

# Winning Space Race with Data Science

Tinh Nhat Pham 1/1/2022



#### Outline

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

### **Executive Summary**

- Summary of methodologies
  - Data collection
  - Data wrangling
  - EDA with data visualization
  - EDA with SQL
  - Build an Interactive Map with Folium
  - Build 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

- 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 SpaceX 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
- Problems you want to find answers
  - Discover new insights by analyzing Falcon 9 rocket launch data in the first stage.
  - Predict whether or not the first stage will successfully land.



#### Methodology

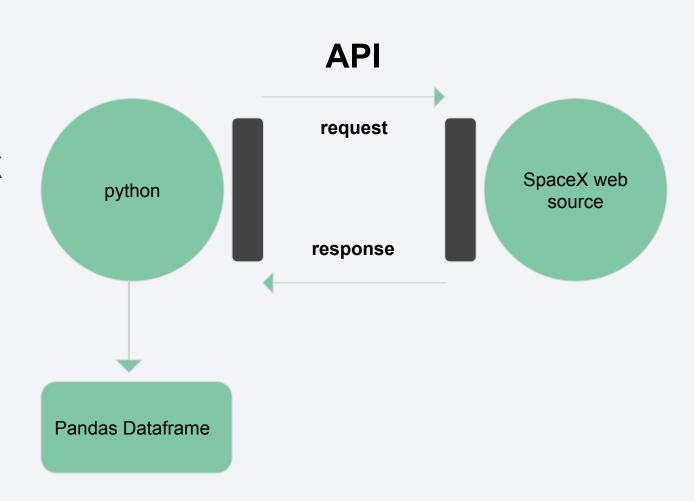
#### **Executive Summary**

- 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 – SpaceX API

#### Key phrases:

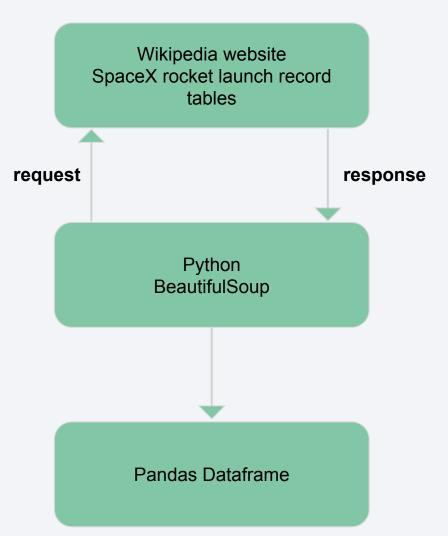
- Send request to get rocket launch data from SpaceX website
- Receive response from the SpaceX website
- Parse response JSON and store in to Pandas Dataframe
- GitHub URL: <u>Click here</u>



#### Data Collection - Scraping

#### Key phrases:

- Send request to get rocket launch record tables from SpaceX Wikipedia
- Receive response from the SpaceX Wikipedia
- Parse response content using BeautifulSoup API
- Store data into Pandas Dataframe
- GitHub URL: <u>Click here</u>

