

Winning Space Race with Data Science

William Mansell March 2024



Outline

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

Executive Summary

- Summary of methodologies
 - Data Collection and Wrangling
 - Exploratory Data Analysis & Data Visualization
 - Exploratory Data Analysis using SQL
 - Building an interactive map with Folium
 - Building a Dashboard with Dash Plotly
 - Predictive Analytics & Modeling
- Summary of all results
 - Collection, Wrangling, Exploration results
 - Predictive Analytics results

Introduction

- Project background and context
 - SpaceX is a successful corporation in the commercial space, with a focus on making space travel affordable. This company advertises Falcon 9 rocket launches costing \$62M – the nearest competitor costing upwards of \$165M. This is accomplished through reuse of the first stage of launch.
 - If we can determine if the first stage will land, we can then determine the cost of a launch using a variety of data science methods.
- Problems you want to find answers
 - How do variables such as payload mass, launch site, number of flights, orbits, and others impact the success rate of the first stage landing?
 - What is the best algorithm that can be created and leveraged for binary classification of success?



Methodology

Executive Summary

- Data collection methodology:
 - We leveraged web scraping from Wikipedia combined with the SpaceX API
- Perform data wrangling
 - We filtered the data to relevant data points and eliminated missing values
 - Leveraged various techniques such as One Hot Encoding to prepare the data for binary classification
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Build, tune and evaluate the classification models that led to best results from our analysis

Data Collection

- Data sets were collected using a combination of API requests from SpaceX and web scraping data from the Wikipedia site. This combination allowed for complete information about the launches, providing us with necessary knowledge for a more detailed analysis
- Data Columns utilized from SpaceX API:
 - FlightNumber, Date, BoosterVersion, PayloadMass, Orbit, LaunchSite, Outcome, Flgihts, GridFins,
 Resued, Legs, LandingPad, Block, ReusedCount, Serial, Longitude, Latitude
- Data Columns utilized from Wikipedia Page Web Scraping:
 - Flight No., Launch Site, Payload, PayloadMass, Orbit, Customer, Launch Outcome, Version Booster, Booster Landing, Date, Time

Data Collection – SpaceX API

Replace missing Filtering dataframe Requesting rocket to results only for values where launch data Falcon 9 applicable Decoding data using Creating dataframe Exporting data to .json and from dictionary **CSV** format normalizing Requesting Constructing information about received data into **Data Collection API** launches from dictionary SpaceX API

Data Collection - Scraping

Requesting
Falcon 9 data
from Wikipedia

Created a dataframe from dictionary

Exporting data to CSV format

Leveraged
BeautifulSoup
for HTML

Input data into dictionary

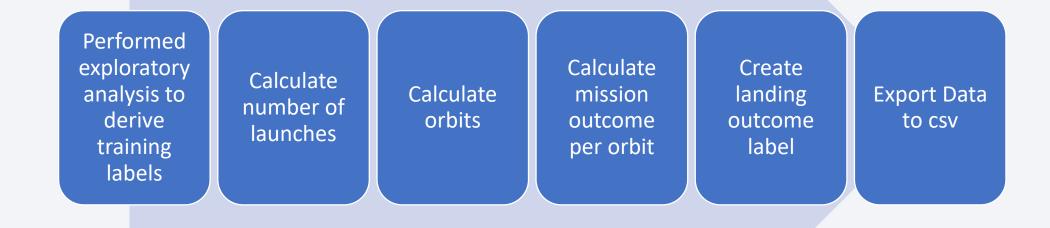
Extracted all columns

Collected and analyzed the data

Data Collection via Web Scraping

Data Wrangling

Data Wrangling



EDA with Data Visualization

- Several visualization charts were plotted to derive insights:
 - 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
 - Success Rate Yearly Trend
- Scatter plots were used to show the relationship between variables
- Bar charts were used to show comparisons amongst categories
- Line charts were used to show trends over time

EDA with Data Visualization

EDA with SQL

- Utilized several SQL queries for data exploration:
 - Display names of each distinct launch site
 - Displayed rows of data where launch sites began with 'CCA'
 - Displayed total payload mass carried by boosters
 - Displayed average payload mass carried by booster ver. F9 v1.1
 - Date when the first successful landing outcome occurred
 - Listing names of boosters which have drone ship success between payload thresholds
 - Etc...

