

# Winning Space Race with Data Science

HaiLv 2024/08/15



### Outline

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- Methodology
- Results
- Conclusion
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### **Executive Summary**

- Summary of methodologies
- Data Collection through API
- Data Collection with Web Scraping
- Data Wrangling
- Exploratory Data Analysis with SQL
- Exploratory Data Analysis with Data Visualization
- Interactive Visual Analytics with Folium
- Machine Learning Prediction
- Summary of all results
- Exploratory Data Analysis result
- Interactive analytics in screenshots
- Predictive Analytics result

#### Introduction

#### Project background and context

SpaceX 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 SpaceX for a rocket launch. This goal of the project is to create a machine learning pipeline to predict if the first stage will land successfully.

#### Problems you want to find answers

- What factors determine if the rocket will land successfully?
- The interaction amongst various features that determine the success rate of a successful landing.
- What operating conditions needs to be in place to ensure a successful landing program.



### Methodology

#### **Executive Summary**

- Data collection methodology:
  - Data from SpaceX was obtained from 2 sources:
    - SpaceX API(<u>https://api.spacexdata.com/v4/rockets/</u>)
  - Web Scraping (https://en.wikipedia.org/wiki/List\_of\_Falcon/\_9/\_and\_Falcon\_Heavy\_I aunches)
- Perform data wrangling
  - Collected data was enriched by creating a landing outcome label based on outcome data after summarizing and analyzing features

# Methodology

#### **Executive Summary**

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Data that was collected until this step were normalized, divided in training and test data sets and evaluated by four different classification models, being the accuracy of each model evaluated using different combinations of parameters

#### **Data Collection**

- Describe how data sets were collected.
- Datasets were collected from SpaceX API
   (https://api.spacexdata.com/v4/rockets/) and from Wikipedia
   (https://en.wikipedia.org/wiki/List\_of\_Falcon/\_9/\_and\_Falcon\_Heavy\_launches),
   using web scraping techniques

### Data Collection – SpaceX API

- SpaceX offers a public API from where data can be obtained and then used;
- This API was used according to the flowchart beside and then data is persisted.
- Source:

   <a href="https://github.com/Lucky-Lee-029/Da">https://github.com/Lucky-Lee-029/Da</a>

   <a href="task-spa">tascience/blob/main/jupyter-labs-spa</a>

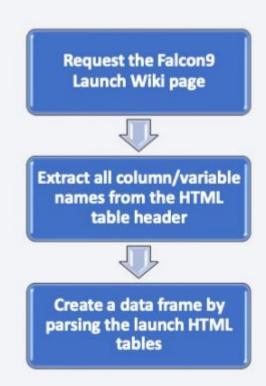
   <a href="cex-data-collection-api.ipynb">cex-data-collection-api.ipynb</a>



### Data Collection - Scraping

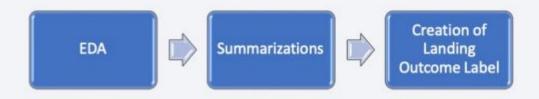
- Data from SpaceX launches can also be obtained from Wikipedia;
- Data are downloaded from Wikipedia according to the flowchart and then persisted.
- Source:

   https://github.com/Lucky-Lee-029/Data
   Science/blob/main/jupyter-labs-webscr aping.ipynb



### **Data Wrangling**

- Initially some Exploratory Data Analysis (EDA) was performed on the dataset.
- Then the summary launches per site, occurrences of each orbit and occurrences of mission outcome per orbit type were calculated.
- Finally, the landing outcome label was created from Outcome column.