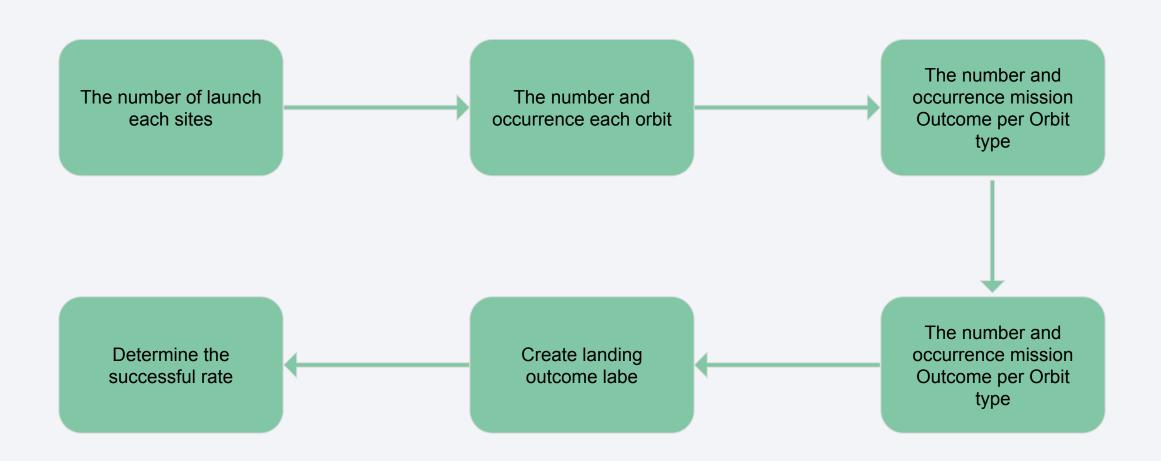


### **Data Wrangling**

#### Key phrases:

- Calculate the number of launches on each site
- Calculate the number and occurrence of each orbit
- Calculate the number and occurrence of mission outcome per orbit type
- Create a landing outcome label from Outcome column
- Determine the successful rate
- GitHub URL: Click here

# Data Wrangling (cont.)



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

- Scatter plot: FlightNumber vs PayMassLoad
  - Check the relationship between flight number vs rocket payload mass based on landing outcome
- Scatter plot: FlightNumber vs Launch Site
  - Check the relationship between flight number vs launch sites based on landing outcome
- Scatter plot: PayloadMass vs Launch Site
  - Check the relationship between flight number vs launch sites based on landing outcome

# EDA with Data Visualization (cont.)

- Bar plot: Successful rate vs Orbit
  - Visualize the relationship between successful rate of each orbit type
- Scatter plot: FlightNumber vs Orbit type
  - Visualize the relationship between Flight number and Orbit type
- Scatter plot: PayloadMass vs Orbit type
  - Visualize the relationship between Payload mass and Orbit type
- Line plot: Successful rate over the years (2010 2020)
  - Visualize the successful rate from 2010 2020, which kept increasing since 2013 till 2020
- GitHub URL: <u>Click here</u>

#### **EDA** with SQL

- Query the names of the unique launch sites in the space mission
- Query the names of the unique launch sites in the space mission
- Query 5 records where launch sites begin with the string 'CCA'
- Query the total payload mass carried by boosters launched by NASA (CRS)
- Query average payload mass carried by booster version F9 v1.1
- List the date when the first successful landing outcome in ground pad was achieved.

# EDA with SQL (cont.)

- List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- List the total number of successful and failure mission outcomes
- List the names of the booster\_versions which have carried the maximum payload mass. Use a subquery
- List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- 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
- GitHub URL: Click here

#### Build an Interactive Map with Folium

- Create circle object to show the location of NASA center and marker to show the name NASA JSC on the folium map
- Create circle objects to show the locations of Launch Sites and marker objects to show the names of the Launch Sites on the folium map
- Create marker objects to show the success/failed launches for each site on the map, and add to the Cluster marker object
- Calculate the distances between the CCAFS SLC-40 to the nearest coastline, highway, rainway, and city. Then draw the lines of the distances.
- GitHub URL: <u>Click here</u>

