

# Winning Space Race with Data Science

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## **Outline**

- Executive Summary
- Introduction
- Methodology
- Results
- Conclusion
- Appendix
- (https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone

## **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

#### **Background and Context**

• In this capstone, we will predict if the Falcon 9 first stage will land successfully. 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. In this module, you will be provided with an overview of the problem and the tools you need to complete the course. Space Y wants to compete with Space X

#### **Problem at Hand**

Space Y tasks us to train a machine learning model to predict successful Stage 1 recovery



# Methodology

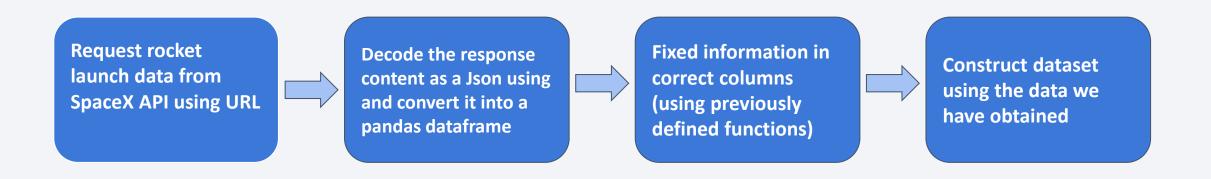
#### **Executive Summary**

- 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**

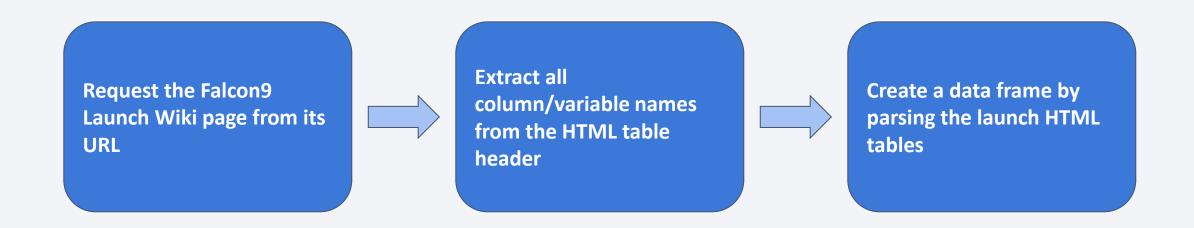
- The data was collected using various methods
- Data collection was done using get request to the SpaceX API.
- Next, we decoded the response content as a Json using .json() function call and turn it into a pandas dataframe using .json\_normalize().
- We then cleaned the data, checked for missing values and fill in missing values where necessary.
- In addition, we performed web scraping from Wikipedia for Falcon 9 launch records with BeautifulSoup.
- The objective was to extract the launch records as HTML table, parse the table and convert it to a pandas dataframe for future analysis.

## Data Collection – SpaceX API



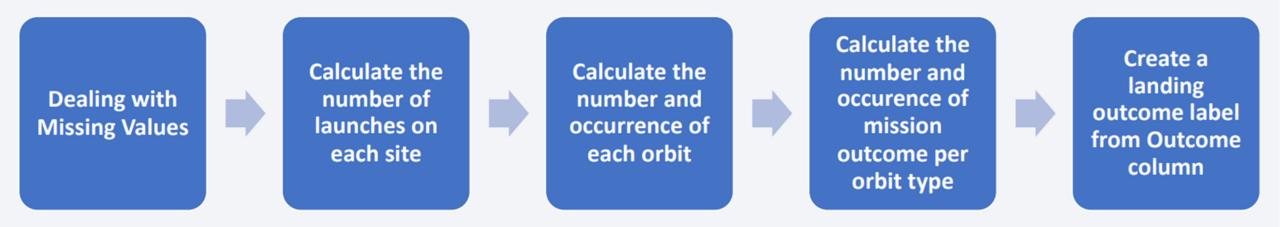
https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone/blob/main/jupyter-labs-spacex-data-collection-api.ipynb

## **Data Collection - Scraping**



https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone/blob/main/jupyter-labs-webscraping.ipynb

## **Data Wrangling**



https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb

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

- The graphs used are
  - Scatter Plot
  - Bar chart
  - Line chart

because they are the ones that best highlight the relationships between the variables considered