• Source:

https://github.com/Lucky-Lee-029/DataScience/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb

#### **EDA** with Data Visualization

- The following SQL queries were performed:
- Names of the unique launch sites in the space mission;
- Top 5 launch sites whose name begins with the string 'CCA';
- 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 achieved;
- Names of the boosters which have success in drone ship and have payload mass between 4000 and 6000 kg;
- Total number of successful and failure mission outcomes;
- Names of the booster versions which have carried the maximum payload mass;
- Failed landing out comes in droneship, their booster versions, and launch site names for in year 2015;
- Rank of the count of landing outcomes (such as Failure (droneship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20.
- Source: <a href="https://github.com/Lucky-Lee-029/DataScience/blob/main/edadataviz.ipynb">https://github.com/Lucky-Lee-029/DataScience/blob/main/edadataviz.ipynb</a>

#### **EDA** with SQL

- To explore data, scatterplots and bar plots were used to visualize the relationship between pair of features:
- Payload Mass X Flight Number, Launch Site X Flight Number, Launch Site X Payload Mass, Orbit and Flight Number, Payload and Orbit
- Source:

https://github.com/Lucky-Lee-029/DataScience/blob/main/jupyter-labs-eda-sql-coursera\_sqllite.ipynb

#### Build an Interactive Map with Folium

- Markers, circles, lines and marker clusters were used with Folium Maps
- Markers indicate points like launch sites;
- Circles indicate highlighted areas around specific coordinates, like NASA Johnson Space Center;
- Marker clusters indicates groups of events in each coordinate, like launches in a launch site; and
- Lines are used to indicate distances between two coordinates.
- Source: <u>https://github.com/Lucky-Lee-029/DataScience/blob/main/lab\_jupyter\_launch\_site\_location.ipynb</u>

#### Build a Dashboard with Plotly Dash

- We built an interactive dashboard with Plotly dash
- We plotted pie charts showing the total launches by a certain sites
- We plotted scatter graph showing the relationship with Outcome and Payload Mass (Kg) for the different booster version
- Source:

https://github.com/Lucky-Lee-029/DataScience/blob/main/spacex\_dash\_app.py

## Predictive Analysis (Classification)

- We loaded the data using numpy and pandas, transformed the data, split our data into training and testing.
- We built different machine learning models and tune different hyperparameters using GridSearchCV.
- We used accuracy as the metric for our model, improved the model using feature engineering and algorithm tuning.
- We found the best performing classification model.
- Source: <u>https://github.com/Lucky-Lee-029/DataScience/blob/main/SpaceX\_Machine%20Learning%20Prediction\_Part\_5.ipynb</u>

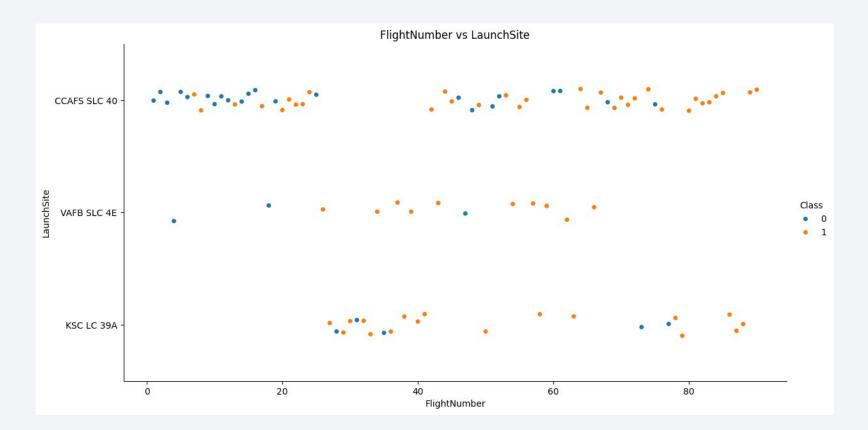
#### Results

- SpaceX uses 4 different launch sites;
- The first launches were done to SpaceX itself and NASA;
- The average payload of F9 v1.1 booster is 2,928 kg;
- The first success landing outcome happened in 2015 fiver year after the first launch;
- Many Falcon 9 booster versions were successful at landing in drone ships having payload above the average; Almost 100% of mission outcomes were successful;
- Two booster versions failed at landing in drone ships in 2015: F9 v1.1 B1012 and F9 v1.1 B1015;
- The number of landing outcomes became as better as years passed.

