

Winning Space Race with Data Science

Simon B. 13.10.2023



Outline

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- Introduction
- Methodology
- Results
- Conclusion
- Appendix

Executive Summary

This presentation summarized methodology and results of a case study into SpaceX.

After collecting data from the SpaceX API and enriching it by Wikipedia data, exploratory analysis is conducted into mission parameters and success factors, and finally a model to predict mission success is created.

Introduction

This project is created as a capstone exercise for the IBM data science course.

It is meant to apply a variety of data science methodology in order to deepen the project owners understanding of them.

Content of this presentation is therefore limited to display this progress.



Methodology

Executive Summary

- Data collection methodology:
 - Describe how data was collected
- Perform data wrangling
 - Describe how data was processed
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - How to build, tune, evaluate classification models

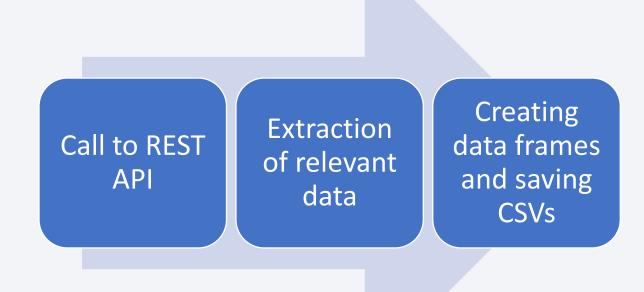
Data Collection

- Base Data was collected from SpaceX accessible online sources
- This was further enriched by web-scraping relevant Wikipedia sources

Data Collection – SpaceX API

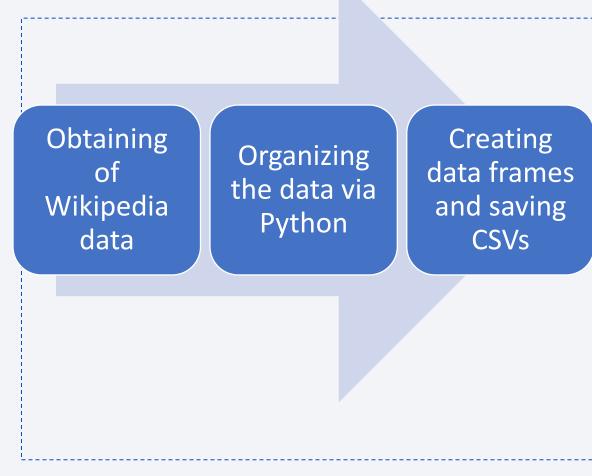
 The REST API was contacted via http calls, specifically its "core" component. Data was then loaded into data frames using the Python library "Pandas"

The Notebook can be found <u>here</u>



Data Collection - Scraping

- Data was scraped from Wikipedia via the Python library "beautifulsoup", relevant data was loaded into data frames and stored as CSV
- The notebook can be found here



Data Wrangling

- Data was cleaned and data types were assigned
- Mission and landing outcomes were calculated based on relevant parameters such as orbit type and launch site
- Notebook can be found here

EDA with Data Visualization

- Outcomes were plotted against potential predictive variables such as flight number and date (to track progression), payload, and orbit type
- Notebook with full set of visuals can be found here

EDA with SQL

- SQL was used to create summarized overviews such as success counts, payload averages, and others. More detailed information on certain booster types was also obtained that way.
- Notebook can be found <u>here</u>

Build an Interactive Map with Folium

- The Python library "Folium" was used to create an interactive map showing lauch sites, missions and their outcomes, as well as distances to certain points of interest
- This enabled a better understanding how launch site locations are chosen and how well they perform as starting points for missions.
- Notebook can be found here

Build a Dashboard with Plotly Dash

- A dashboard with insights into success rates based on launch site and payload was created
- This allows a quick interactive overview
- Code file can be found here

Predictive Analysis (Classification)

- Mission outcome prediction models were created and tested via grid search
- Predictive variables were chosen and prepared
 - Normalizing numerical variables
 - One-hot encoding categorical variables
- Notebook can be found here