• GITHUB URL: https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone/blob/main/edadataviz.ipynb

## **EDA** with SQL

#### SQL queries performed

- Display the names of the unique launch sites in the space mission
- Display 5 records where launch sites begin with the string 'CCA'
- Display the total payload mass carried by boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1
- List the date when the first successful landing outcome in the ground pad was achieved.
- List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000.
- List of total number of successful and failure mission outcomes
- List the names of the booster\_versions which have carried the maximum payload mass.
- List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- GITHUB URL: https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-sql-coursera\_sqllite.ipynb

## Build an Interactive Map with Folium

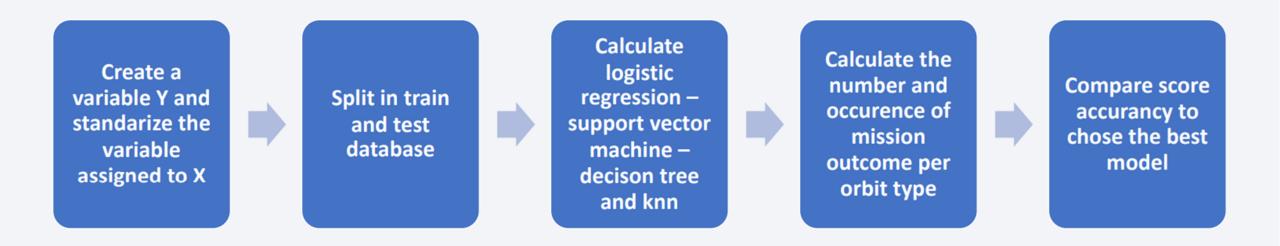
- Summary of map objects:
- Markers: Show a geo location from latitude and longitude data
- Cluster: Show a group of markers
- Circles: Show a single location
- Lines: Show distance between two

• GITHUB URL: https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone/blob/main/lab\_jupyter\_launch\_site\_location.ipynb

## Build a Dashboard with Plotly Dash

- Summary of plots:
- Bar: Show categories differences
- Line: Reports time series changes
- Pie: Shows the percentage of events
- Tree: Shows complex relationship of variables in interactive way
- Map: Shows variables of states on a map

## Predictive Analysis (Classification)



GITHUB URL: https://github.com/imadomar64/IBM-Applied-Data-Science-Capstone/blob/main/SpaceX\_Machine%20Learning%20Prediction.ipynb

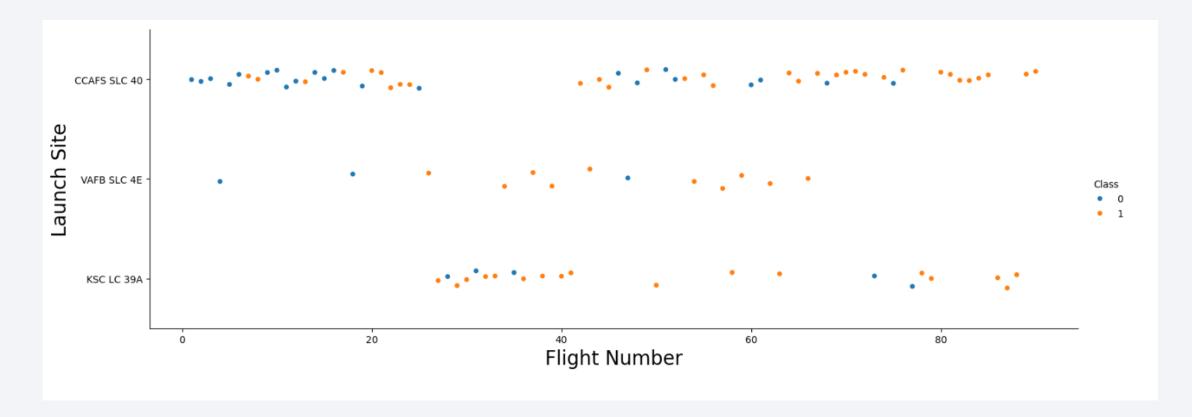
### Results

- Exploratory data analysis results
  - Both API and web scraping are capable to collect Xspace data
- Interactive analytics demo in screenshots
  - EDA with SQL is effective for data filtering
  - EDA with interactive visualization provides informative information
  - Plotly Dash is powerful to show instant data change
- Predictive analysis results
  - Decision Tree Classifier Algorithm has the best accuracy of predicting.