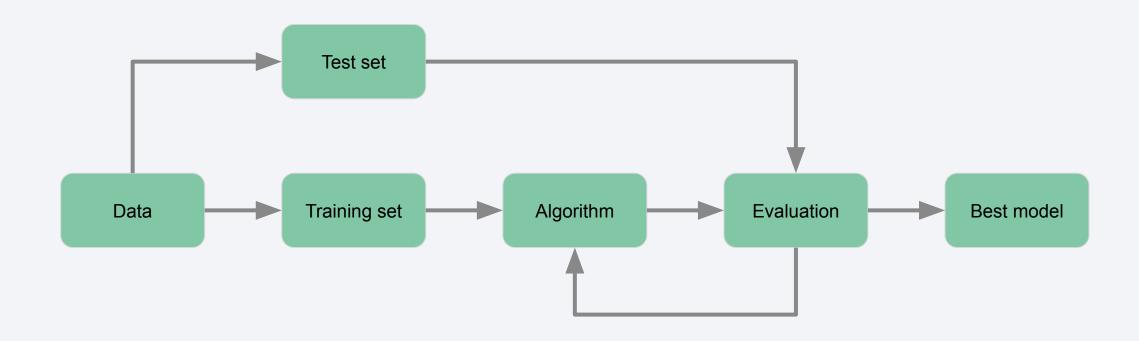
#### Build a Dashboard with Plotly Dash

- Summarize what plots/graphs and interactions you have added to a dashboard
- Explain why you added those plots and interactions
- Add drop down list of Launch sites
- Add a pie chart to show the success/failed launch rate when choosing the site from drop down list
- Add a slider to choose the minimum and maximum payload mass
- Add a scatter plot to show the success/failed launches when choosing the site from drop down list and the min/max payload mass from slider
- GitHub URL: <u>Click here</u>

### Predictive Analysis (Classification)

- Load the data and store in Pandas Dataframe
- Create features (X) and target (Y)
- Standardize the features X
- Create Train set and Test set by using X and Y
- Using GridSearchCV to find the best parameters for each classifier:
  - Logistic Regression
  - Support Vector Machine
  - Decision Tree
  - K Nearest Neighbors
- Calculate performance score each classifier by predict Test set.
- Evaluate the best performing classification model.
- GitHub URL: <u>Click here</u>