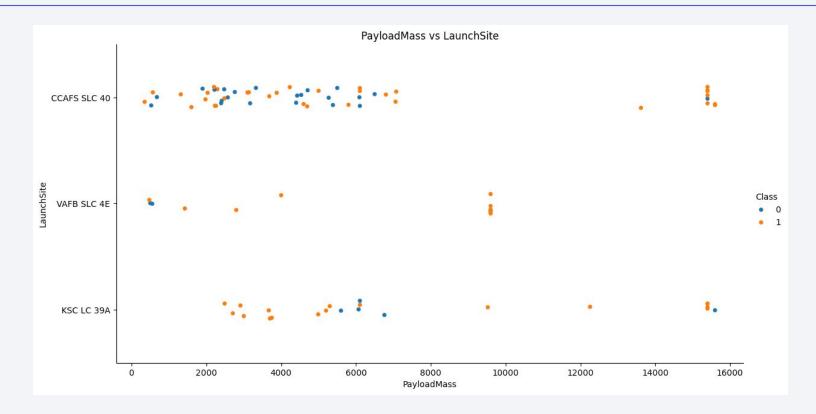


### Flight Number vs. Launch Site

• From the plot, we found that the larger the flight amount at a launch site, the greater the success rate at a launch site.

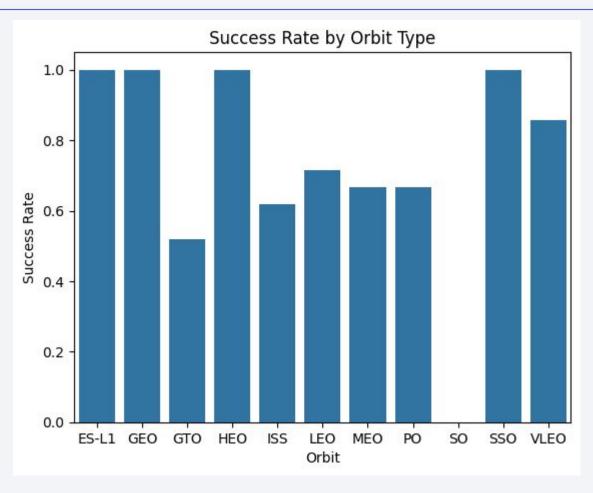


# Payload vs. Launch Site



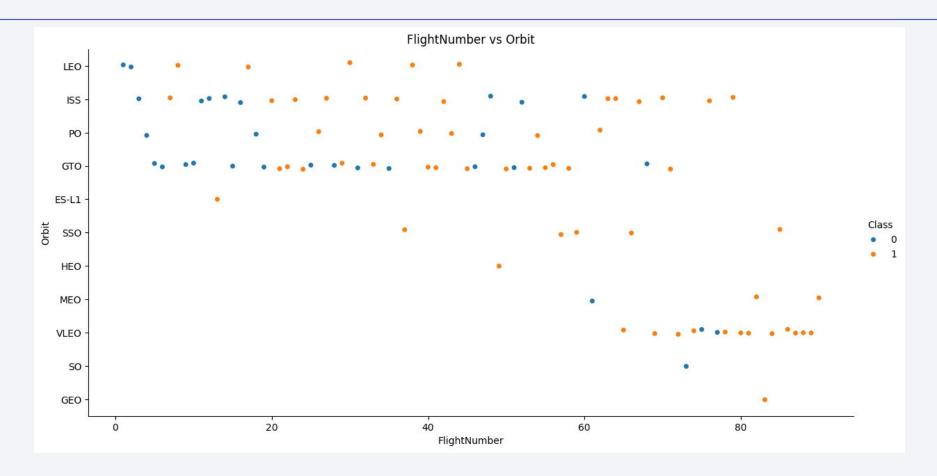
The higher payload mass, the higher successful rate for the rocket