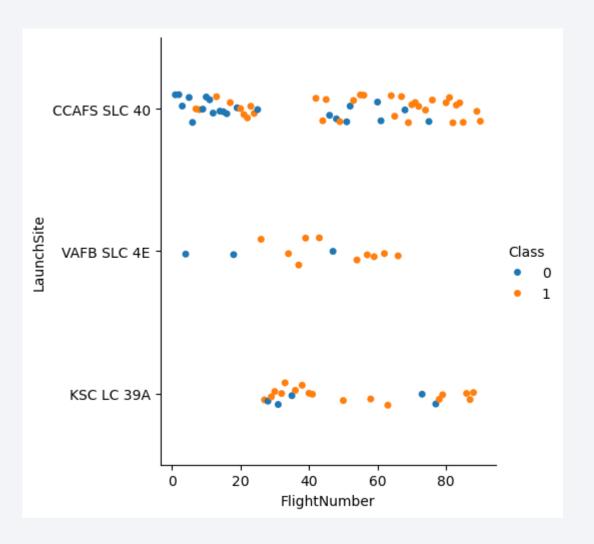
Results

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



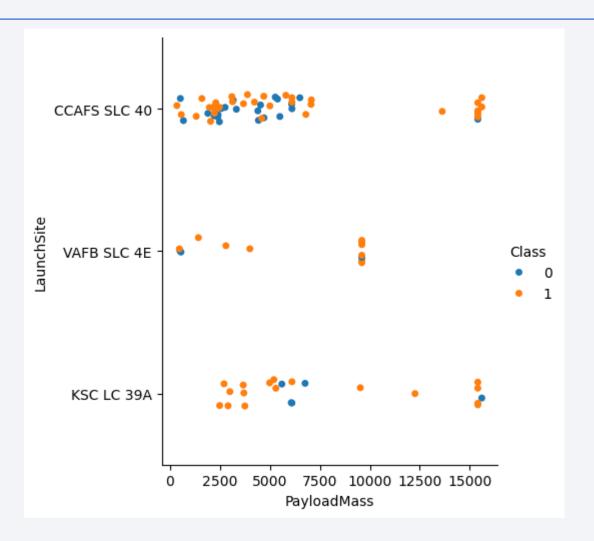
Flight Number vs. Launch Site

- Flights over time can be seen broken down by the three launch sites
- Red dots indicate successful missions, blue dots failures



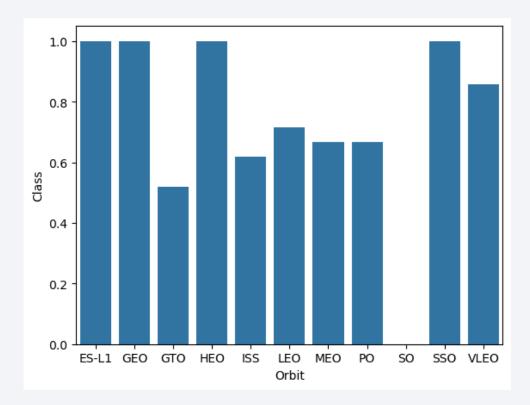
Payload vs. Launch Site

- Payload massed can be seen broken down by the three launch sites
- Red dots indicate successful missions, blue dots failures



