

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

1. 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.

2. Problems you 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

Executive Summary

Data collection methodology:

Via SpaceX Rest API

Web Scrapping from Wikipedia

Perform data wrangling

Filtering the data

Dealing with missing values

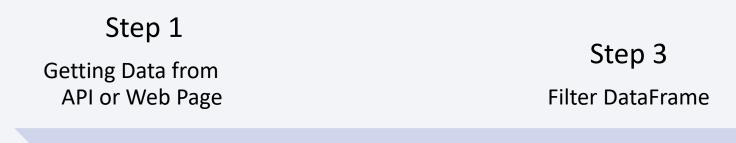
Using One Hot Encoding to prepare the data to a binary classification

- Perform exploratory data analysis (EDA) using <u>visualization</u> and <u>SQL</u>
- Perform interactive visual analytics using <u>Folium</u> and <u>Plotly Dash</u>
- Perform predictive analysis using classification models

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

Data Collection

- •Data collection is a process of grouping data from different sources and media in order to filter and obtain information that can be analyzed.
- Data Columns are 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 are obtained by using Wikipedia Web Scraping: Flight No., Launch site, Payload, PayloadMass, Orbit, Customer, Launch outcome, Version Booster, Booster landing, Date, Time





Step 2

Make a DataFrame

Generate a CSV file

Step 4

Data Collection – SpaceX API





Data Collection - Scraping

Falcon 9 launch data from Wikipedia

```
1 | static url = "https://en.wikipedia.org/w/index.php?title=List of Falcon 9 and Falcon Heavy launches&oldid=1027686922"
2 response = requests.get(static url)
```

Creating a BeautifulSoup object from the HTML

soup=BeautifulSoup(response.text, "html.parser")

Extracting all column names from the HTML header

html tables=soup.find all('table') print(html tables)

Collecting the data HTML tables

6 | for columnas in first_launch_table.find_all("th",scope="col"): column_names.append(columnas.text.removesuffix("\n"))

Constructing data we have obtained into a dictionary

```
for table number, table in enumerate(soup.find_all('table', "wikitable plainrowheaders collapsible")):
    for rows in table.find_all("tr"):
        if rows.th:
           if rows.th.string:
                flight_number=rows.th.string.strip()
                flag=flight_number.isdigit()
           flag=False
        row=rows.find_all('td')
if flag:
           launch_dict['Flight No.'].append(flight_number)
```

Creating a dataframe from the dictionary

df=pd.DataFrame(launch_dict)

Exporting the data to CSV

1 df.to_csv('spacex_web_scraped_modificado.csv', index=False)