# Predictive Analysis (Classification) (cont.)



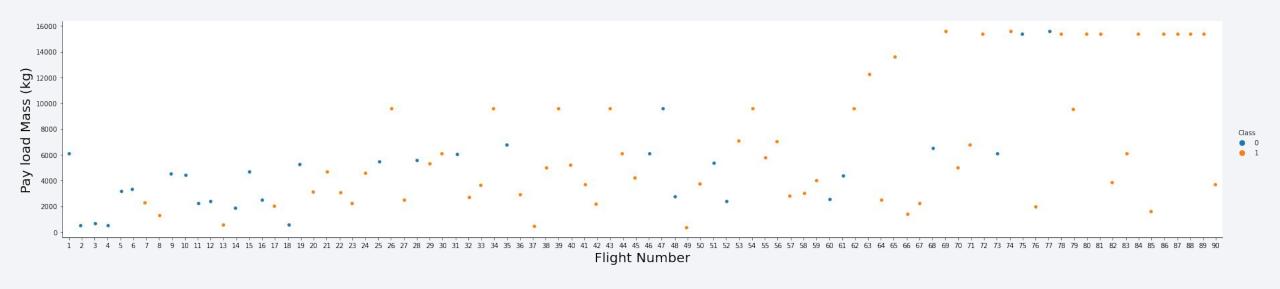
**Model Development Process** 

#### Results

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

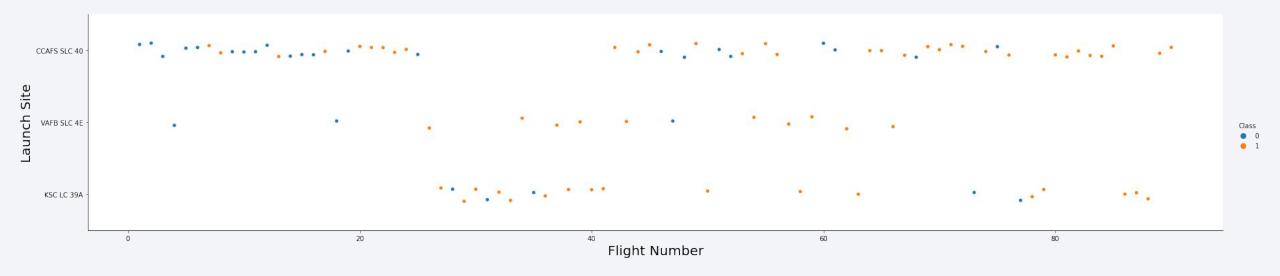


# Flight Number vs. Payload Mass



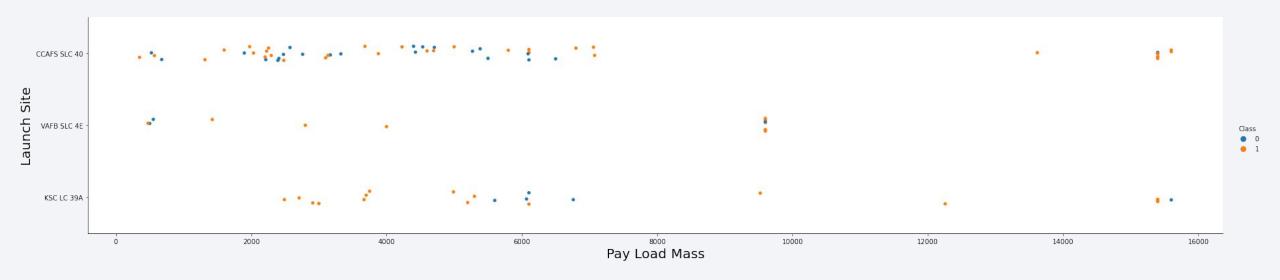
- The plot shows that as the number of flights increases, so does the likelihood of a successful landing.
- The greater the payload mass, the better the landing outcome.

# Flight Number vs. Launch Site



- The higher flight number the more success.
- CCAFS SLC 40 launch site has the lowest successful rate (60%)
- VAFB SLC 4E and KSC LC 39A has the most successful rate (77%)

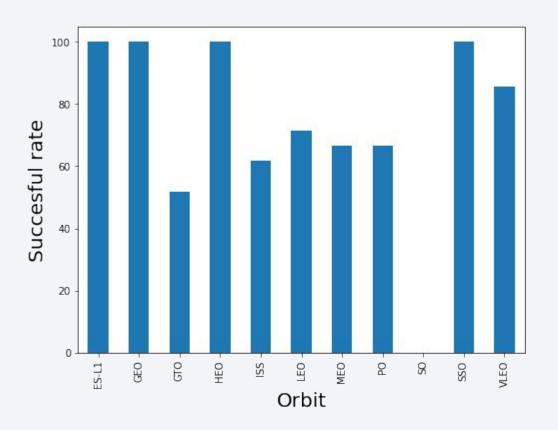
#### Payload vs. Launch Site



- The VAFB-SLC launchsite there are no rockets launched for heavy payload mass (greater than 10000)
- The CCAFS SLC 40 launch site has the most rockets launched, but mostly below 8000
- The KSC LC 39A has the most successful rate compare to other sites

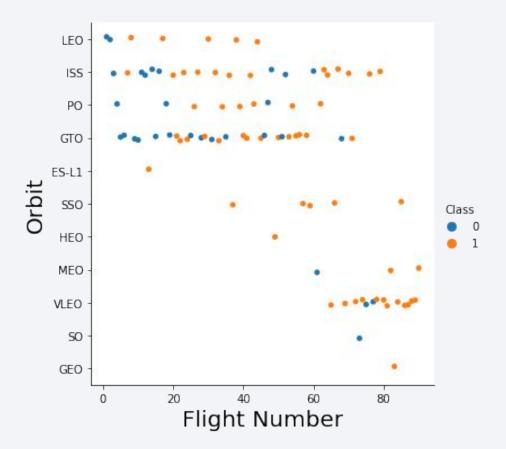
### Success Rate vs. Orbit Type

- ES-L1, GEO, HEO, SSO are the orbits that have 100% success rate.
- GTO is the orbit that has lowest success rate (about 55%)
- ISS, LEO, MEO, PO orbits have similar success rate (60%-70%)