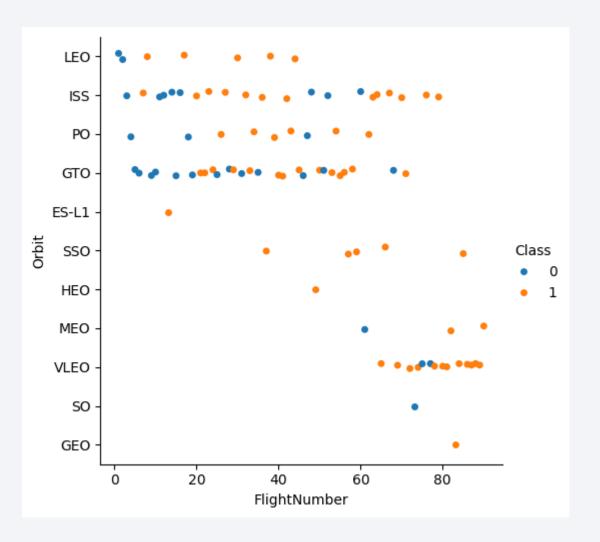
Success Rate vs. Orbit Type

 This bar chart shows the mission success rates for different orbit types



Flight Number vs. Orbit Type

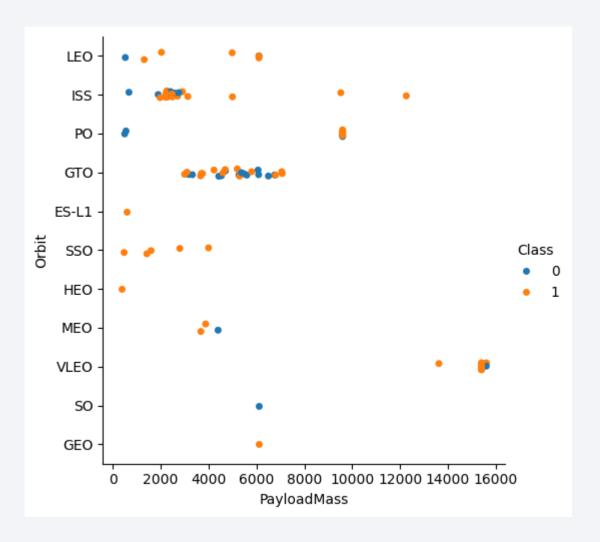
- This plot shows the orbit types by flight number, giving an overview over when missions started going to those orbits
- Red dots indicate successful missions, blue dots failures



Payload vs. Orbit Type

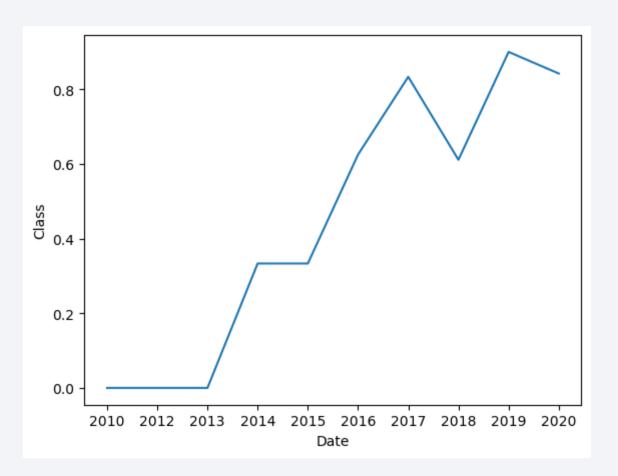
 The plot shows payload masses broken down by orbit types

 Red dots indicate successful missions, blue dots failures



Launch Success Yearly Trend

 The graph shows mission success rate over time



All Launch Site Names

Launch_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

Launch Site Names Begin with 'CCA'

In [14]: %sql select * from SPACEXTABLE where Launch Site like "CCA%" limit 5 * sqlite:///my_data1.db Done. Out[14]: Date **Booster Version Launch Site** Payload PAYLOAD_MASS_KG_ Orbit Customer Mission_Outcome Landing_Outcome (UTC) Dragon 2010-CCAFS LC-Spacecraft F9 v1.0 B0003 Failure (parachute) LEO Success SpaceX 04-06 40 Qualification Unit Dragon demo flight C1, two NASA CCAFS LC-LEO 15:43:00 F9 v1.0 B0004 CubeSats, Failure (parachute) (COTS) Success 08-12 barrel of NRO Brouere cheese Dragon CCAFS LC-LEO NASA 07:44:00 F9 v1.0 B0005 demo flight 525 No attempt Success (ISS) (COTS) CCAFS LC-LEO NASA SpaceX 00:35:00 F9 v1.0 B0006 500 No attempt Success 08-10 CRS-1 (ISS) (CRS) CCAFS LC-SpaceX LEO NASA 15:10:00 F9 v1.0 B0007 No attempt Success 01-03 (ISS) CRS-2 (CRS)

Total Payload Mass

Average Payload Mass by F9 v1.1

First Successful Ground Landing Date

Successful Drone Ship Landing with Payload between 4000 and 6000

Total Number of Successful and Failure Mission Outcomes

Boosters Carried Maximum Payload

```
%sql select Booster_Version from SPACEXTABLE where PAYLOAD_MASS__KG_ = (SELECT MAX(PAYLOAD_MASS__KG_) from SPACEXTABLE)
         * sqlite:///my_data1.db
        Done.
Out[42]: 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
```