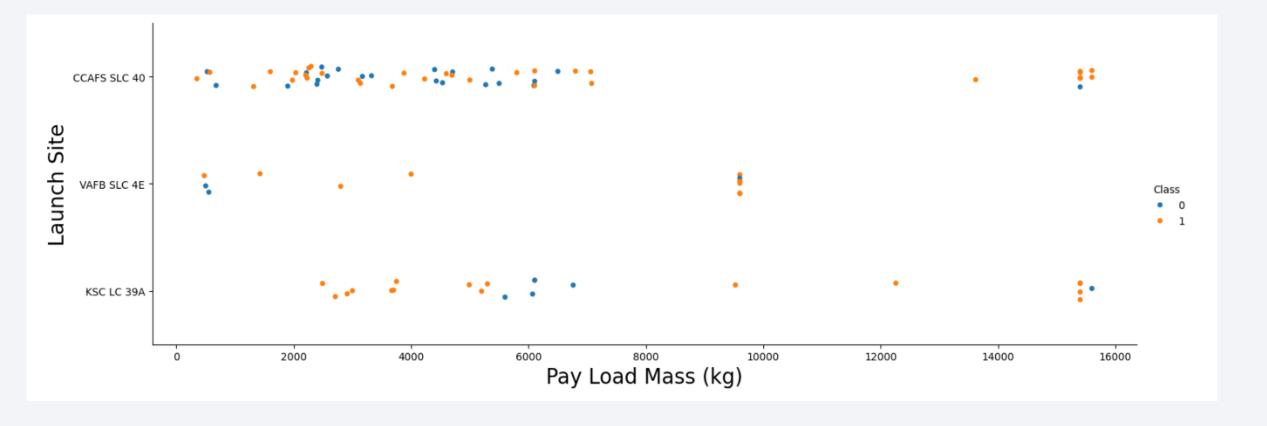
## Flight Number vs. Launch Site

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



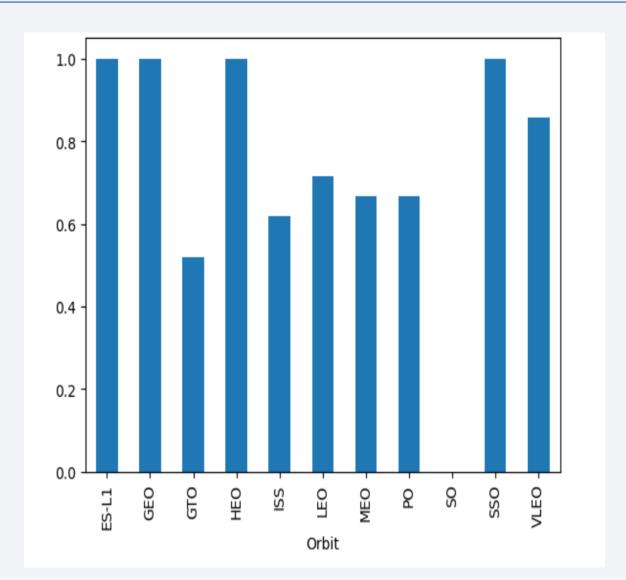
## Payload vs. Launch Site

Blue indicates successful launch and orange indicates unsuccessful launch. Payload mass appears to fall mostly between 0-6000 kg. Different launch sites also seem to use different payload mass.