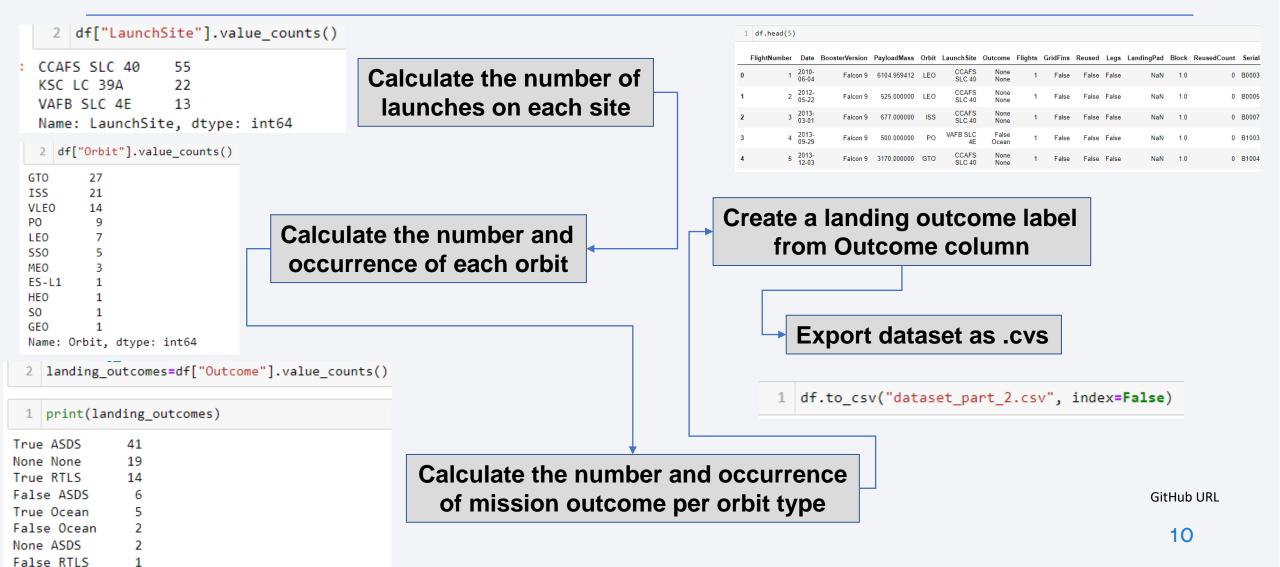
Time	Date	Booster landing	Version Booster	outcome	Customer	Orbit	Payload mass	Payload
18:45	4 June 2010	Failure	F9 v1.0B0003.1	Success	SpaceX	LEO	0	Dragon Spacecraft Qualification Unit
15:43	8 December 2010	Failure	F9 v1.0B0004.1	Success	NASA	LEO	0	Dragon
07:44	22 May 2012	No attempt	F9 v1.0B0005.1	Success	NASA	LEO	525 kg	Dragon
00:35	8 October 2012	No attempt	F9 v1.0B0006.1	Success	NASA	LEO	4,700 kg	SpaceX CRS-1
15:10	1 March 2013	No attempt	F9 v1.0B0007.1	Success	NASA	LEO	4,877 kg	SpaceX CRS-2

GitHub URL

9

Data Wrangling

Name: Outcome, dtype: int64



EDA with Data Visualization

Scatter Plots

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

- 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

Line Plot

Line plots show trends in data over time (time series)

Success Rate Yearly Trend

Bar Plot

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

Payload Mass vs Orbit Type

EDA with SQL

Performed SQL queries:

- Names of the unique launch sites in the space mission
- 5 records where launch sites begin with the string 'CCA'.
- The total payload mass carried by boosters launched by NASA (CRS).
- Average payload mass carried by booster version F9 v1.1.
- Date when the first successful landing outcome in ground pad was acheived.
- 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.
- Names of the booster_versions which have carried the maximum payload mass.
- The records which will display the month names, failure landing_outcomes in drone ship ,booster versions, launch_site for the months in year 2015.
- Rank the count of successful landing_outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

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.



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.

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

Calculating the accuracy on the test data using the method .score() for all models

Applying GridSearchCV on LogReg, SVM, Decision Tree, and KNN models

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

Examining the confusion matrix for all models

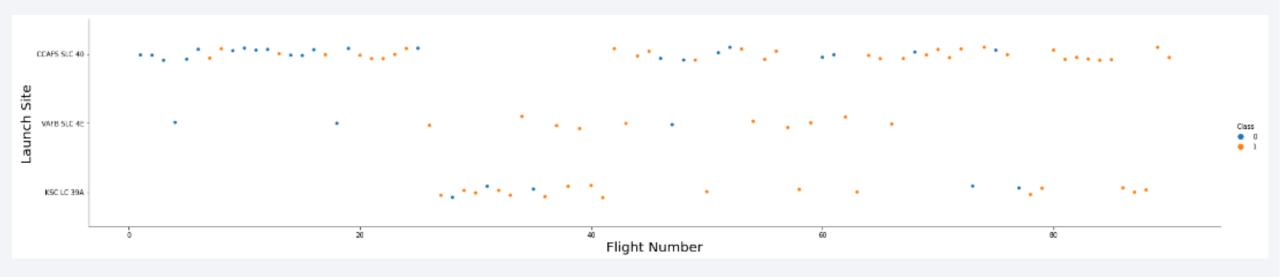
Finding the method performs best by examining the Jaccard score and F1 score metrics