# Success Rate vs. Orbit Type



ES-L1, GEO, HEO, SSO, VLEO had the most success rate

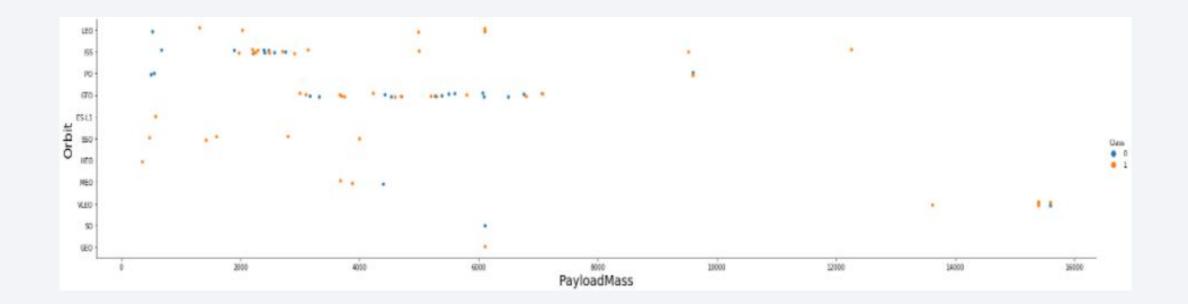
# Flight Number vs. Orbit Type



We observe that in the LEO orbit, success is related to the number of flights whereas in the GTO orbit, there is no relationship between flight number and the orbit.

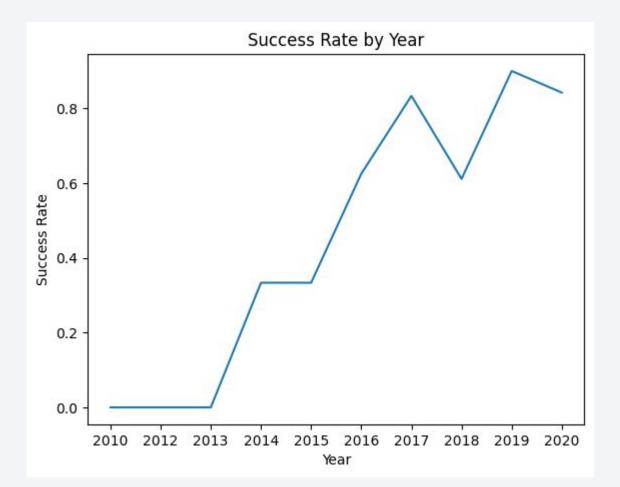
# Payload vs. Orbit Type

• We can observe that with heavy payloads, the successful landing are more for PO, LEO and ISS orbits.



## Launch Success Yearly Trend

From the plot, we can observe that success rate since 2013 kept on increasing till 2020.



#### All Launch Site Names

• Use DISTINCT to find to unique name of launch site:



# Launch Site Names Begin with 'CCA'

• LIMIT 5 records that names begin with 'CCA'

<pre>%sql select * from SPACEXTABLE where Launch_Site like 'CCA%' limit 5  * sqlite:///my_data1.db Done.</pre>									
2010- 06-04	18:45:00	F9 v1.0 B0003	CCAFS LC- 40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
2010- 12-08	15:43:00	F9 ∨1.0 B0004	CCAFS LC- 40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
2012- 05-22	7:44:00	F9 v1.0 B0005	CCAFS LC-	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
2012- 10-08	0:35:00	F9 v1.0 B0006	CCAFS LC-	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
2013-	15:10:00	F9 v1.0 B0007	CCAFS LC-	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

#### **Total Payload Mass**

 We calculated the total payload carried by boosters from NASA as 45596 using the query below

```
Display the total payload mass carried by boosters launched by NASA (CRS)

%sql select sum(PAYLOAD_MASS__KG_) from SPACEXTABLE where Customer = "NASA (CRS)"

* sqlite://my_data1.db
)one.

sum(PAYLOAD_MASS__KG_)

45596
```

### Average Payload Mass by F9 v1.1

 We calculated the total payload carried by boosters F9 v1.1 using the query below