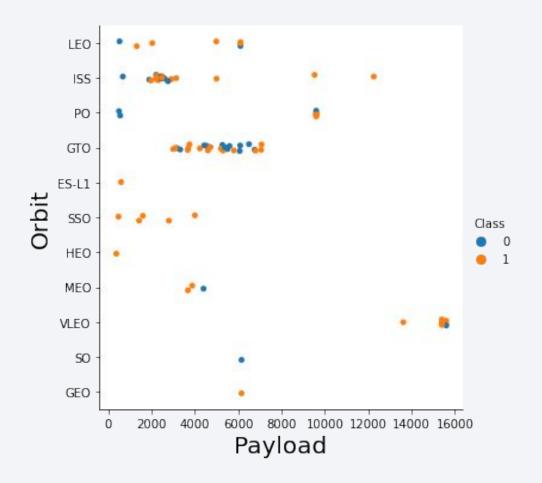
### Flight Number vs. Orbit Type

- LEO is the only orbit where the success rate is related to the amount of flights.
- In other orbits, there is no relationship between success rate and flight number.



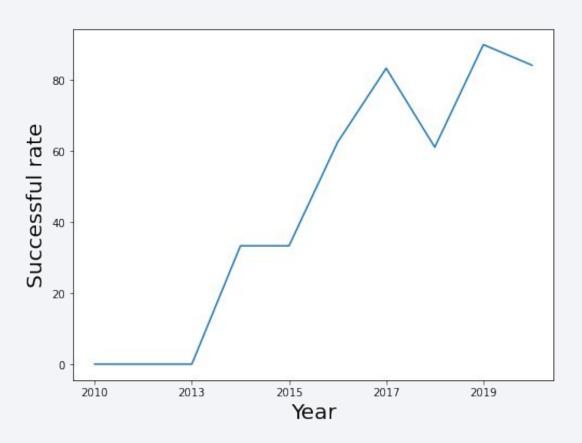
### Payload vs. Orbit Type

- We can see that LEO, ISS, and PO have a higher success rate when the payload mass is large.
- There is no relationship between success rate and payload in other orbits.



### Launch Success Yearly Trend

- The success rate kept increasing since 2013.
- The reasons could be the improvement after each failure and the advancement of technology over time.



#### All Launch Site Names

• Find the names of the unique launch sites

```
%%sql
select distinct LAUNCH_SITE
from SPACEXDATASET
```

• Use the keyword *distinct* to select the unique launch sites

### Launch Site Names Begin with 'CCA'

• Find 5 records where launch sites begin with `CCA`

```
%%sql
select distinct LAUNCH_SITE
from SPACEXDATASET
where launch_site like 'CCA%'
```

Use distinct and where to select the launch sites with the name starts with
 CCA

#### **Total Payload Mass**

Calculate the total payload carried by boosters from NASA

```
%%sql
select sum(PAYLOAD_MASS__KG_) as total_payload_mass
from SPACEXDATASET
where CUSTOMER = 'NASA (CRS)'
```

 Use sum function to summary the total payload mass and where statement to apply condition (customer is NASA)

#### Average Payload Mass by F9 v1.1

Calculate the average payload mass carried by booster version F9 v1.1

```
%%sql
select avg(PAYLOAD_MASS__KG_) as average_payload_mass
from SPACEXDATASET
where BOOSTER_VERSION = 'F9 v1.1'
```

 Use avg function to calculate the average of payload mass and where statement to apply condition (BOOSTER\_VERSION is F9 v1.1)

#### First Successful Ground Landing Date

• Find the dates of the first successful landing outcome on ground pad

```
%%sql
select min(DATE)
from SPACEXDATASET
where LANDING_OUTCOME = 'Success (ground pad)'
```

- Use *min* function to get the first date (minimum) of the first successful landing outcome on ground pad.
- Use where statement to apply condition (LANDING\_OUTCOME is Success (ground pad))