2015 Launch Records

```
In [74]:
          %%sql
          select
              substr(Date,0,5) as year_,
              substr(Date, 6,2) as month,
              Booster Version,
              Launch Site,
              count(Mission Outcome) as launches,
              sum(case when Landing Outcome = "Failure (drone ship)" then 1 else 0 end) as landing fails droneship
          from SPACEXTABLE
          where substr(Date,0,5)='2015'
          group by year , month, Launch Site, Booster Version
         * sqlite:///my data1.db
        Done.
Out[74]: year_ month Booster_Version Launch_Site launches landing_fails_droneship
          2015
                         F9 v1.1 B1014 CCAFS LC-40
          2015
                         F9 v1.1 B1015 CCAFS LC-40
          2015
                         F9 v1.1 B1016 CCAFS LC-40
          2015
                        F9 v1.1 B1018 CCAFS LC-40
          2015
                         F9 v1.1 B1012 CCAFS LC-40
          2015
                   11 F9 v1.1 B1013 CCAFS LC-40
          2015
                   12
                           F9 FT B1019 CCAFS LC-40
                                                                                0
```

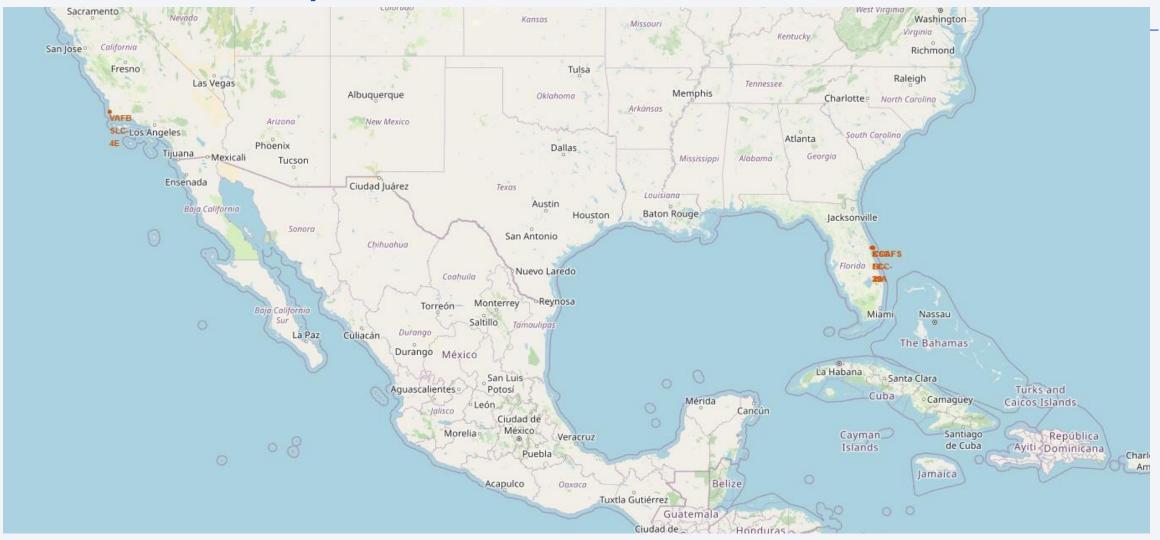
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

