

Winning Space Race with Data Science

<Name> <Date>



Outline

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

Executive Summary

Methodologies

- data collection with SpaceX API and web scraping from Wikipedia
- EDA with visualization

EDA with SQL

- Dashboard for interactive data analysis and geospacer analysis
- Machine learning classification models evaluation

Summary of all results

- Relationship between variables
- Model selection
- Effective prediction

Introduction

• Project background and context

SpaceX is a private aerospace agency that provides services of transportation and launches using the Falcon 9 rocket; a rocket that can land its first stage for it to be reutilized, reducing cost in historical amounts.

Problem

We want to know what determines if a landing is successful and create a model to make predictions of landing success.





Methodology

Executive Summary

- Data collection methodology:
 - The data was collected using web scraping from Wikipedia and with the SpaceX API
- Perform data wrangling
 - Categorical features were treated with One Hot Encoding
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Sci Kit Learn was used to develop, optimize and test ML models

Data Collection

- With the SpaceX API and get request method the data was collected and later the json from the data was converted to a pandas dataframe.
- The data was cleaned and missing value were treated, taking care of the integrity of the data.
- Web Scraping from Wikipedia was performed in order to obtain launch records of the Falcon 9. Parsing the HTML tables and converting its information to a pandas dataframe.

Data Collection – SpaceX API

Use of the SpaceX API

• Complete Notebook in Git Hub.

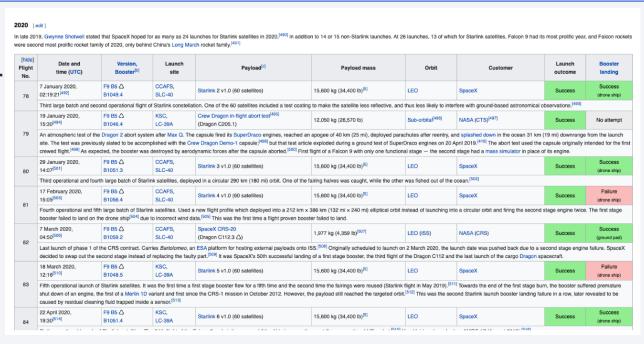
```
spacex_url="https://api.spacexdata.com/v4/launches/past"

response = requests.get(spacex_url)
```

Data Collection - Scraping

 Beautiful Soup was used for the web scraping of data from Wikipedia.

• Complete Notebook in GitHub.



TASK 1: Request the Falcon9 Launch Wiki page from its URL

First, let's perform an HTTP GET method to request the Falcon9 Launch HTML page, as an HTTP response.

```
# use requests.get() method with the provided static_url
# assign the response to a object
data = requests.get(static_url).text

Create a BeautifulSoup object from the HTML response

# Use BeautifulSoup() to create a BeautifulSoup object from a response text content
soup = BeautifulSoup(data, 'html5lib')
```

Data Wrangling

- First, we examined the missing values and data types.
- We differentiate between numerical and categorical data type in columns.
- Created a "class" column to make the launch success or fail a numerical value with 1 and 0.
- Complete Notebook in GitHub.

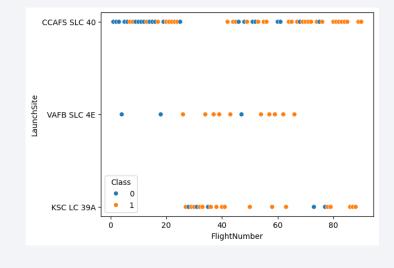
EDA with Data Visualization

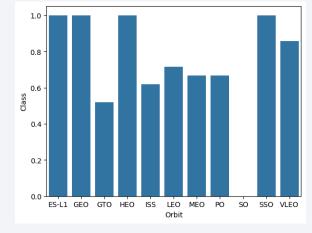
- We created mainly scatter plots colored by the outcome of the landing to understand the relationship between magnitudes like flight number and payload mass with other variables like launch site or orbit type.
- Bar plot to see the success rate of each orbit type.
- Line plot to show a time series for the success of the landings.