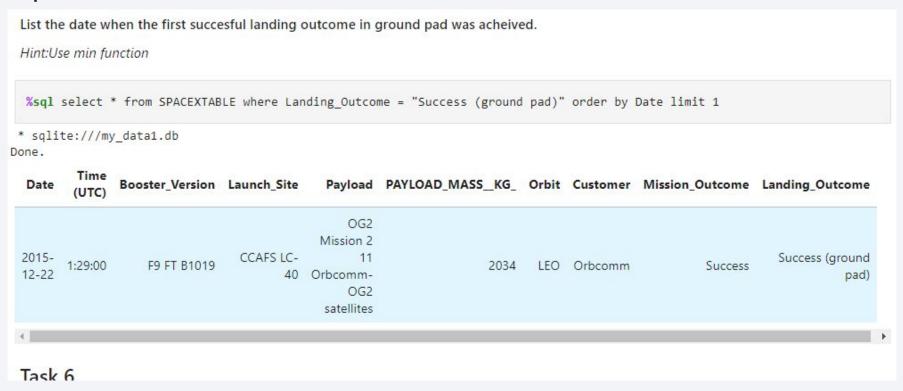
```
%sql select avg(PAYLOAD_MASS__KG_) from SPACEXTABLE where Booster_Version like "F9 v1.1%"

* sqlite:///my_data1.db
Done.
avg(PAYLOAD_MASS__KG_)

2534.6666666666665
```

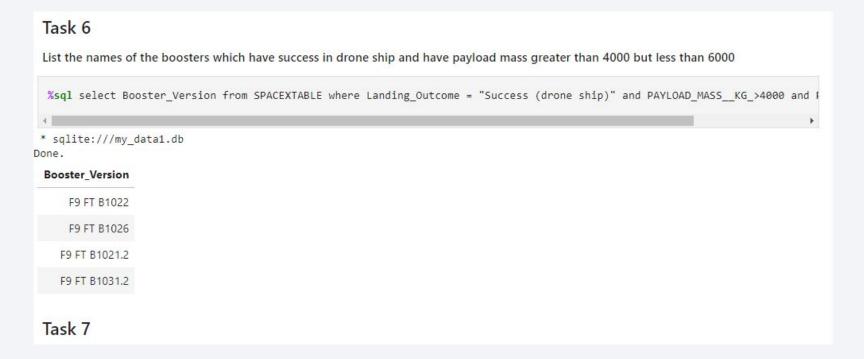
### First Successful Ground Landing Date

 We observed that the dates of the first successful landing outcome on ground pad was 22nd December 2015



#### Successful Drone Ship Landing with Payload between 4000 and 6000

 We used the WHERE clause to filter for boosters which have successfully landed on drone ship and applied the AND condition to determine successful landing with payload mass greater than 4000 but less than 6000



#### Total Number of Successful and Failure Mission Outcomes

 We used wildcard like '%' to filter for WHERE MissionOutcome was a success or a failure.

```
List the total number of successful and failure mission outcomes
 countSuccess = %sql select count(*) from SPACEXTABLE WHERE Landing Outcome like "Success%"
  countFail = %sql select count(*) from SPACEXTABLE WHERE Landing Outcome like "Failure%"
  print(countSuccess)
 print(countFail)
* sqlite:///my_data1.db
Done.
* sqlite:///my_data1.db
Done.
+----+
 count(*)
+----+
 count(*)
+----+
    10
+----+
```

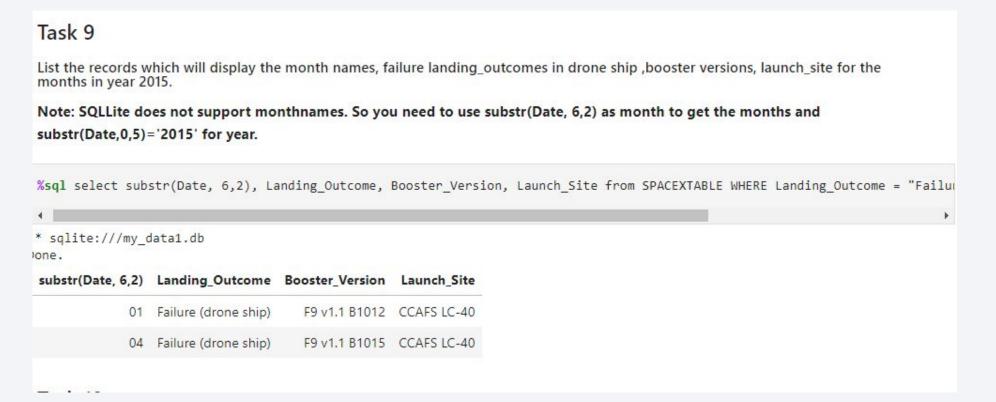
### **Boosters Carried Maximum Payload**

 We determined the booster that have carried the maximum payload using a subquery in the WHERE clause and the MAX() function.