EDA with SQL

Build an Interactive Map with Folium

Folium Map Markers

- Added circle markers, pop-up labels, and text labels using latitude and longitude coordinates
- Added circle markers, pop-up labels, and text labels using latitude and longitude coordinates to show geographical locations
- Launch Outcomes for each Site
 - Colored markers of success [green] and unsuccessful [red] launches using markerclusters to identify success rates for each site
- Distances Between Launch Sites
 - Colored lines to show distances between various launch sites

Build a Dashboard with Plotly Dash

- Pie Chart showing successful launches
- Slider of payload mass range
- Scatter Chart of payload mass vs success rate for the different booster versions

 Unfortunately, this was the only part of the analysis that didn't bear fruit, as the tools were not functioning properly

Predictive Analysis (Classification)

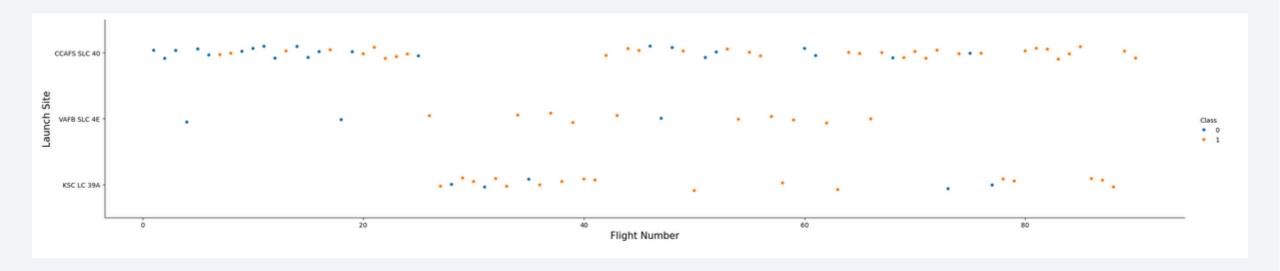
Calculating accuracy Used LogReg, SVM, Created numpy array across various Decision Tree, and K-N for "class" data variables to determine models best model Machine Learning Created a Developed and Standardized the data GridSearchCV to **Prediction** applied confusion with StandardScaler establish the best matrix for each model parameters Found the best model Split the data into a Transformed the data based on Jaccard training set and a and then "fit" it Score and F1 Score testing set functions

Results

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

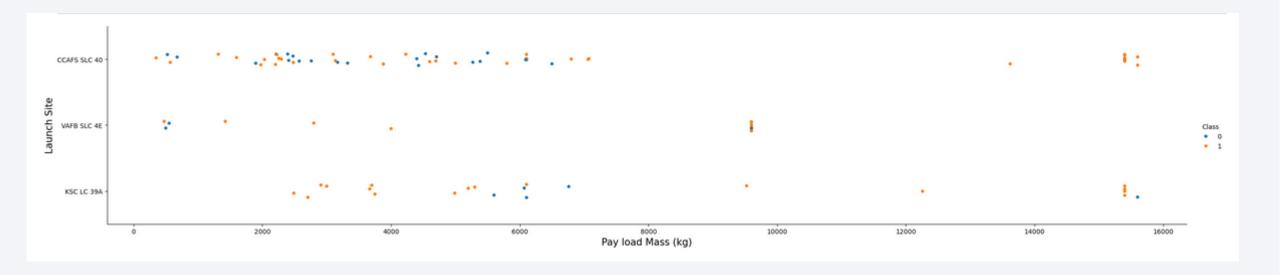


Flight Number vs. Launch Site



- Earliest flights failed; later flights succeeded
- CCAFS SLC 40 launch site has majority of launches
- VAFB SLC 4E has a high success rate. KSC LC 39A is mediocre