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

 List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

```
%%sql
select BOOSTER_VERSION
from SPACEXDATASET
where LANDING__OUTCOME = 'Success (drone ship)' and
(PAYLOAD_MASS__KG_ > 4000 and PAYLOAD_MASS__KG_ < 6000)
```

Use the comparison operators > and < to select the</li>
 PAYLOAD MASS KG that greater than 4000 but less than 6000

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

Calculate the total number of successful and failure mission outcomes

```
%%sql
select MISSION_OUTCOME, count(MISSION_OUTCOME) as total
from SPACEXDATASET
group by MISSION_OUTCOME
```

 Use count function to count the number of mission outcomes, and using group by statement to group the value by the mission outcomes.

### **Boosters Carried Maximum Payload**

 List the names of the booster which have carried the maximum payload mass

```
%%sql
select BOOSTER_VERSION
from SPACEXDATASET
where PAYLOAD_MASS__KG_ = (select
max(PAYLOAD_MASS__KG_) from SPACEXDATASET)
```

 Use the subquery to return the maximum payload mass(max function) and get the booster version that has the payload equal to the maximum payload mass

#### 2015 Launch Records

 List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015

```
%%sql
select LANDING__OUTCOME, BOOSTER_VERSION, LAUNCH_SITE
from SPACEXDATASET
where year(DATE) = 2015 and LANDING__OUTCOME = 'Failure (drone
ship)'
```

• Use *year* function to extract the year from DATE to select the year in 2015

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

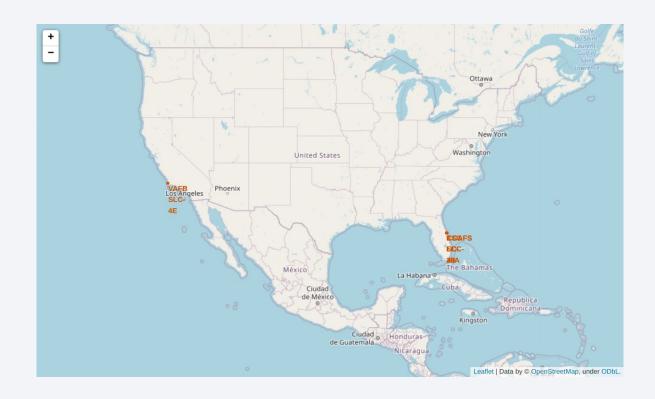
 Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order

```
%%sql select LANDING__OUTCOME, count(LANDING__OUTCOME) as total from SPACEXDATASET where DATE between '2010-06-04' and '2017-03-20' group by LANDING__OUTCOME order by total desc
```

- Use *between* operator to get the date from 2010-06-04 and 2017-03-20.
- Use group by statement to group result by landing outcome
- Use order by statement with desc to sort the result in descending total order.