#### 2015 Launch Records

 We used a combinations of the WHERE clause, LIKE, AND, and BETWEEN conditions to filter for failed landing outcomes in drone ship, their booster versions, and launch site names for year 2015e



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

 We selected Landing outcomes and the COUNT of landing outcomes from the data and used the WHERE clause to filter for landing outcomes BETWEEN 2010-06-04 to 2010-03-20.
 We applied the GROUP BY clause to group the landing outcomes and the ORDER BY clause to order the grouped landing outcome in descending order.

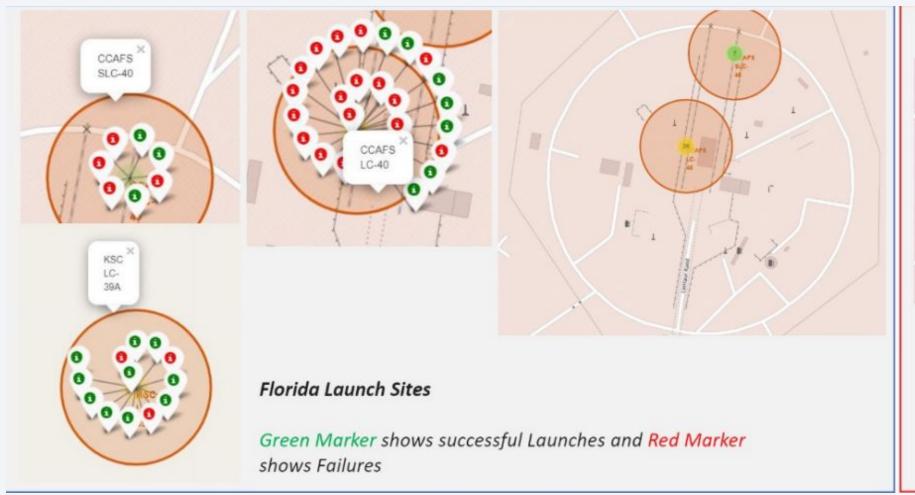


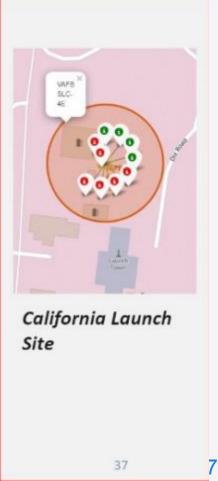


### All launch sites global map markers

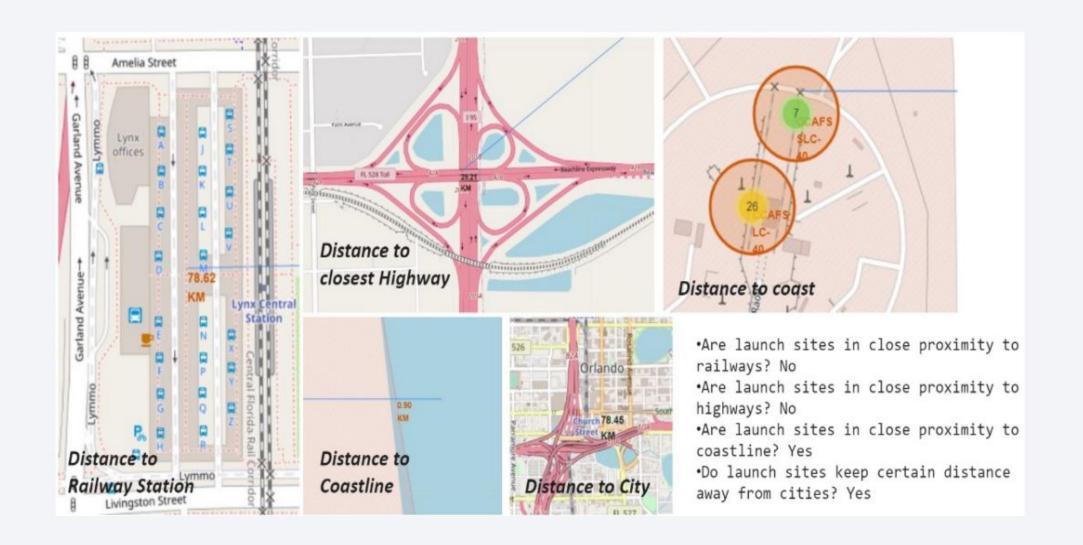