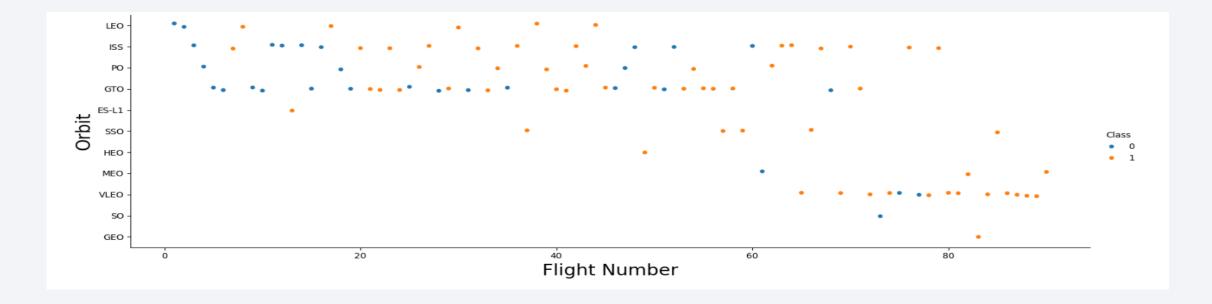
## Success Rate vs. Orbit Type

- ES-L1 (1), GEO (1), HEO (1)
   have 100% success rate
   (sample sizes in parenthesis)
   SSO (5) has 100% success rate
- VLEO (14) has decent success rate and attempts
- SO (1) has 0% success rate
- GTO (27) has the around 50% success rate but largest sample



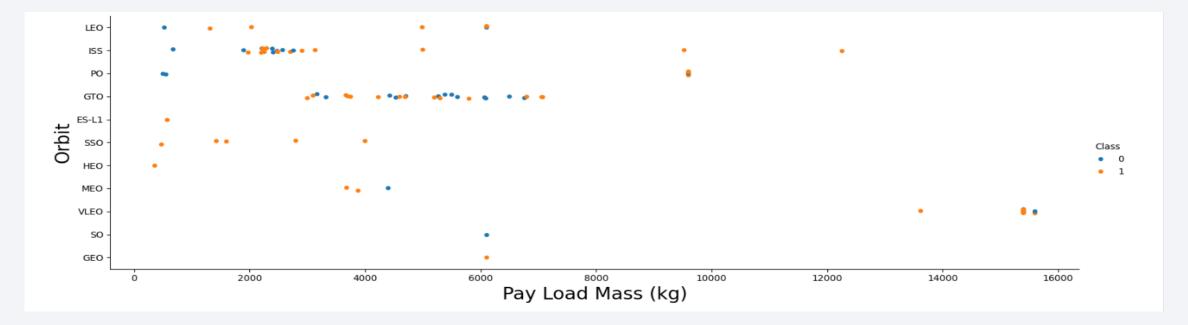
## Flight Number vs. Orbit Type

- Blue indicates successful launch and orange indicates unsuccessful launch.
- Launch Orbit preferences changed over Flight Number.
- Launch Outcome seems to correlate with this preference.
- SpaceX started with LEO orbits which saw moderate success LEO and returned to VLEO in recent launches
- SpaceX appears to perform better in lower orbits or Sun-synchronous orbits



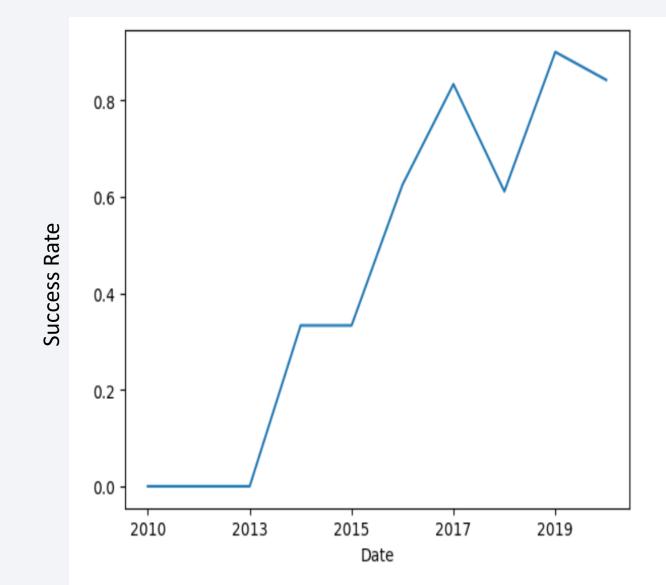
## Payload vs. Orbit Type

- Blue indicates successful launch and orange indicates unsuccessful launch.
- Payload mass seems to correlate with orbit
- LEO and SSO seem to have relatively low payload mass
- The other most successful orbit VLEO only has payload mass values in the higher end of the range



## Launch Success Yearly Trend

- Success generally increases over time since
   2013 with a slight dip in
   2018
- Success in recent years at around 80%