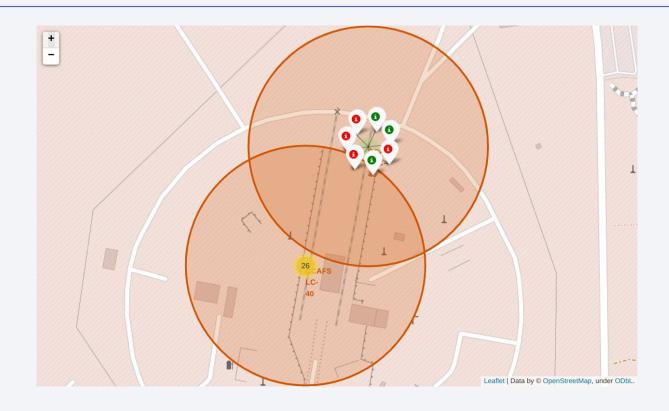


#### Launch site locations



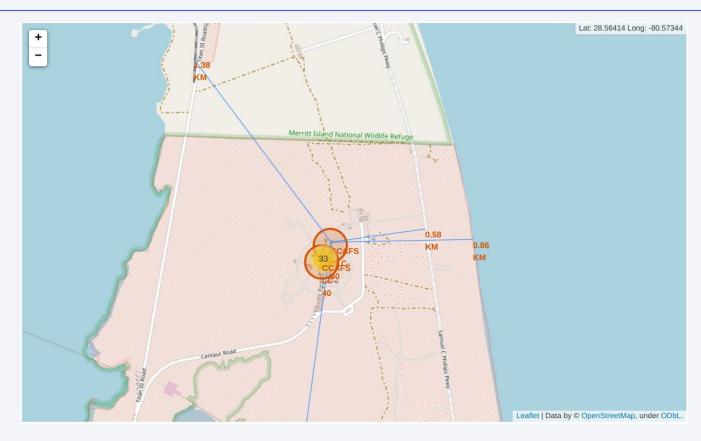
- All the launch sites in proximity to the Equator line, it's because of the rotational speed of Earth that helps the launch.
- All the launch sites is located very close to the coast to reduce the risks over populated areas.

## Landing outcome on each landing site



- Based on the color-labeled markers in marker clusters, we can see the KSC
   LC-39A has the most successful landing outcome (10 success out of 13 landings)
- CCAFS LC-40 has the lowest success rate, with 19 failures out of 26 landings.

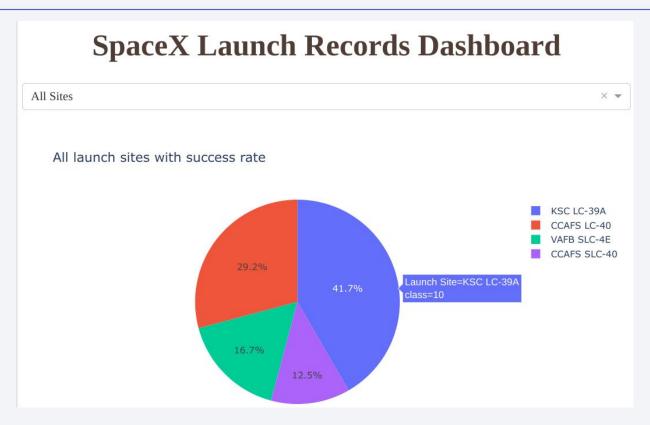
## Launch site vs its proximities



- The launch site (CCAFS SLC-40) in close proximity to railway: 1.38 Km
- The launch site (CCAFS SLC-40) in close proximity to highway: 0.58 Km
- The launch site (CCAFS SLC-40) in close proximity to coastline: 0.86 Km
- The launch site (CCAFS SLC-40) keeps 49.86 Km away from Melbourne city



#### Launch sites success rate



- The KSC LC-39A has the highest landing success rate (41.7 percent of total rate).
- CCFS SLC-40 has the lowest landing success rate (12.5%)

### Launch site - success ratio



• KSC LC-39A has the highest launch success ration (76.9% success vs 23.1% fail).

## Success rate - Payload vs Landing outcome



 Payload range 0 - 5000: FT version have the largest success rate vs V1.1 version has the most failure rate.

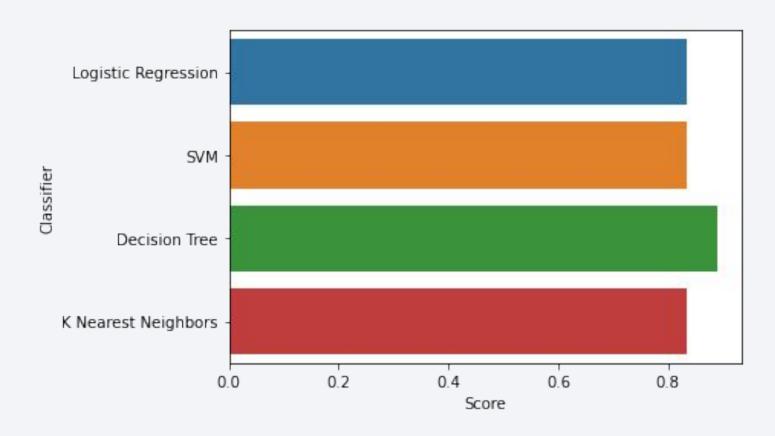
## Success rate - Payload vs Landing outcome



- Payload range 5000 9600: FT version have the most success rate vs B4 version has the most failure rate.
- In heavy payload mass, only FT and B4 boosters are used (greater than 5000 kg).