0.2

2010 2012 2013 2014 2015 2016 2017 2018 2019 2020

Notebook to see the visualizations.





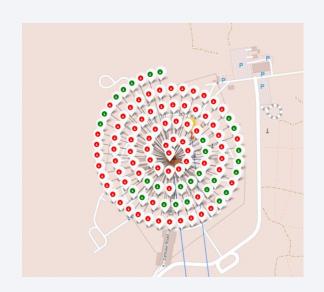
EDA with SQL

- Queries to find the name of all launch sites and the launch sites with "CCA" in its name.
- Query to find the total payload mass carried and launched by NASA
- Query to find mean payload mass carried by the Falcon 9 v1.1
- Query to find the date of the first successful landing
- Query to find the name of the boosters with a successful landing in drone ship and that are in a specific payload mass range.
- Query to find the number of success vs failure in missions
- Query to find the booster versions that have carried more mass
- Query to summarize the data for drone ship failures
- Query to rank the landing outcomes
- Notebook with all the SQL queries

Build an Interactive Map with Folium

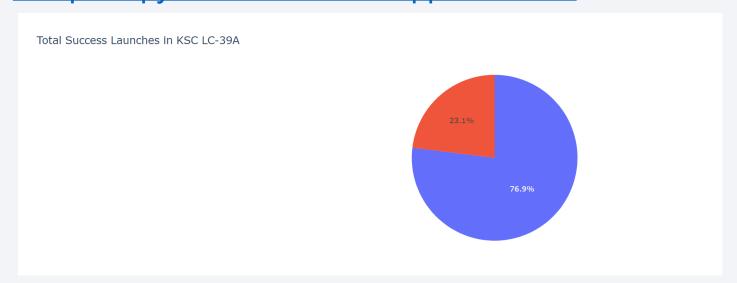
- We marked each launch site and its success and failure cases. We also measured its distance to coastlines, cities, and other important locations to understand if there was any relation between the outcome and the nature of the launch location
- Complete Notebook in GitHub.





Build a Dashboard with Plotly Dash

- The dashboard lets the user select the between each or all launch site and then it shows a pie graph with the success vs failure outcomes of the selection and also a scatter plot of the payload mass vs the outcome, coloring the points by its booster version.
- With these characteristics it is easier to find relationships between the main variables in our analysis.
- Complete python code for the app in GitHub.



Predictive Analysis (Classification)

- With our data, first we divided the variables between dependent and independent variables. Then the train and test set were selected with 80% and 20% respectively.
- Testing for logistic regression, support vector machine, decision tree and Knearest neighbor, first its respective hyperparameters were optimized with GridSearchCV.
- We calculate the accuracy of each model and obtain a confusion matrix plot. Finally comparing its accuracy, we can select the better one.
- Complete Notebook with the development of models.

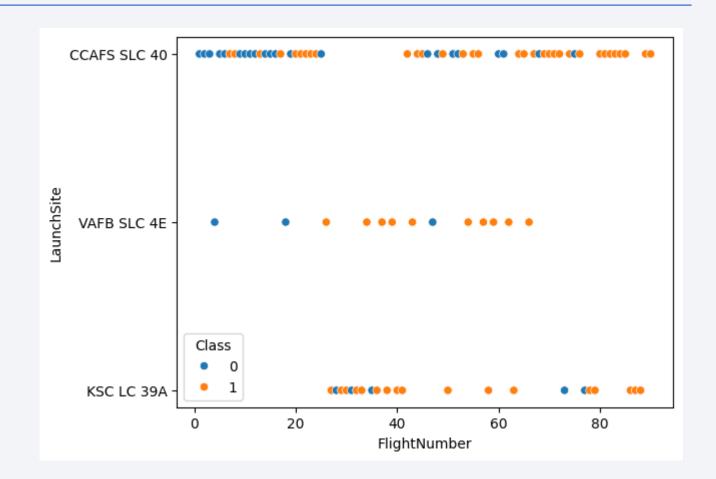
Results

- We found that the variables *Flight Number*, *Payload Mass*, *Launch Site* and *Orbit Type* are related to the outcome of the mission; and we can use it to develop a statistical model.
- A dashboard for gaining insight from the data was created
- The decision tree model had the highest accuracy