#### All Launch Site Names

- CCAFS LC-40
- CCAFS SCL-40
- KSC LC-39A
- VAFB SLC-4E

```
%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
```

# 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_Outcome
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 v1.0 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- 40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
2012- 10-08	0:35:00	F9 v1.0 B0006	CCAFS LC- 40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
2013-03-01	15:10:00	F9 v1.0 B0007	CCAFS LC- 40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

# **Total Payload Mass**

```
%sql select sum(payload_mass__kg_) as total_payload_mass from SPACEXDATASET where customer = 'NASA (CRS)';

* sqlite:///my_data1.db

hone.

total_payload_mass

45596
```

## Average Payload Mass by F9 v1.1

# First Successful Ground Landing Date

```
%sql select min(date) as first_successful_landing from SPACEXDATASET where Landing_Outcome = 'Success (ground pad)';

* sqlite://my_data1.db
Done.

first_successful_landing

2015-12-22
```

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

 %sql select booster\_version from SPACEXDATASET where Landing\_Outcome = 'Success (drone ship)' and payload\_mass\_\_kg\_ between 4000 and 6000;

#### Booster\_Version

F9 FT B1022

F9 FT B1026

F9 FT B1021.2

F9 FT B1031.2

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

```
%sql select mission_outcome, count(*) as total_number from SPACEXDATASET group by mission_outcome;
 * sqlite:///my_data1.db
Done.
             Mission_Outcome total_number
               Failure (in flight)
                      Success
                                         98
                      Success
 Success (payload status unclear)
```

# **Boosters Carried Maximum Payload**

```
%sql select booster_version from SPACEXDATASET where payload_mass__kg_ = (select max(payload_mass__kg_) from SPACEXDATASET);
  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

**%sql** SELECT date, substr(Date, 6, 2) AS Month, BOOSTER\_VERSION, LAUNCH\_SITE FROM SPACEXTBL WHERE LANDING\_OUTCOME = 'Failure (drone ship)' AND substr(Date,0,5)='2015';

Date	Month	Booster_Version	Launch_Site
2015-01-10	01	F9 v1.1 B1012	CCAFS LC-40
2015-04-14	04	F9 v1.1 B1015	CCAFS LC-40

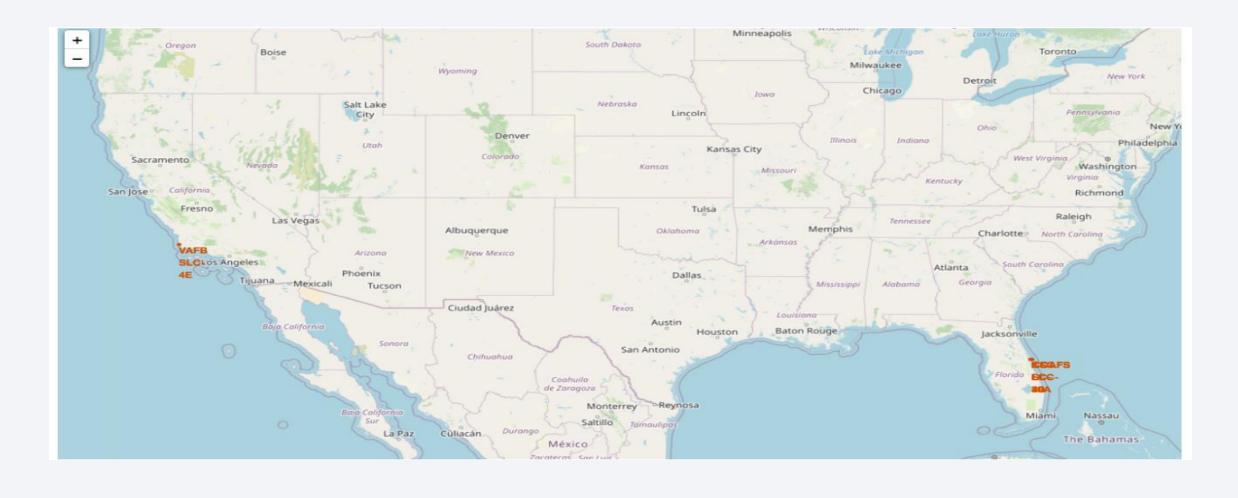
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