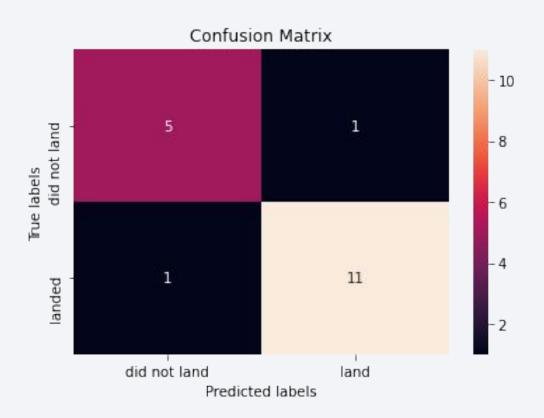
### Classification Accuracy



As the chart shows, the Decision Tree classifier has the highest score (88% accuracy)

#### **Confusion Matrix**

- The confusion matrix plot shows the accuracy of the model predictions:
  - 11 True Positives vs 1 False Negative
  - 5 True Negatives vs 1 False Positive



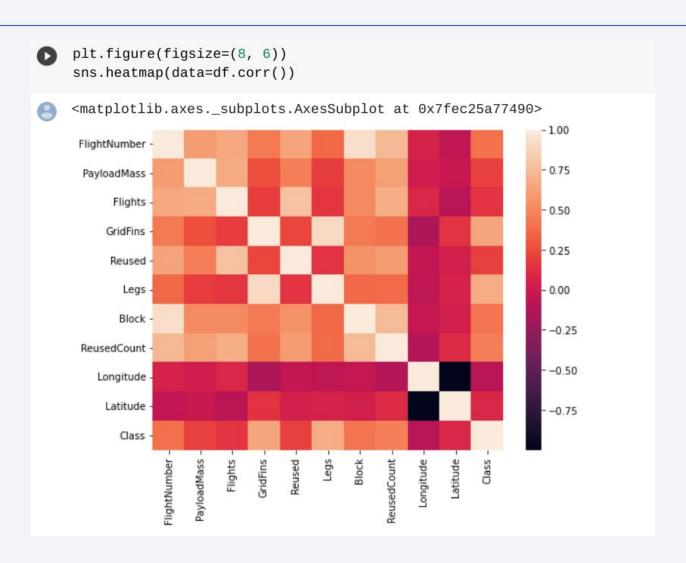
#### Conclusions

- The greater payload mass, the better landing outcome.
- KSC LC-39A has the most success rate compare to others.
- ES-L1, GEO, HEO, SSO are the orbits that have 100% success rate.
- The reasons could be the improvement after each failure and the advancement of technology over time.
- All the launch sites in proximity to the Equator line
- All the launch sites is located very close to the coast to reduce the risks over populated areas.
- The Decision Tree Classifier is the best model.

# **Appendix**

- Use seaborn heatmap plot to show the correlation in the dataset.
- Show the flight number over the year.

# **Appendix**



# **Appendix**

```
flight_by_year_df = df.groupby('Year')['FlightNumber'].sum()
    flight_by_year_df.plot(figsize=(8, 6))
    plt.xlabel("Year", fontsize=20)
    plt.ylabel("Flight Number", fontsize=20)
    Text(0, 0.5, 'Flight Number')
        1400
        1200
     Flight Number
        200
                        2013
                                    2015
                                               2017
                                                          2019
             2010
                                      Year
```