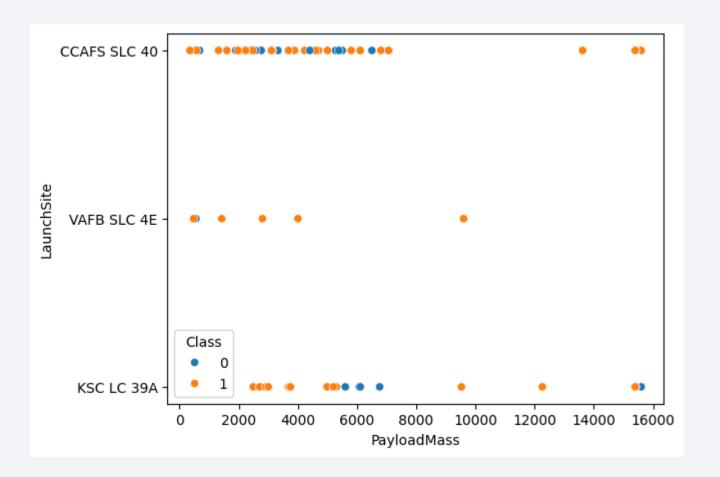
Flight Number vs. Launch Site

- Figure: Flight Number vs. Launch Site
- For the CCAFS SLC 40 and VAFB SLC 4E launch sites, we can see that after the flight number 20 there are more successful landings.
- While for the KSC LC 39A there aren't any launches before but has more successful landings regularly.



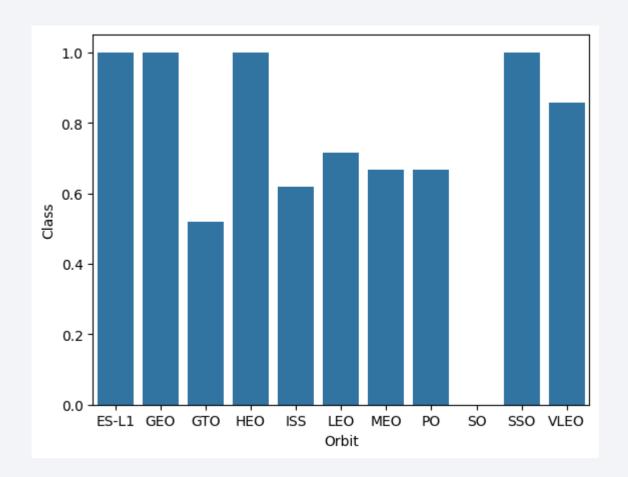
Payload vs. Launch Site

- Figure: Payload vs. Launch
 Site
- High payload mass launches are less common
- With very few launches, the VAFB SLC 4E launch site has mainly successful landings



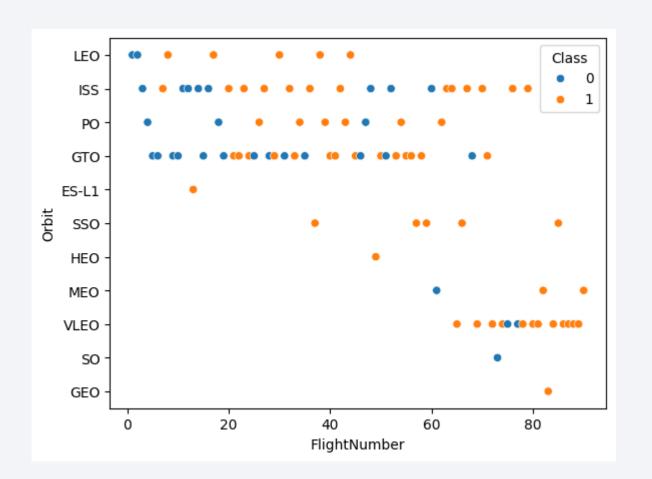
Success Rate vs. Orbit Type

- Figure: Bar plot with the success rate for each orbit type
- ES-L1, GEO, HEO and SSO
 have success rate of 100%,
 while SO of 0% however, we
 can't know from the plot how
 big or small is the population
 of each orbit type.