# Markers showing launch sites with color labels



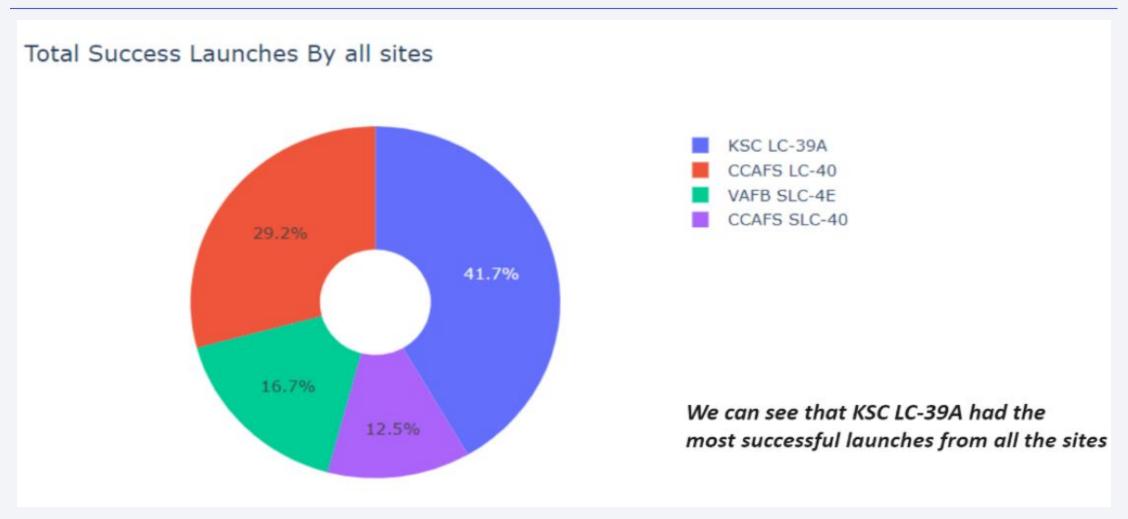


#### Launch Site distance to landmarks

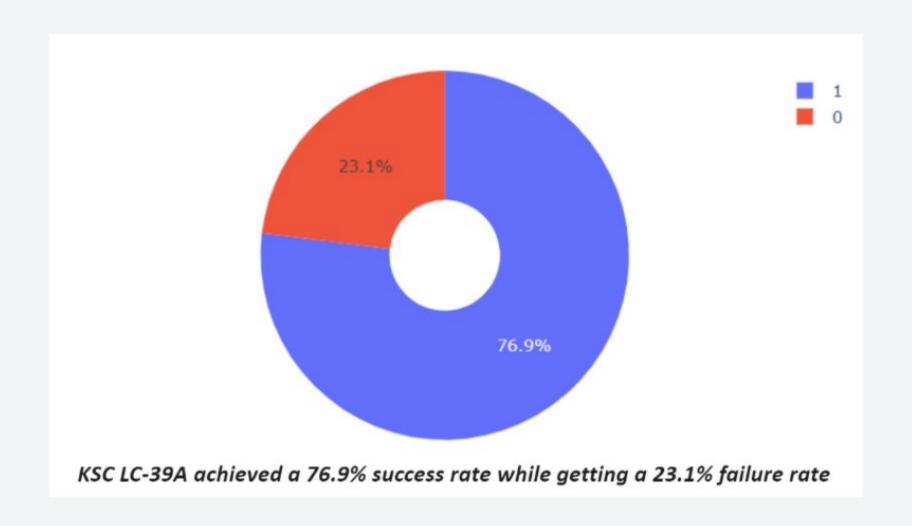




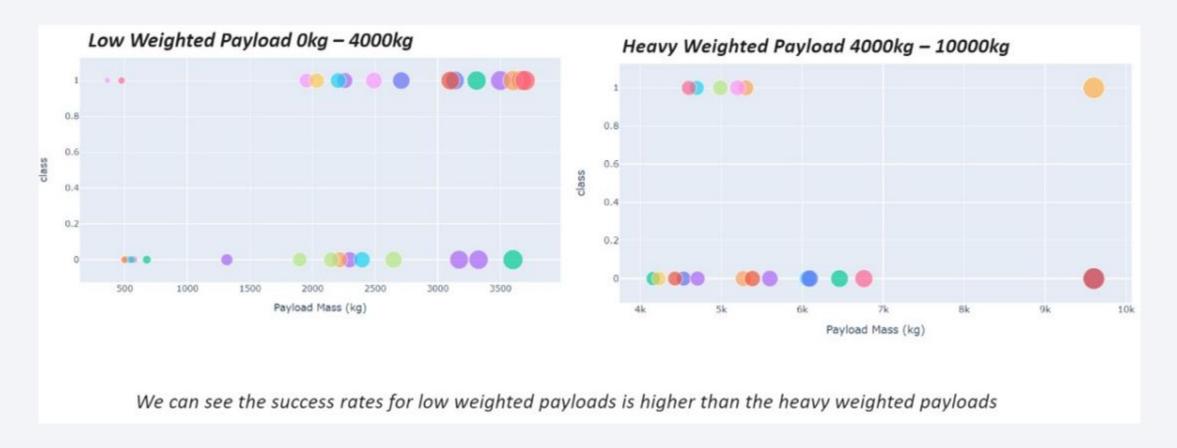
#### Pie chart showing the success percentage achieved by each launch site



#### Pie chart showing the Launch site with the highest launch success ratio



# Scatter plot of Payload vs Launch Outcome for all sites, with different payload selected in the range slider





### **Classification Accuracy**

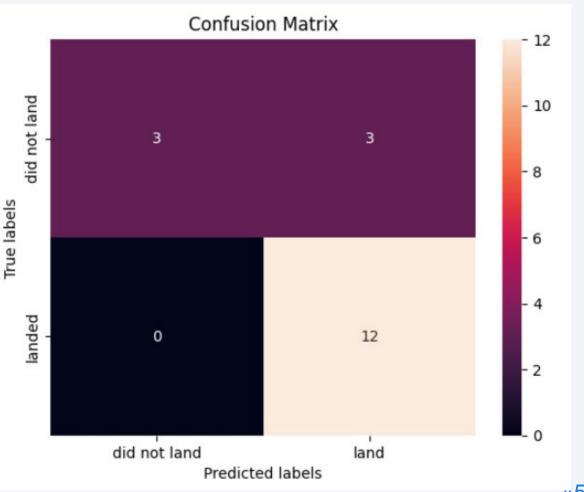
• The decision tree classifier is the model with the highest classification accuracy

```
print("tuned hpyerparameters :(best parameters) ",tree_cv.best_params_)
print("accuracy :",tree_cv.best_score_)

tuned hpyerparameters :(best parameters) {'criterion': 'entropy', 'max_depth': 8, 'max_features': 'sqrt', 'min_samples_lea
f': 4, 'min_samples_split': 2, 'splitter': 'random'}
accuracy : 0.8767857142857143
```

#### **Confusion Matrix**

 The confusion matrix for the decision tree classifier shows that the classifier can distinguish between the different classes. The major problem is the false positives .i.e., unsuccessful landing marked as successful landing by the classifier.



#### Conclusions

- The larger the flight amount at a launch site, the greater the success rate at a launch site.
- Launch success rate started to increase in 2013 till 2020.
- Orbits ES-L1, GEO, HEO, SSO, VLEO had the most success rate.
- KSC LC-39A had the most successful launches of any sites.
- The Decision tree classifier is the best machine learning algorithm for this task.

