

# Outline

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



# **Executive Summary**

## **SUMMARY OF METHODOLOGIES**

- Data collection via API and Web Scraping
- Exploratory Data Analysis with Data Visualization
- Exploratory Data Analysis with SQL
- Building an interactive map with Folium
- Building a Dashboard with Plotly Dash
- Predictive analysis with classification models

# **SUMMARY OF ALL RESULTS**

- Exploratory Data Analysis results
- Interactive analytics via dashboards
- Predictive analysis results

# Introduction

### PROJECT BACKGROUND AND CONTEXT

SpaceX is the most successful company of the commercial space age, making space travel affordable. The company 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. Based on public information and machine learning models, we are going to predict if SpaceX will reuse the first stage.

### PROBLEMS WE WANT TO FIND ANSWERS

- What are the main characteristics of a successful or failed landing?
- How do variables such as payload mass, launch site, number of flights, and orbits affect the success of the first stage landing?
- What are the conditions which will allow SpaceX to achieve the best landing success rate?



# Methodology

# **Executive Summary**

# Data collection methodology:

- SpaceX REST API
- Web Scrapping from Wikipedia

# Perform data wrangling:

- Dropping unnecessary columns
- One Hot Encoding for classification models

Perform exploratory data analysis (EDA) using visualization and SQL

Perform interactive visual analytics using Folium and Plotly Dash

# Perform predictive analysis using classification models

• Building, tuning and evaluation of classification models to ensure the best results

# **Data Collection**

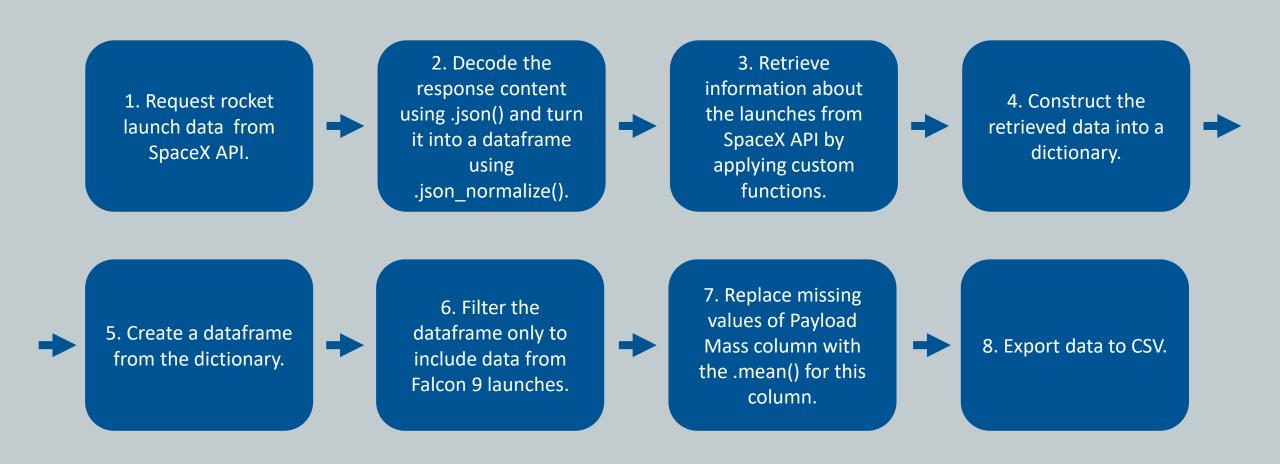
- Data collection process involved a combination of API requests from **SpaceX REST API** and **Web Scraping** data from a table in SpaceX's Wikipedia entry.
- Data obtained from SpaceX REST API:

Flight Number	LaunchSite	Legs	Longitude
Date	Outcome	Landing Pad	Latitude
Booster Version	Flights	Block	
Payload Mass	Grid Fins	Reused Count	
Orbit	Reused	Serial	