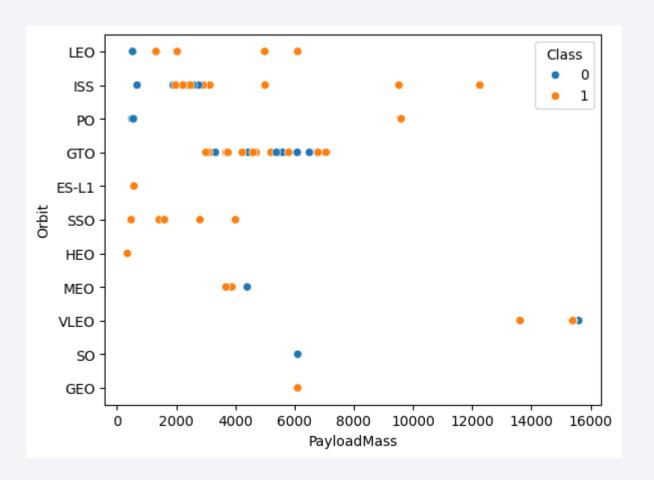
Flight Number vs. Orbit Type

- Flight number vs. Orbit type
- This plot solves some problems from the previous bar plot. We now can see that GEO, SO, HEO and ES-L1 have very low population and making conclusions from there might be problematic without caution. However, for SSO the 100% success rate is well populated
- LEO, ISS, PO and MEO seem to have upgraded their success frequency with flight number.



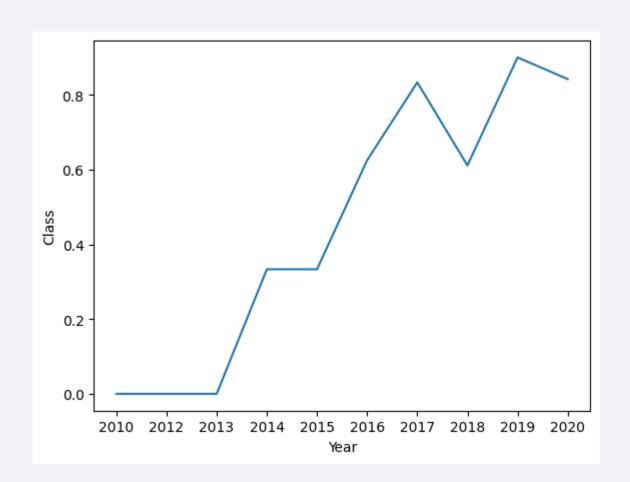
Payload vs. Orbit Type

- Figure payload vs. orbit type
- ISS and VLEO orbits use to have higher payload mass than others
- LEO, GTO, SSO and MEO use to have lower payload mass



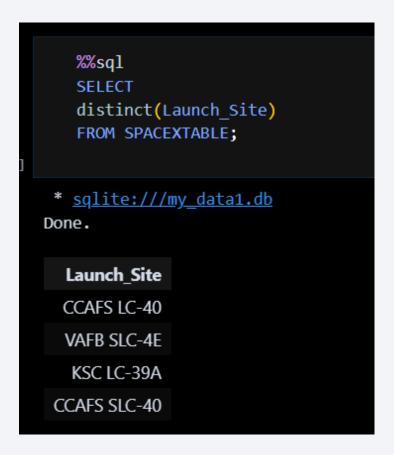
Launch Success Yearly Trend

- Figure: yearly average success rate
- It is notorious that with time each year often has a higher success rate
- Between 2015 and 2017 there was a big acceleration in the success rate change
- 2015, 2018 and 2020 are the exception for the growth in the success rate



