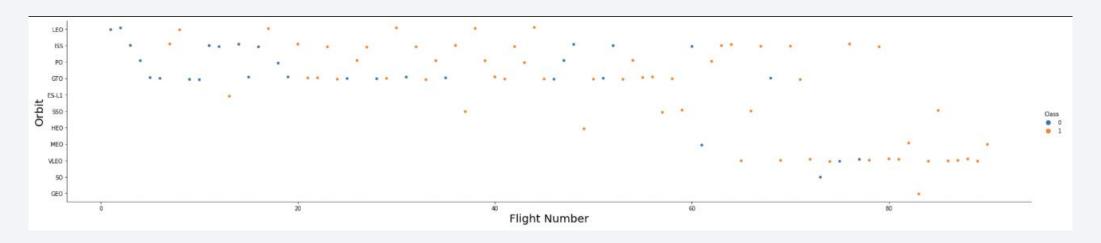
- For every launch site the higher the payload mass, the higher the success rate.
- Most of the launches with payload mass over 7000 kg were successful.
- KSC LC 39A has a 100% success rate for payload mass under 5500 kg too.

Success Rate vs. Orbit Type

- Orbits with 100% success rate:
 - ES-L1, GEO, HEO, SSO
- Orbits with 0% success rate:
 - **-** SO
- Orbits with success rate between 50% and 85%:
 - GTO, ISS, LEO, MEO, PO



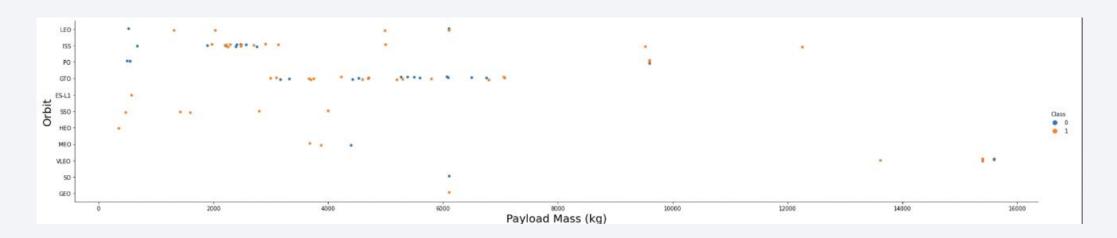
Flight Number vs. Orbit Type



Explanation:

•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



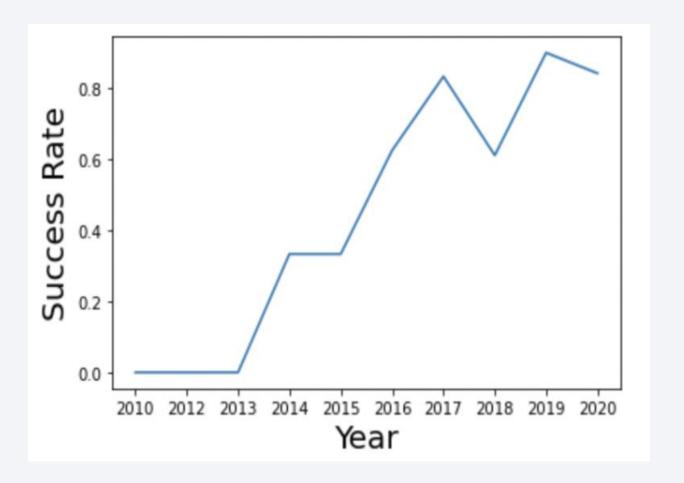
Explanation:

• Heavy payloads have a negative influence on GTO orbits and positive on GTO and Polar LEO (ISS) orbits.

Launch Success Yearly Trend

Explanation:

• The success rate since 2013 kept increasing till 2020.