Results

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



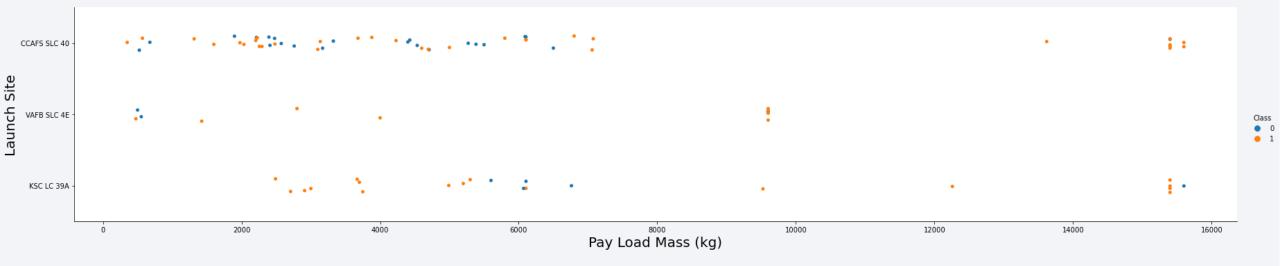
Flight Number vs. Launch Site



Explanation:

- 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



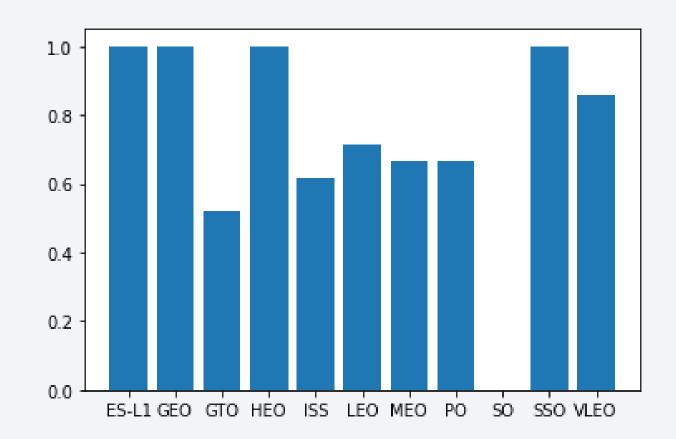
Explanation:

- 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.

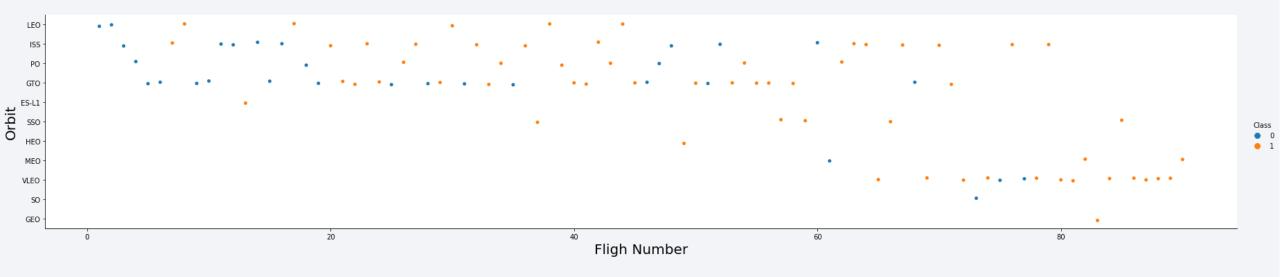
Success Rate vs. Orbit Type

Explanation:

- 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



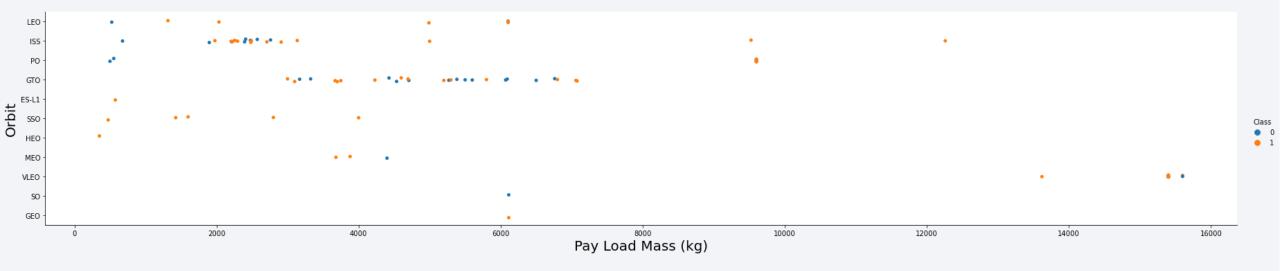
Flight Number vs. Orbit Type



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

Explanation:

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

Launch Site Names Begin with 'CCA'

```
1 # SELECT * FROM SPACEXTBL WHERE Launch_Site LIKE "CCA%" limit 5 # Funciona en el sqlite browser
2 %sql SELECT * FROM SPACEXTBL WHERE "Launch Site" LIKE "CCA%" limit 5;
                Time (UTC)
                            Booster Version
                                             Launch Site
                                                                           Payload
                                                                                                      PAYLOAD_MASS__KG_
        Date
                                                                                                                             Orbit
   1 04-06-2010 18:45:00
                           F9 v1.0 B0003
                                            CCAFS LC-40 Dragon Spacecraft Qualification Unit
                                                                                                                                    Spa
                                                                                                                         0 LEO
                                            CCAFS LC-40 Dragon demo flight C1, two CubeSats, barrel of ...
                                                                                                                         0 LEO (ISS) NAS
   2 08-12-2010 15:43:00
                           F9 v1.0 B0004
                                            CCAFS LC-40 Dragon demo flight C2
                                                                                                                       525 LEO (ISS) NAS
   3 22-05-2012 07:44:00
                           F9 v1.0 B0005
   4 08-10-2012 00:35:00
                                            CCAFS LC-40 SpaceX CRS-1
                                                                                                                       500 LEO (ISS) NAS
                           F9 v1.0 B0006
                                            CCAFS LC-40 SpaceX CRS-2
                                                                                                                       677 LEO (ISS) NAS
   5 01-03-2013 15:10:00
                           F9 v1.0 B0007
```

Explanation:

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

Total Payload Mass

```
# SELECT sum(PAYLOAD_MASS__KG_) from SPACEXTBL where Payload like "%crs%" # Funciona en el sqlite browser
%sql SELECT sum(PAYLOAD_MASS__KG_) from SPACEXTBL where Payload like "%crs%"
...

total_payload_mass
45596
```

Explanation:

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

Average Payload Mass by F9 v1.1

```
# SELECT AVG(PAYLOAD_MASS__KG_) FROM SPACEXTBL WHERE Booster_Version LIKE "%F9 v1.1%" # Funciona en el sqlite browser
%sql SELECT AVG(PAYLOAD_MASS__KG_) FROM SPACEXTBL WHERE Booster_Version LIKE "%F9 v1.1%"

AVG(PAYLOAD_MASS__KG_)

AVG(PAYLOAD_MASS__KG_)

2534.66666666667
```

Explanation:

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

First Successful Ground Landing Date

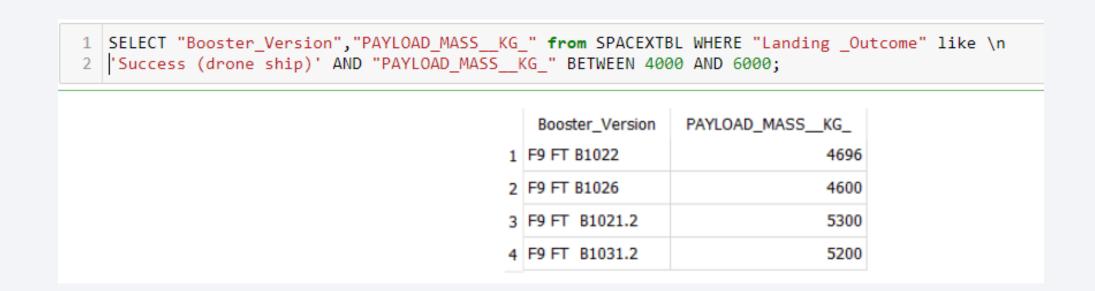
```
# select min(Date) as first_successful_landing from SPACEXTBL WHERE "Landing _Outcome" like 'Success (ground pad)';"
%sql select min(Date) as first_successful_landing from SPACEXTBL WHERE "Landing _Outcome" like ='Success (ground pad)';
...

first_successful_landing
2015-12-22
```