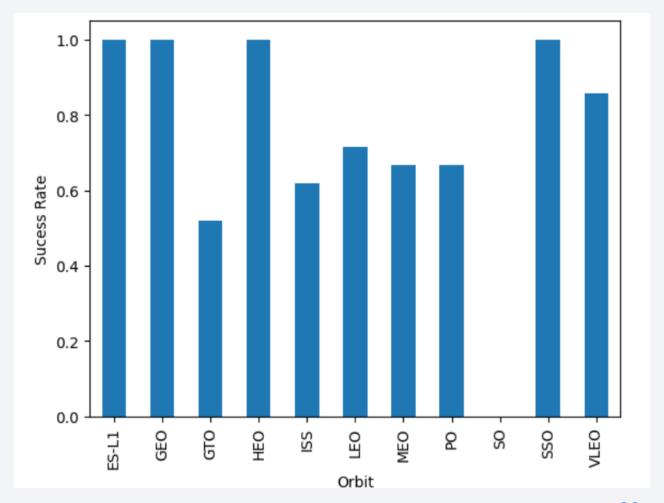
Payload vs. Launch Site



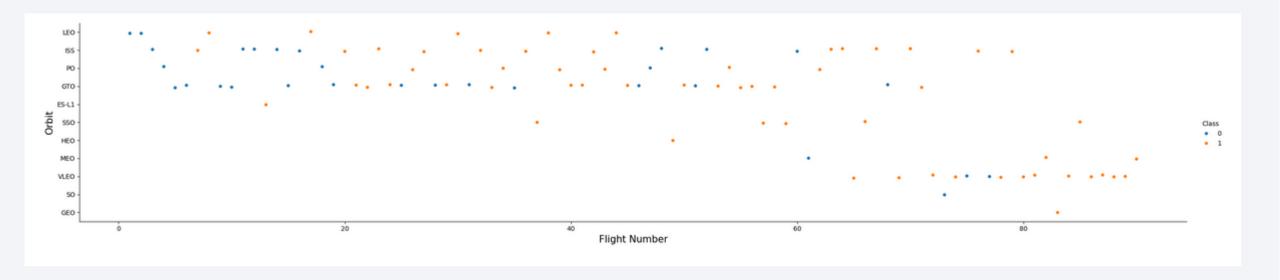
- Launch sites with a larger payload mass have a higher success rate
- Majority of launches with payloads >= 6500 were successful, with few outliers
- KSC LC 39A almost 100% successful after 6500 payload

Success Rate vs. Orbit Type

- ES-L1, GEO, HEO have 100% success rates.
 - SSO is very close
- SO has a 0% success rate recommend removing orbit
- Remainder have greater than
 50%



Flight Number vs. Orbit Type



- LEO Orbit success appears related to the number of flights
- No relationship gleamed from GTO Orbit

Payload vs. Orbit Type



- Heavy payloads impact the successful landing and positive landing rates for POLAR, LEO and ISS
- GTO has both positive and negative landing rates
- I used bigger font for the axes because I felt like it I guess

Launch Success Yearly Trend

- Notated that the success rate since 2013 kept increasing until 2020
- Unable to develop chart to reflect these findings due to error

All Launch Site Names

```
%sql SELECT DISTINCT "LAUNCH_SITE" FROM SPACEXTBL
 * sqlite:///my_data1.db
Done.
  Launch_Site
  CCAFS LC-40
  VAFB SLC-4E
   KSC LC-39A
 CCAFS SLC-40
```

• SQL query that selects distinctly unique launch site names from the database

Launch Site Names Begin with 'CCA'

%sql SELECT * FROM SPACEXTBL WHERE "LAUNCH_SITE" LIKE '%CCA%' LIMIT 5
 * sqlite:///my_data1.db
Done.

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG_	Orbit	Customer	Mission_Outcome	Landing_Outc
2010-06-04	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (paracl
2010-12-08	15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (paracl
2012-05-22	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No atte
2012-10-08	0:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No atte
2013-03-01	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No atte

 SQL query selects 5 rows from the database where the launch sites begin with the character string 'CCA'

Total Payload Mass

```
%sql SELECT SUM("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE "CUSTOMER" = 'NASA (CRS)'

* sqlite://my_data1.db
Done.

SUM("PAYLOAD_MASS__KG_")

45596
```

 SQL query calculates the sum of the payload mass column where the customer is NASA (CRS)

Average Payload Mass by F9 v1.1

 SQL query calculates the average of the payload mass from the database where the booster version is F9 v1.1

```
%sql SELECT AVG("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE "BOOSTER_VERSION" LIKE '%F9 v1.1%'

* sqlite://my_data1.db
Done.

AVG("PAYLOAD_MASS__KG_")

2534.66666666666665
```

First Successful Ground Landing Date

 SQL query derives the first successful landing date by leveraging the "min" function

```
%sql SELECT MIN("DATE") FROM SPACEXTBL WHERE "Landing_Outcome" LIKE '%Success%'

* sqlite://my_data1.db
Done.

MIN("DATE")

2015-12-22
```

Successful Drone Ship Landing with Payload between 4000 and 6000

• SQL query selects the booster version column from the database where the landing outcome column reflects "success (drone ship), but only if the Payload column is between 4000 and 6000

```
%sql SELECT "BOOSTER_VERSION" FROM SPACEXTBL WHERE "Landing_Outcome" = 'Success (drone ship)' AND "PAYLOAD_MASS__KG_" > 4000
* sqlite:///my_data1.db
Done.

Booster_Version
F9 FT B1022
F9 FT B1021.2
F9 FT B1031.2
```

Total Number of Successful and Failure Mission Outcomes

 SQL query selects the count of how many successful mission outcomes there were, along with how many failures

Task 7

List the total number of successful and failure mission outcomes

```
%sql SELECT (SELECT COUNT("MISSION_OUTCOME") FROM SPACEXTBL WHERE "MISSION_OUTCOME" LIKE '%Success%') AS SUCCESS, (SELECT CO
```

* sqlite:///my_data1.db Done.