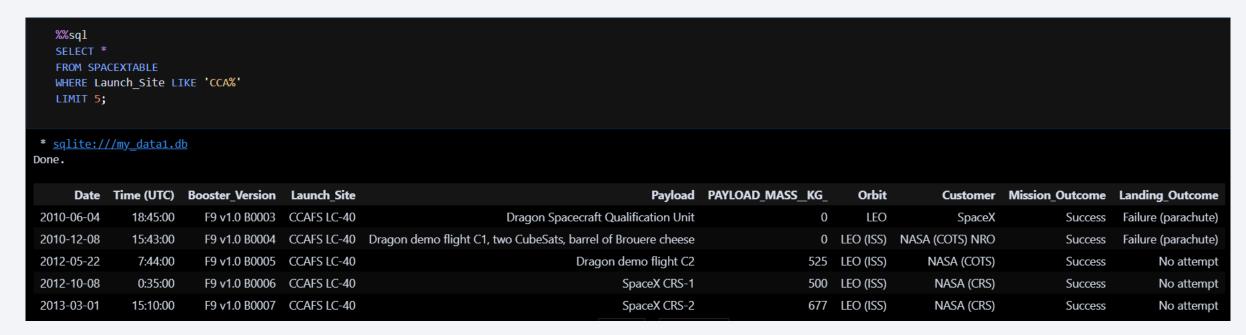
All Launch Site Names

• To not repeat the Launch_Site values, the use of DISTINCT(Launch_Site) is needed.



Launch Site Names Begin with 'CCA'

 We used the WHERE clause with a LIKE 'CCA%' to select only the values for Launch_Site that start with CCA



Total Payload Mass

 With a sum of the PAYLOAD_MASS_KG and selecting only the launches for NASA as Customer we can find the total payload mass for NASA launches to be of 48,213 kg

```
%%sql
SELECT
sum(PAYLOAD_MASS__KG_)
FROM SPACEXTABLE
WHERE Customer LIKE '%NASA (CRS)%';

* sqlite:///my_data1.db
Done.

sum(PAYLOAD_MASS__KG_)
48213
```

Average Payload Mass by F9 v1.1

• We can calculate the average payload mass carried by booster version F9 v1.1 using the function AVG(PAYLOAD_MASS_KG_) and selecting with the WHERE clause only the F9 v1.1 Booster_Version. Finding that it is of 2928.4 kg.

```
%%sql
SELECT
avg(PAYLOAD_MASS__KG_)
FROM SPACEXTABLE
WHERE Booster_Version LIKE 'F9 v1.1';

* sqlite://my_data1.db
Done.

avg(PAYLOAD_MASS__KG_)
2928.4
```

First Successful Ground Landing Date

• With the WHERE clause we can find the Landing_Outcome 'Success (ground pan)' and select the date only, ordering with ORDER BY in ascendent order and selecting the first element, we find the date of the first success landing in ground pad to be December 22 of 2015.

```
%%sql
SELECT
"Date"
FROM SPACEXTABLE
WHERE Landing_Outcome LIKE 'Success (ground pad)'
ORDER BY "Date" ASC LIMIT 1;

* sqlite:///my_data1.db
Done.

Date
2015-12-22
```

Successful Drone Ship Landing with Payload between 4000 and 6000

• To list the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000, first we select the Booster_Version column from our table, and we used the WHERE clause with double statement, PAYLOAD_MASS_KG has to be in the given range, and also to have Landing_Outcome as 'Success (drone ship)

```
%%sql
SELECT
Booster_Version
FROM SPACEXTABLE
WHERE PAYLOAD_MASS__KG__ BETWEEN 4000 AND 6000
AND Landing_Outcome LIKE 'Success (drone ship)';

* sqlite:///my_data1.db
Done.

Booster_Version
    F9 FT B1022
    F9 FT B1026
F9 FT B1021.2
F9 FT B1031.2
```