Explanation:

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

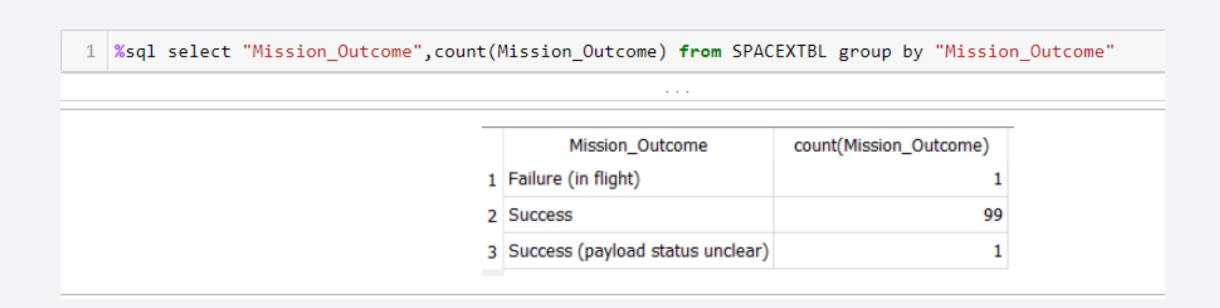
Successful Drone Ship Landing with Payload between 4000 and 6000



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.

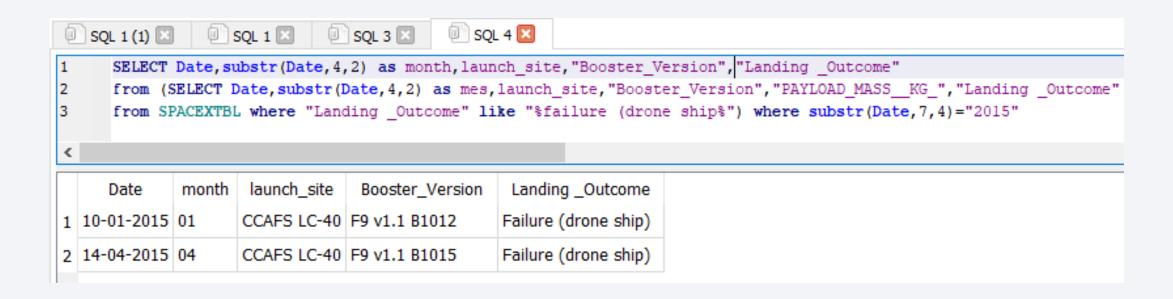
Boosters Carried Maximum Payload

```
%sql select Booster_Version , PAYLOAD_MASS_KG_ from SPACEXTBL where PAYLOAD_MASS_KG_=(select max("PAYLOAD_MASS_KG_") \n _
                                                                                                        from SPACEXTBL)
* sqlite:///my_data1.db
Done.
Booster_Version PAYLOAD_MASS__KG_
  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
```

Explanation:

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

2015 Launch Records



Explanation:

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

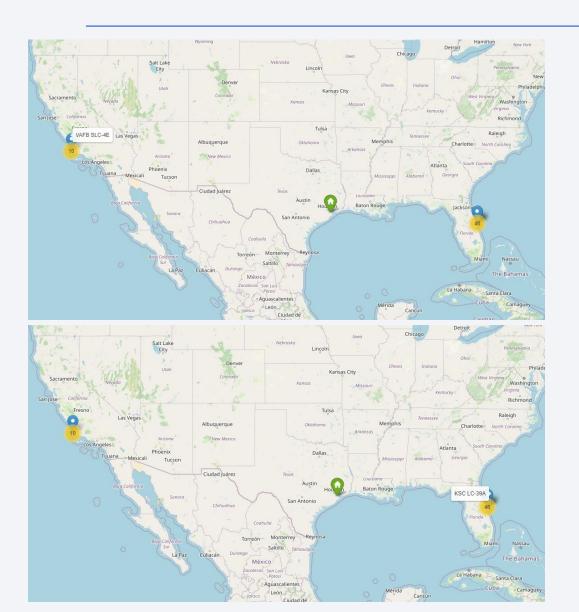
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

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.