All Launch Site Names

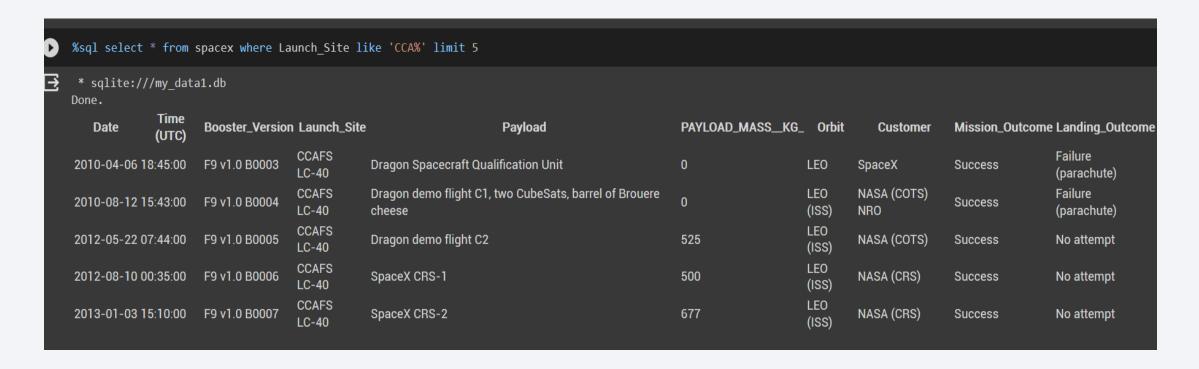
```
% % sql select distinct Launch_Site from spacex

* sqlite://my_data1.db
Done.
    Launch_Site
    CCAFS LC-40
    VAFB SLC-4E
    KSC LC-39A
    CCAFS SLC-40
```

Explanation:

• Displaying the names of the unique launch sites in the space mission

Launch Site Names Begin with 'CCA'



Explanation:

•Displaying 5 records where launch sites begin with the string 'CCA'.

Total Payload Mass

```
[ ] %sql select sum(PAYLOAD_MASS__KG_) as total_payload_mass from spacex where Customer='NASA (CRS)'
    * sqlite:///my_data1.db
    Done.
    total_payload_mass
    45596
```

Explanation:

•Displaying the total payload mass carried by boosters launched by NASA (CRS).

Average Payload Mass by F9 v1.1

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

* sqlite://my_data1.db
Done.
average_payload_mass
2928.4
```

Explanation:

•Displaying average payload mass carried by booster version F9 v1.1.

First Successful Ground Landing Date

Explanation:

•Listing the date when the first successful landing outcome in ground pad was achieved.

Successful Drone Ship Landing with Payload between 4000 and 6000

```
# sqlite://my_data1.db
Done.
Booster_Version
F9 FT B1022
F9 FT B1021.2
F9 FT B1031.2
# sqlite://sqlite://my_data1.db
```

Explanation:

•Listing the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000.

Total Number of Successful and Failure Mission Outcomes

Explanation:

•Listing the total number of successful and failure mission outcomes.

Boosters Carried Maximum Payload

```
%sql select distinct(Booster Version) from spacex where PAYLOAD MASS KG =(select max(PAYLOAD MASS KG ) from spacex)
 * sqlite:///my data1.db
Done.
Booster_Version
F9 B5 B1048.4
F9 B5 B1049.4
F9 B5 B1051.3
F9 B5 B1056.4
F9 B5 B1048.5
F9 B5 B1051.4
F9 B5 B1049.5
F9 B5 B1060.2
F9 B5 B1058.3
F9 B5 B1051.6
F9 B5 B1060.3
F9 B5 B1049.7
```

Explanation:

•Listing the names of the booster versions which have carried the maximum payload mass.

2015 Launch Records

```
* sqlite://my_data1.db
Done.
month Date Booster_Version Launch_Site Landing_Outcome
5- 2015-10-01 F9 v1.1 B1012 CCAFS LC-40 Failure (drone ship)

* CAFS LC-40 Failure (drone ship)

* CCAFS LC-40 Failure (drone ship)

* CCAFS LC-40 Failure (drone ship)
```

Explanation:

•Listing the failed landing outcomes in drone ship, their booster versions and launch site names for the months in year 2015.

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

```
%sql select Landing_Outcome, count(Landing_Outcome) as count from spacex where Date between '2010-06-04' and '2017-03-20' group by Landing_Outcome order by COUNT(Landing_Outcome) DESC