SUCCESS FAILURE

100

Boosters Carried Maximum Payload

• SQL query selects unique booster version rows from the database where the payload column is the maximum value



2015 Launch Records

```
%sql SELECT substr("DATE", 6, 2) AS MONTH, "BOOSTER_VERSION", "LAUNCH_SITE" FROM SPACEXTBL WHERE "LANDING_OUTCOME" = 'Failure (drone ship)' AND substr("DATE", 0, 5) = '2015'

* sqlite://my_datal.db
Done.

MONTH Booster_Version Launch_Site

01 F9 v1.1 B1012 CCAFS LC-40

04 F9 v1.1 B1015 CCAFS LC-40
```

 SQL query shows the month, booster version and launch sites with failed landings in 2015

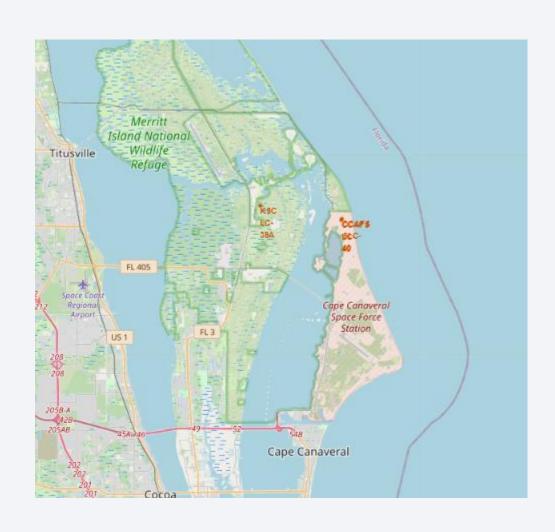
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20



• 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

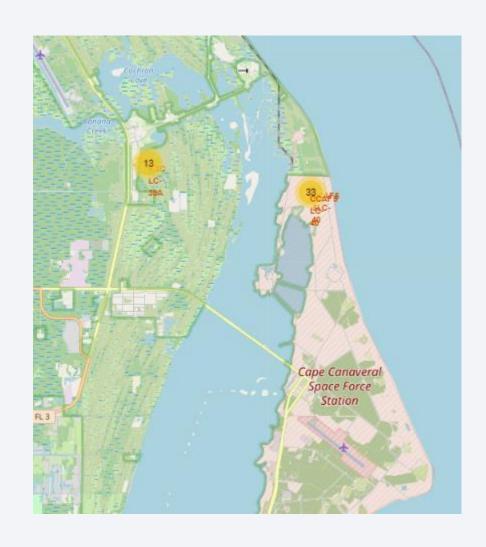


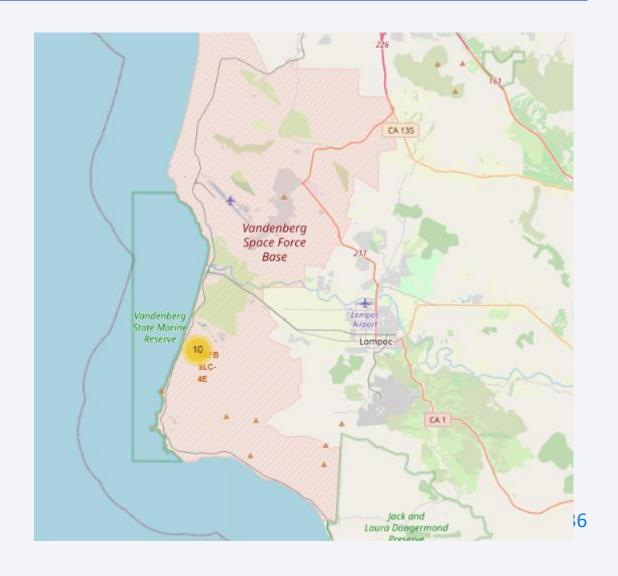
Folium Map 1





Folium Map 2





Folium Map 3





Plotly Code

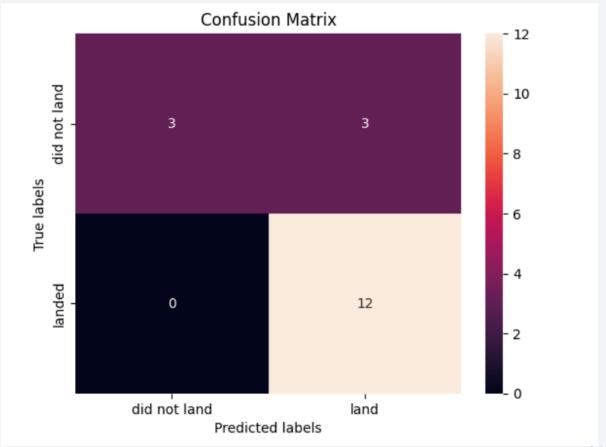
• Provided software did not work – link below to code

Link to code



Confusion Matrix

 SVM model had the most accurate confusion matrix, with an accuracy of 83.33%



Conclusions

• We've leveraged many data science techniques to delve deeper into the relevant data to determine how to approach our competition with a price that cannot be beaten, while ensuring our data backs up our claims pertaining to first stage success of launch.