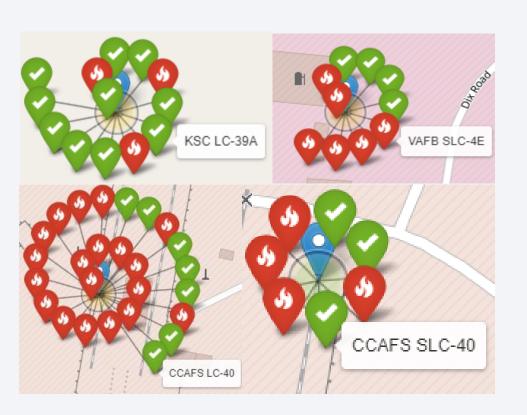


All Launch Sites on Folium Map



The ideal launch sites are near the equator and near coastlines. As can be seen on the map, the three launch sites are as far south as the US geography allows and very close to its coasts.

All launch sites' location markers

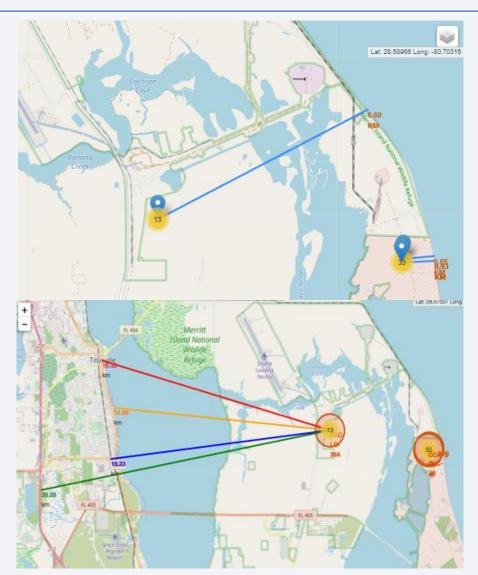


Green Marker shows successful launches and Red Marker shows failures.

From these screenshots its easily understandable that KSC LC-39A has the maximum probability of success

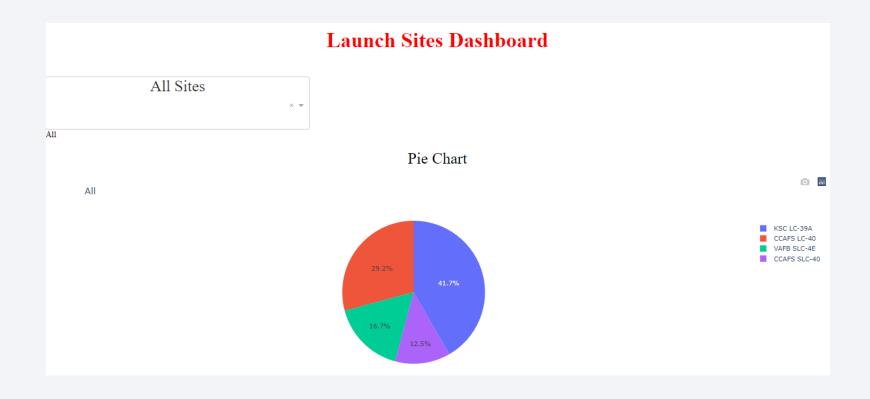
Distance from the launch site KSC LC-39A to its proximities

- Are launch sites in close proximity to railways?Yes, distance < 2 Km
- Are launch sites in close proximity to highways?No, distance < 15 km
- Are launch sites in close proximity to coastline?Yes, distance < 5 Km
- Do launch sites keep certain distance away from cities?
 Yes, distance > 15 km