* sqlite:///my_data1.db
Done.
    Landing_Outcome count
No attempt 10
Success (ground pad) 5
Success (drone ship) 5
Failure (drone ship) 5
Controlled (ocean) 3
Uncontrolled (ocean) 2
Precluded (drone ship) 1
Failure (parachute) 1
```

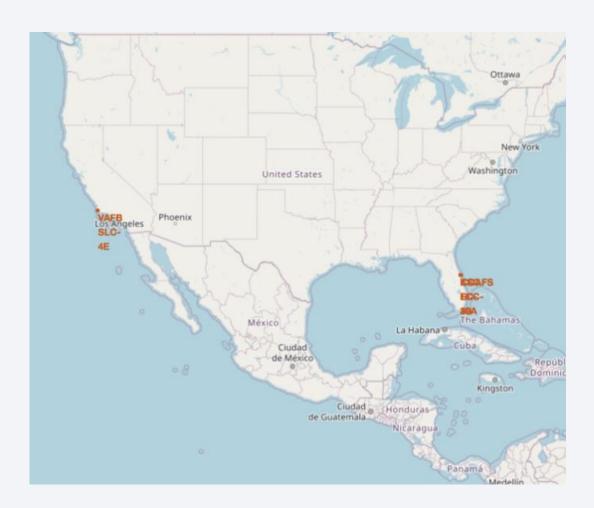
Explanation:

•Ranking 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.



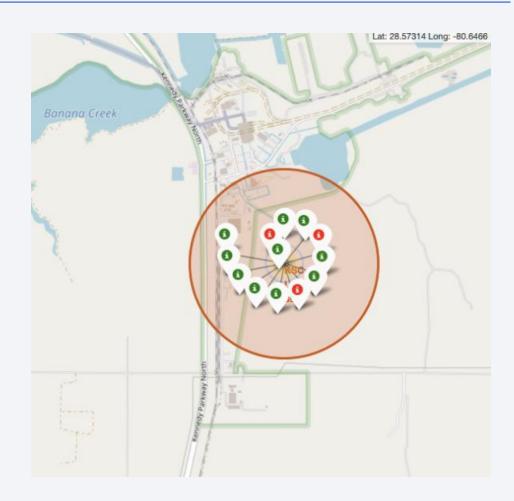
All launch sites' location markers on a global map

- Most of Launch sites are in proximity to the Equator line. The land is moving faster at the equator than any other place on the surface of the Earth. Anything on the surface of the Earth at the equator is already moving at 1670 km/hour. If a ship is launched from the equator it goes up into space, and it is also moving around the Earth at the same speed it was moving before launching. This is because of inertia. This speed will help the spacecraft keep up a good enough speed to stay in orbit.
- All launch sites are in very close proximity to the coast, while launching rockets towards the ocean it minimises the risk of having any debris dropping or exploding near people



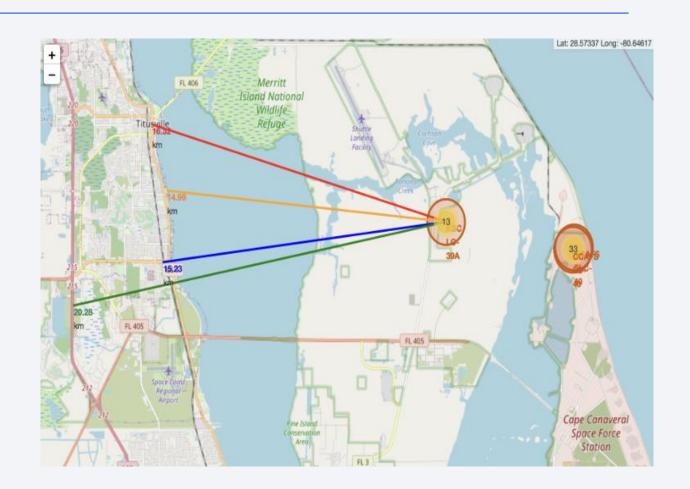
Colour-labeled launch records on the map

- From the colour-labeled markers we should be able to easily identify which launch sites have relatively high success rates.
 - Green Marker = Successful Launch
 - Red Marker = Failed Launch
- Launch Site KSC LC-39A has a very high Success Rate.



Distance from the launch site KSC LC-39A to its proximities

- From the visual analysis of the launch site KSC LC-39A we can clearly see that it is:
 - relative close to railway (15.23 km)
 - relative close to highway (20.28 km)
 - relative close to coastline (14.99 km)
- Also the launch site KSC LC-39A is relative close to its closest city Titusville (16.32 km).
- Failed rocket with its high speed can cover distances like 15-20 km in few seconds. It could be potentially dangerous to populated areas





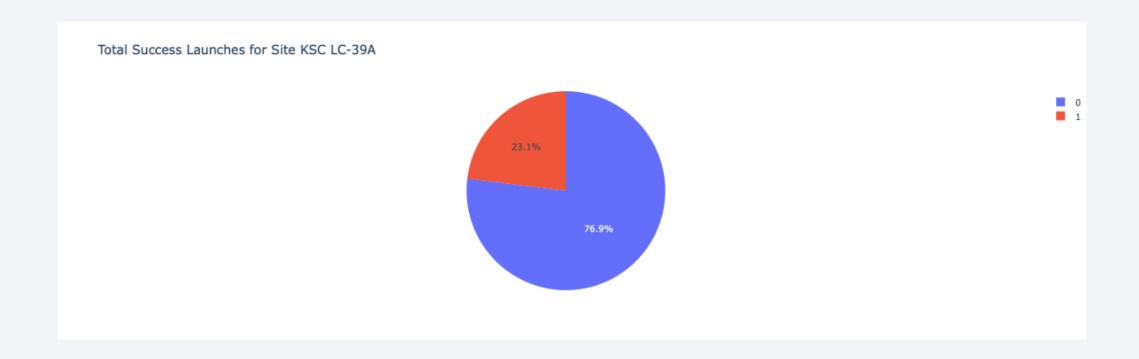
Launch success count for all sites



Explanation:

• The chart clearly shows that from all the sites, KSC LC-39A has the most successful launches

Launch site with highest launch success ratio



Explanation:

• KSC LC-39A has the highest launch success rate (76.9%) with 10 successful and only 3 failed landings.

Payload Mass vs. Launch Outcome for all sites

Explanation:

• The charts show that payloads between 2000 and 5500 kg have the highest success rate.







Classification Accuracy

Explanation:

- Based on the scores of the Test Set, we can not confirm which method performs best.
- Same Test Set scores may be due to the small test sample size (18 samples). Therefore, we tested all methods based on the whole Dataset.
- The scores of the whole Dataset confirm that the best model is the Decision Tree Model. This model has not only higher scores, but also the highest accuracy.

Scores and Accuracy of the Test Set

	LogReg	SVM	Tree	KNN
Jaccard_Score	0.800000	0.800000	0.800000	0.800000
F1_Score	0.888889	0.888889	0.888889	0.888889
Accuracy	0.833333	0.833333	0.722222	0.833333

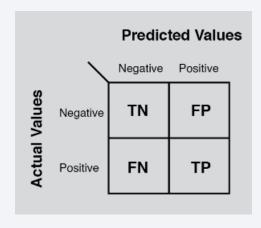
Scores and Accuracy of the Entire Data Set

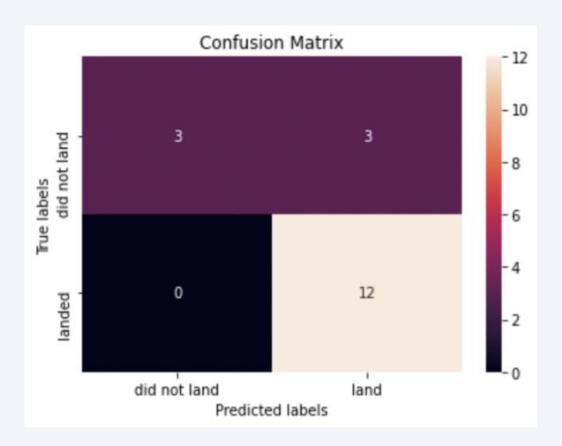
	LogReg	SVM	Tree	KNN
Jaccard_Score	0.833333	0.845070	0.808824	0.819444
F1_Score	0.909091	0.916031	0.894309	0.900763
Accuracy	0.866667	0.877778	0.855556	0.855556

Confusion Matrix

Explanation:

• Examining the confusion matrix, we see that logistic regression can distinguish between the different classes. We see that the major problem is false positives.





Conclusions

- Decision Tree Model is the best algorithm for this dataset.
- Launches with a low payload mass show better results than launches with a larger payload mass.
- Most of launch sites are in proximity to the Equator line and all the sites are in very close proximity to the coast.
- The success rate of launches increases over the years.
- KSC LC-39A has the highest success rate of the launches from all the sites.
- Orbits ES-L1, GEO, HEO and SSO have 100% success rate.

Appendix

Special Thanks to:

Coursera

IBM