Total Number of Successful and Failure Mission Outcomes

 To calculate the total number of successful and failure mission outcomes we used two subqueries where each one counts the elements of the selection success of failure.

```
%%sql
SELECT
(SELECT count(*) FROM SPACEXTABLE WHERE upper(Mission_Outcome) LIKE '%SUCCESS%') Success,
(SELECT count(*) FROM SPACEXTABLE WHERE upper(Mission_Outcome) LIKE '%FAILURE%') Failure;

* sqlite:///my_data1.db
Done.

Success Failure
100 1
```

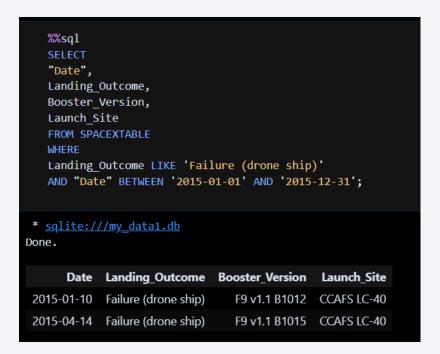
Boosters Carried Maximum Payload

 Selecting the Booster_Version column and with the clause WHERE to select only the PAYLOAD_MASS__KG with the same value of max(PAYLOAD_MASS__KG_) obtained from a subquery, we list the names of the booster which have carried the maximum payload mass

```
%%sql
   SELECT
   Booster Version
   FROM SPACEXTABLE
   WHERE PAYLOAD MASS KG IN (SELECT max(PAYLOAD MASS KG ) FROM SPACEXTABLE);
 * 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
```

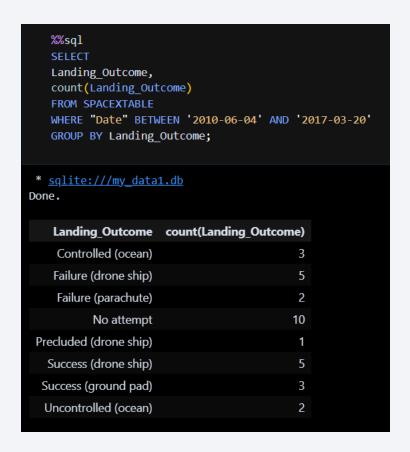
2015 Launch Records

• We list the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015 by selecting the mentioned columns alongside date, and with the clause WHERE select only the 'Failure (drone ship) in Landing_Outcome and the range of 2015 for date.



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

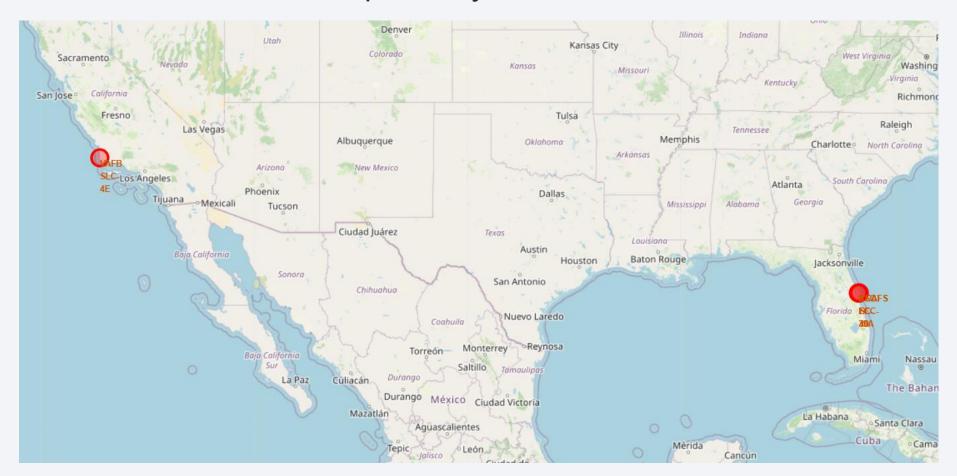
 We rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad))
between the date 2010-06-04 and 2017-03-20, in
descending order by selecting the Landing_Outcome
column and counting its elements from our table,
also with the WHERE clause we select only the data
range mentioned, and we group our elements for the
sum by Langind_Outcome with GROUP BY, finally to
order we use ORDER BY DESC.