```
%%sql select Landing_Outcome, count(*) as count_outcomes from SPACEXDATASET
        where date between '2010-06-04' and '2017-03-20'
        group by Landing Outcome
        order by count_outcomes desc;
 * sqlite:///my_data1.db
Done.
    Landing Outcome count outcomes
           No attempt
                                    10
   Success (drone ship)
    Failure (drone ship)
  Success (ground pad)
     Controlled (ocean)
   Uncontrolled (ocean)
     Failure (parachute)
 Precluded (drone ship)
```

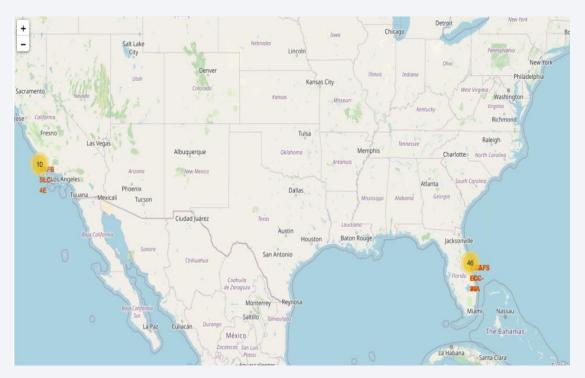


## All Launch Sites' Location Markers

All the launches are near USA, Florida, and California



## Color-labeled Launch Outcomes





Green means successful Red means Failure

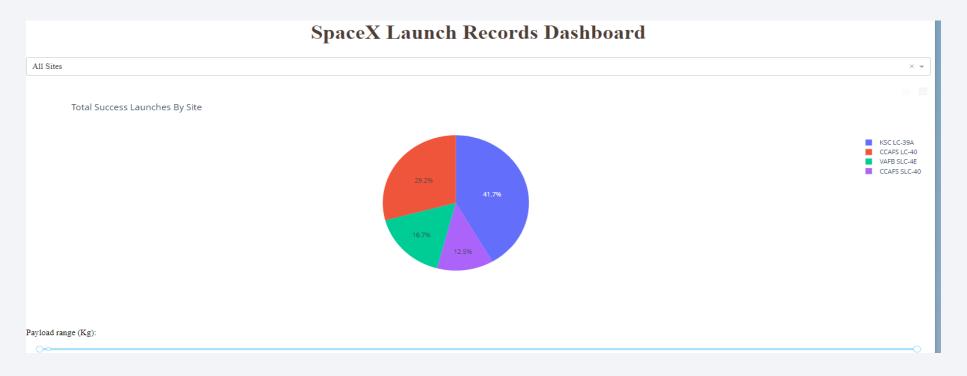
## Launch Sites to its Proximities



All distances from launch sites to its proximities, they weren't far from railway tracks.



### Launch Success Count

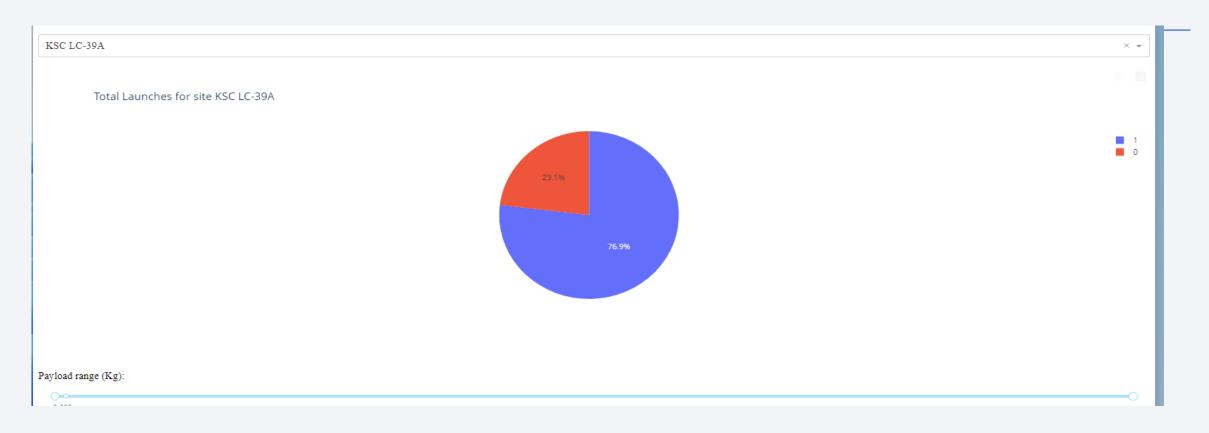


KSC LC-39A has the highest success score with 41.7%

CCAFS LC-40 comes next with 29.2%

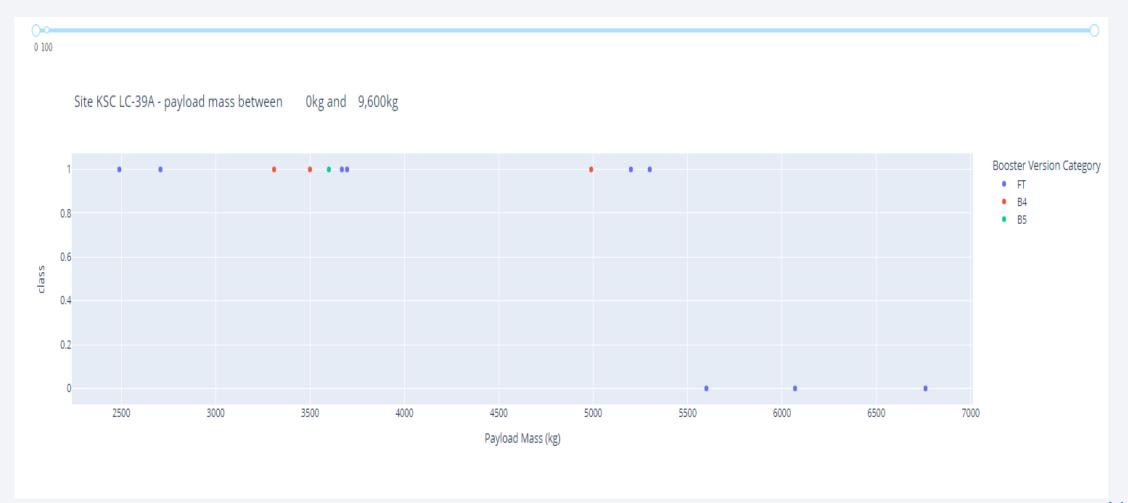
Finally, VAFB SLC-4E and CCAFS SLC-40 with 16.7% and 12.5% respectively

## Launch Site with Highest Score



KSC LC-39A has the highest score with 76.9% with payload range of 2000 kg – 10000 kg, and FT booster version has the highest score