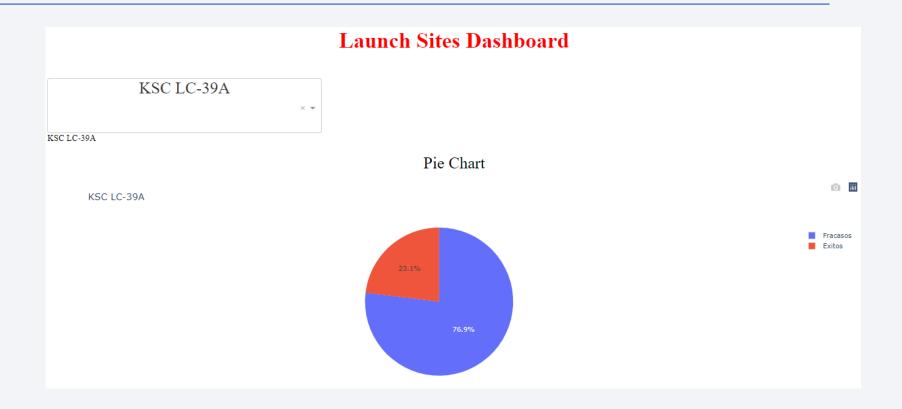
Launch Sites Dashboard



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

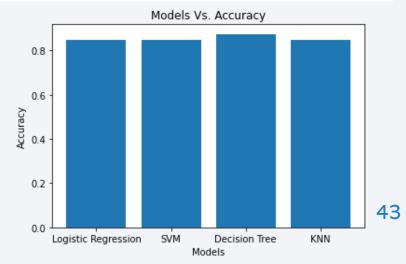
```
In [76]: | Model=['Logistic Regression', 'SVM', 'Decision Tree', 'KNN']

accuracy=[logreg_cv.score(X_test,Y_test),svm_cv.score(X_test,Y_test),knn_cv.score(X_test,Y_accuracy=[logreg_cv.best_score_,svm_cv.best_score_,knn_cv.best_score_]

accuracy=[logreg_cv.score(X_test,Y_test),svm_cv.score(X_test,Y_test),knn_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_test),knn_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_test),knn_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_cv.score(X_test,Y_accuracy=[logreg_c
```

Explanation:

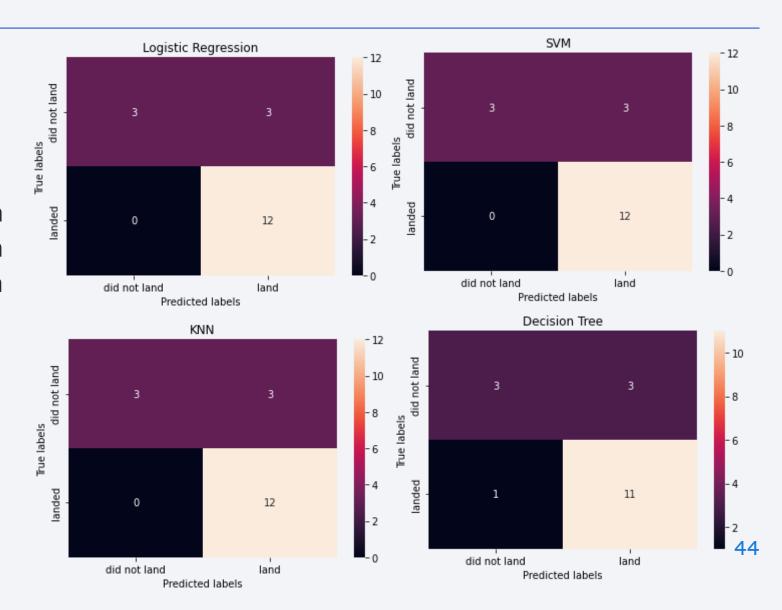
 The scores of the whole Dataset confirm that the best model is the Decision Tree Model. This model has not only higher scores, but also the highest accuracy.



Confusion Matrix

Explanation:

 Examining the confusion matrix, we see that Decision Tree can distinguish between the different classes.



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.
- KSC LC-39A has the highest success rate of the launches from all the sites.
- 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.
- Orbits ES-L1, GEO, HEO and SSO have 100% success rate.

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