Launch Sites with Folium

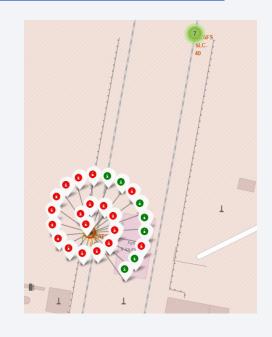
• All launch sites are in the USA, specifically in California and Florida.



Launch Site by Outcome



- We can see the landing outcome of each launch by launch site by zooming in and clicking on the site
- Also each circle is colored for easy evaluation of the success rate





Evaluating Distances

- We can evaluate the distance from the launch sites to different places of interest.
- For example, coastlines or cities

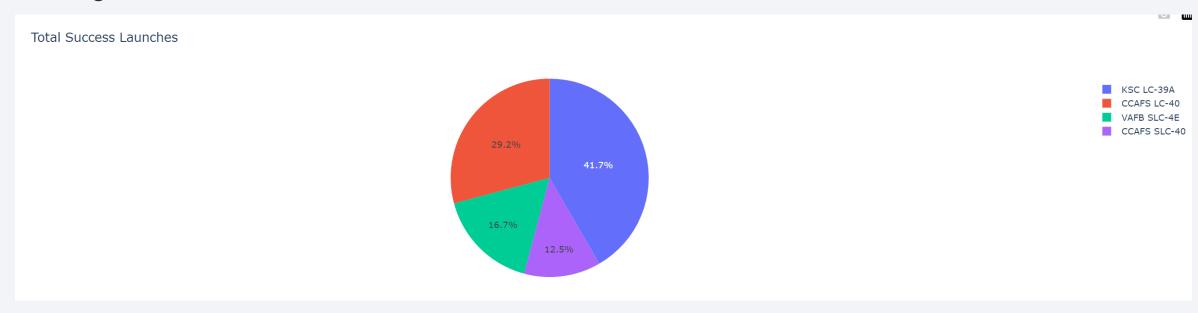






Success Count by Launch Site

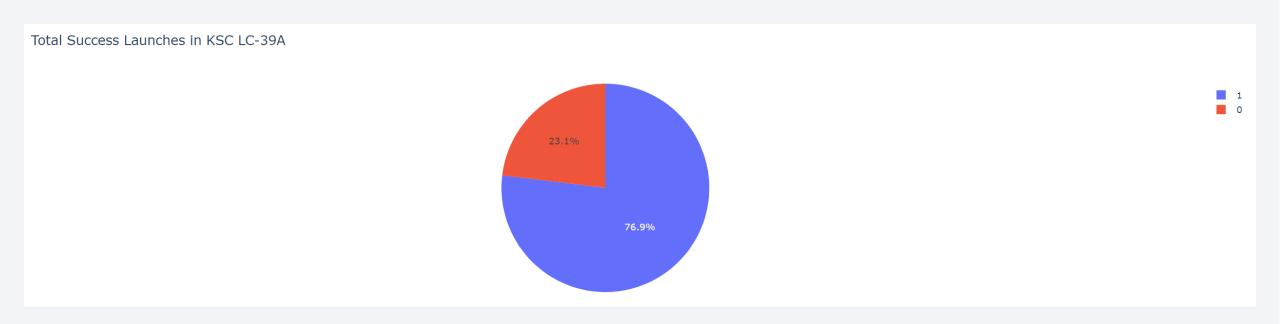
• Figure: Pie chart with the success rate for each launch site



• We can see that the launch site with highest success is KSC LC-39A, with 41.7% of all success