 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

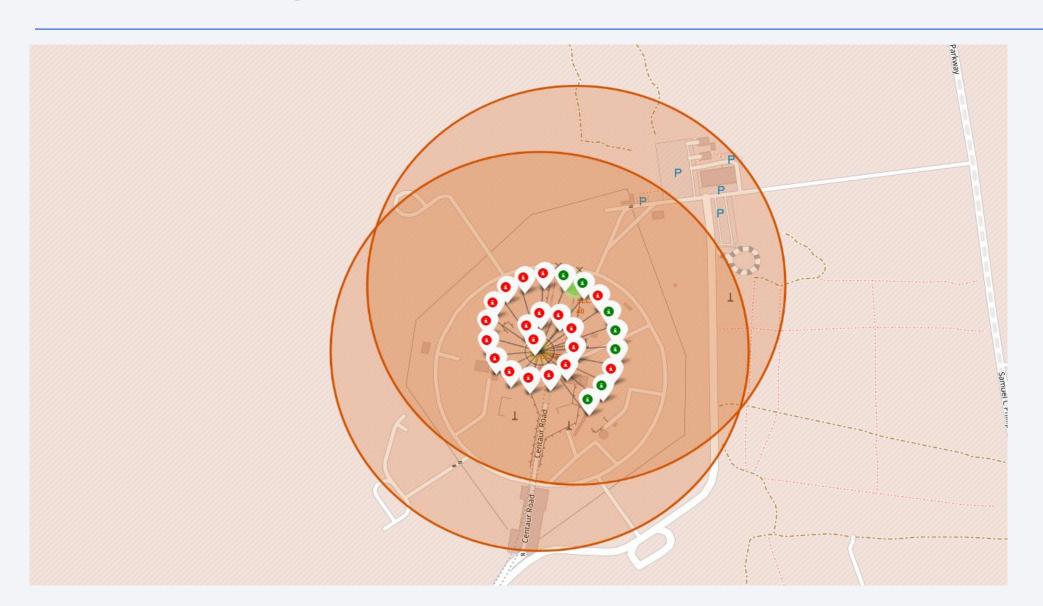
Present your query result with a short explanation here