# Payload vs. Launch Outcome





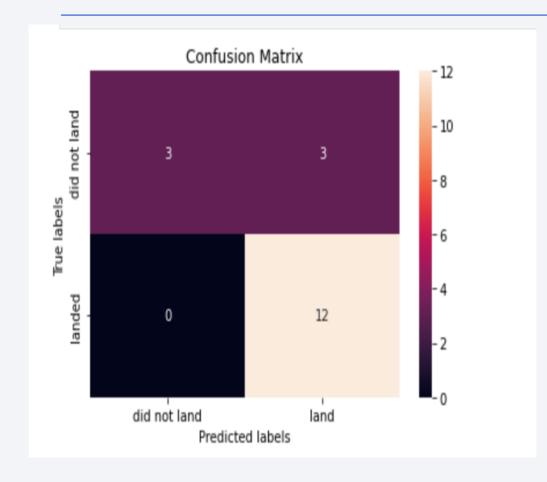
## Classification Accuracy

All models had virtually the same accuracy on the test set at 83.33% accuracy. It should be noted that test size is small at only sample size of 18.

This can cause large variance in accuracy results, such as those in Decision Tree Classifier model in repeated runs.

We likely need more data to determine the best model.

#### **Confusion Matrix**



- Since all models performed the same for the test set, the confusion matrix is the same across all models. The models predicted 12 successful landings when the true label was successful landing.
- The models predicted 3 unsuccessful landings when the true label was unsuccessful landing.
- The models predicted 3 successful landings when the true label was unsuccessful landings (false positives). Our models over predict successful landings.

#### Conclusions

#### We can conclude that:

- 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.

## **Appendix**

• Include any relevant assets like Python code snippets, SQL queries, charts, Notebook outputs, or data sets that you may have created during this project