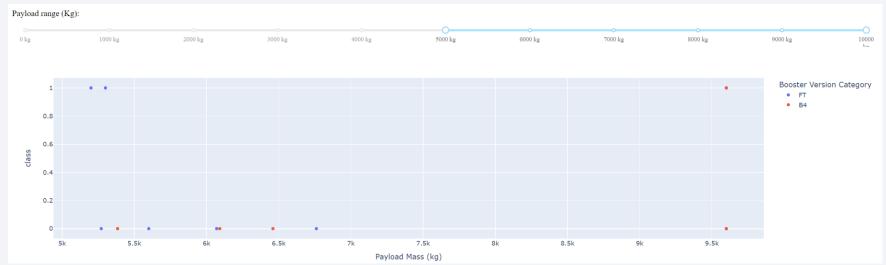
KSC LC-39A success rate

• The success rate of this launch site is 77.9%, with only 23.1% of its landings being classified as failure.



Success and Payload mass





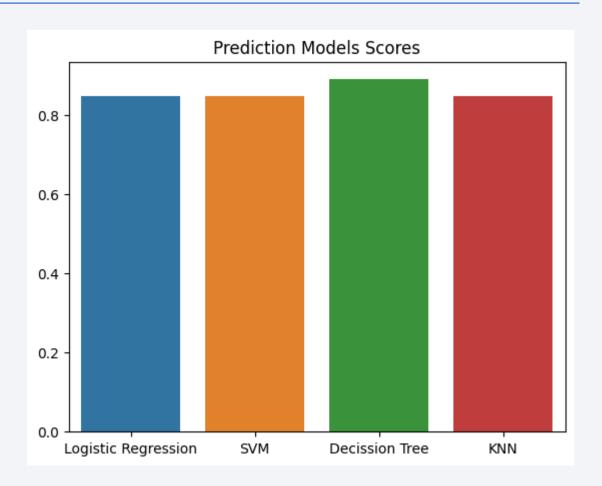
- For heavier payload masses, only FT and B4 booster versions are used
- The success is more common in the lower payload masses



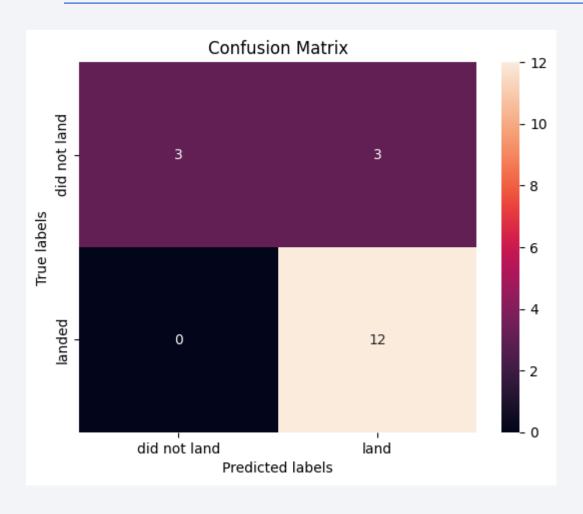
Classification Accuracy

 Figure: bar chart with the accuracy scores for each one of the models evaluated

The highest accuracy is found in the decision tree model



Confusion Matrix



- Figure: confusion matrix for the decision tree model
- The model predicted effectively all the 12 launches that landed
- While for the not successful landing cases, the model predicted effectively that 3 wouldn't land, but failed at other 3 flights that didn't landed and labeled as successful
- The model is better at predicting successful landings than unsuccessful landings



Conclusions

- We found that multiple variables are related between them
- The success rate for the landings have been generally upgrading with time
- The payload mass cargo has been increasing with time, and is significantly influential in the success on a landing
- Multiple classification models were evaluated, with a best fit for a simple decision tree
- We can effectively predict the outcome of a launch mission based on the public data obtained, and therefore use it for a better planning of future missions

Appendix

• Complete GitHub with all relevant files here