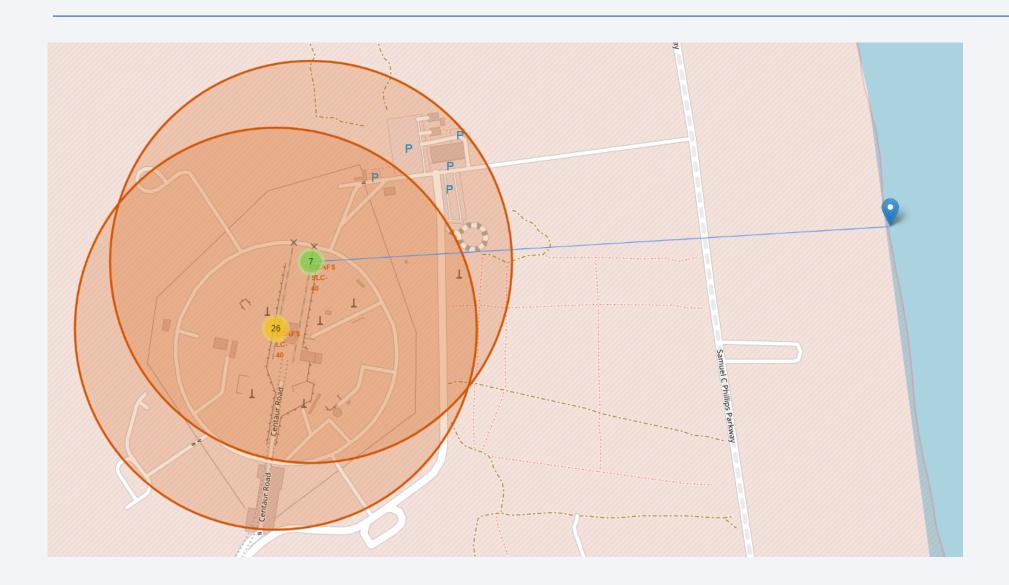
<Folium Map Screenshot 1>



<Folium Map Screenshot 2>

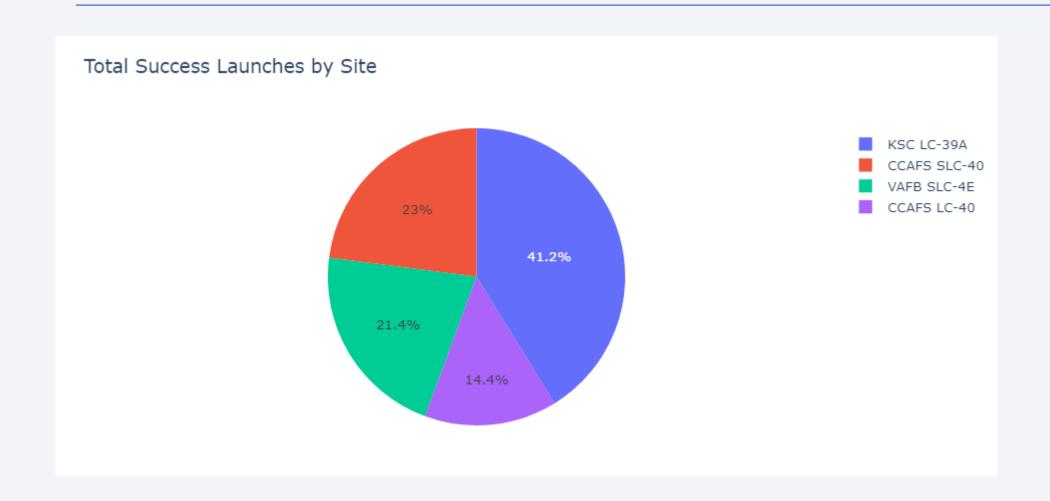


<Folium Map Screenshot 3>

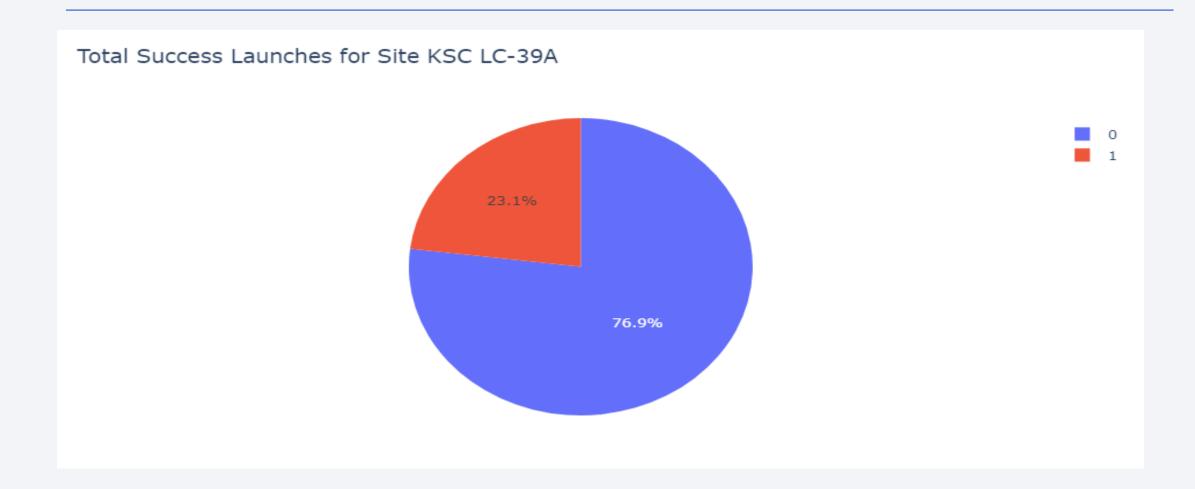




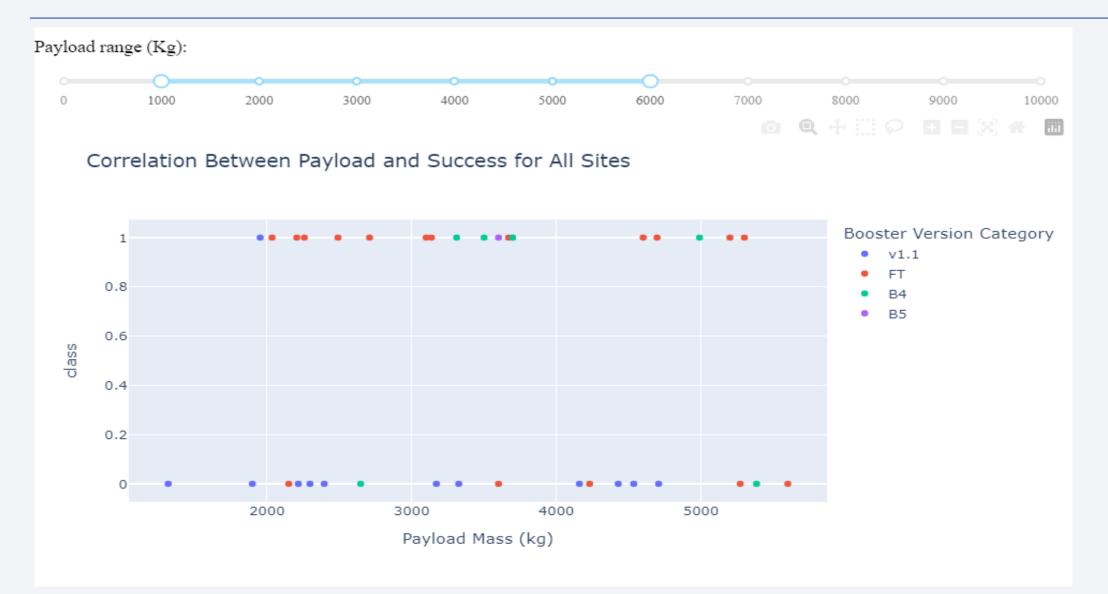
< Dashboard Screenshot 1>



< Dashboard Screenshot 2>

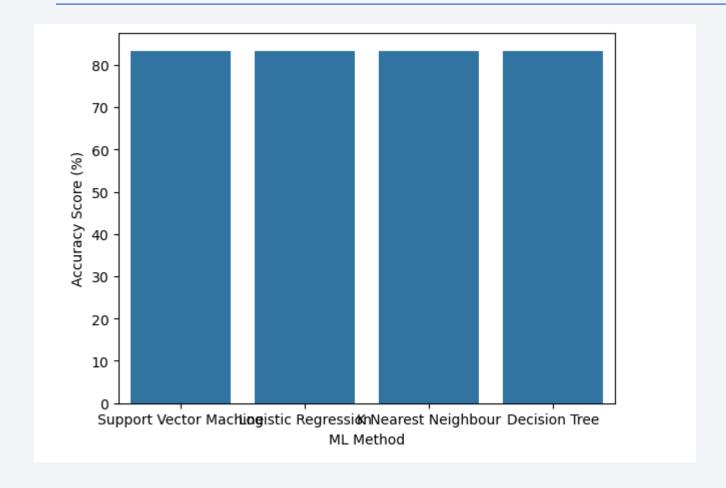


< Dashboard Screenshot 3>

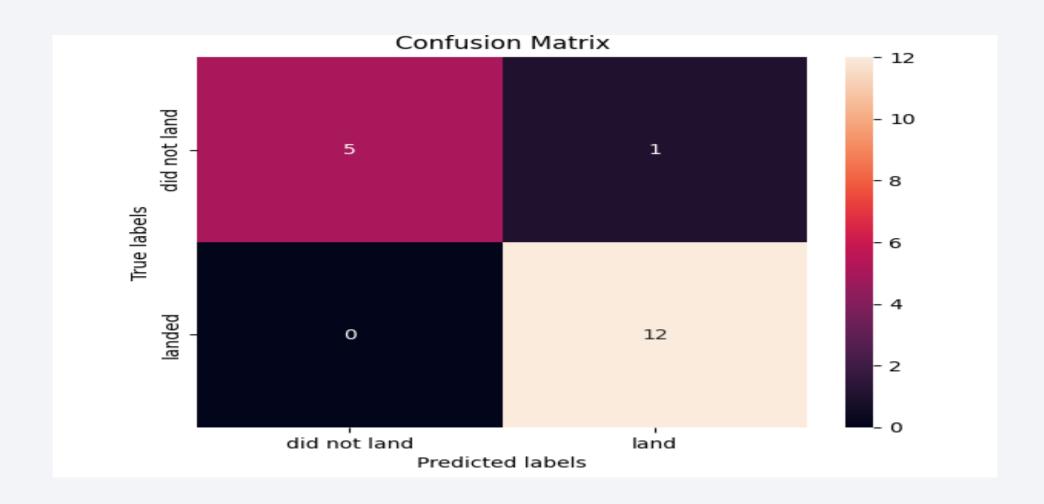




Classification Accuracy



Confusion Matrix



Conclusions

- Data showed interesting insights into SpaceX progress over time and success factors
- A model was built with decent success to create launch outcomes

Appendix

• Link to full repository