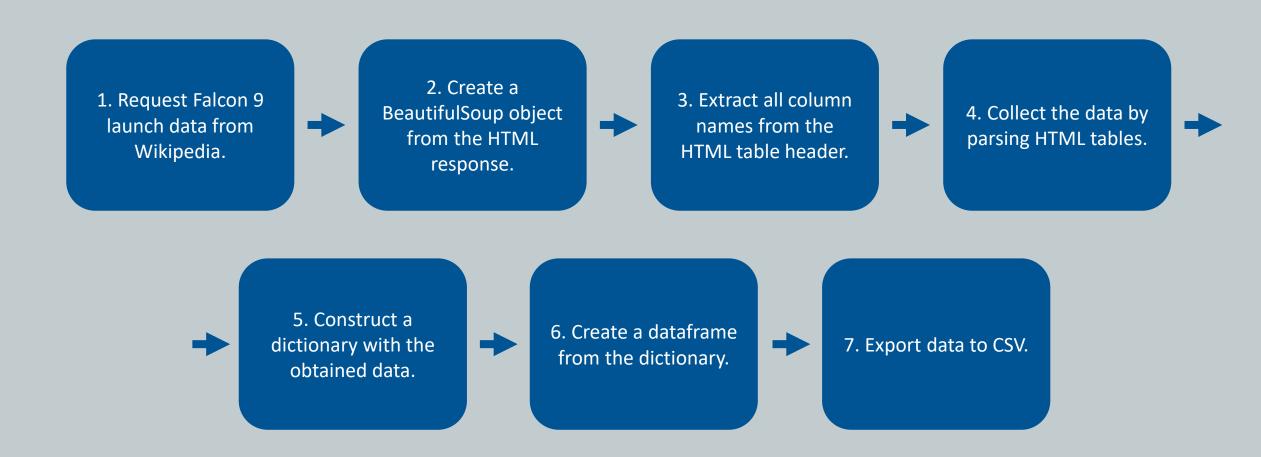
Data Columns are obtained by using Wikipedia Web Scraping:

Flight Number	Payload Mass	Launch Outcome	Date
Launch Site	Orbit	Version Booster	Time
Payload	Customer	Booster Landing	

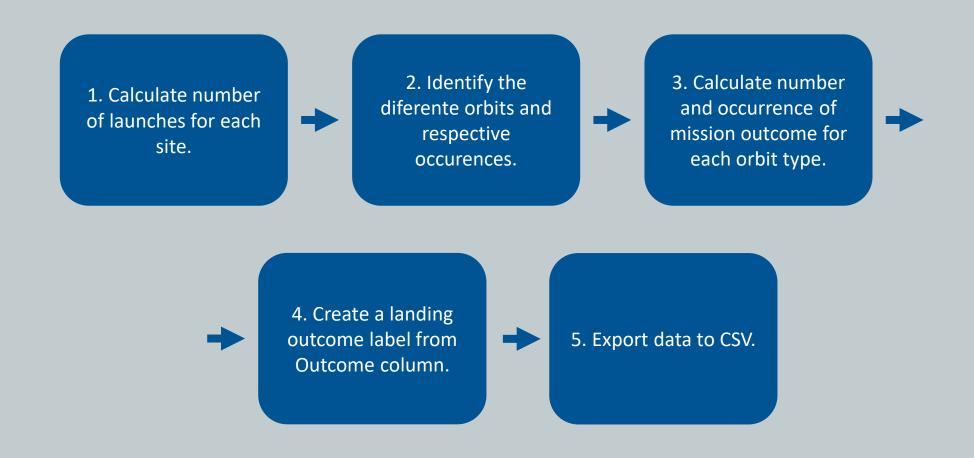
# Data Collection - SpaceX API



# **Data Collection - Scraping**



# Data Wrangling



# **EDA** with SQL

# SQL queries performed in this Lab:

- 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 ground pad was achieved.
- List the names of 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.
- List the records which will display the month names, failure landing outcomes in drone ship, booster versions, launch site for the months in year 2015.
- Rank the count of successful landing outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

# **EDA** with Data Visualization

# The following charts were plotted:

Flight Number vs. Payload Mass	Flight Number vs. Orbit Type	
Flight Number vs. Launch Site	Payload Mass vs Orbit Type	
Payload Mass vs. Launch Site	Success Rate Yearly Trend	
Orbit Type vs. Success Rate		

<u>Scatter plots</u> show the relationship between variables. If a relationship exists, they could be used in machine learning model. <u>Bar charts</u> show comparisons among discrete categories. The goal is to show the relationship between the specific categories being compared and a measured value. <u>Line charts</u> show trends in data over time (time series).

# Build an Interactive Map with Folium

# The following objects were added to the map, centered on NASA Johnson Space Center at Houson, Texas:

- Red circle on NASA Johnson Space Center's coordinates, with label showing its name (folium.Circle, folium.map.Marker).
- Red circles on each launch site coordinates, with label showing launch site name (folium.Circle, folium.map.Marker, folium.features.Divlcon).
- The grouping of points in a cluster to display different informations for the same coordinates (folium.plugins.MarkerCluster).
- Markers to show successful and unsuccessful landings. Green for successful landing and Red for unsuccessful landing (folium.map.Marker, folium.lcon).
- Markers to show distance between launch site to key locations (railway, highway, coastway, city) and a plot line between them (folium.map.Marker, folium.PolyLine, folium.features.DivIcon).

These objects were selected to add context to the data gathered. They show launch sites, points of interest in their surroundings, and the number of successful and unsuccessful landings.

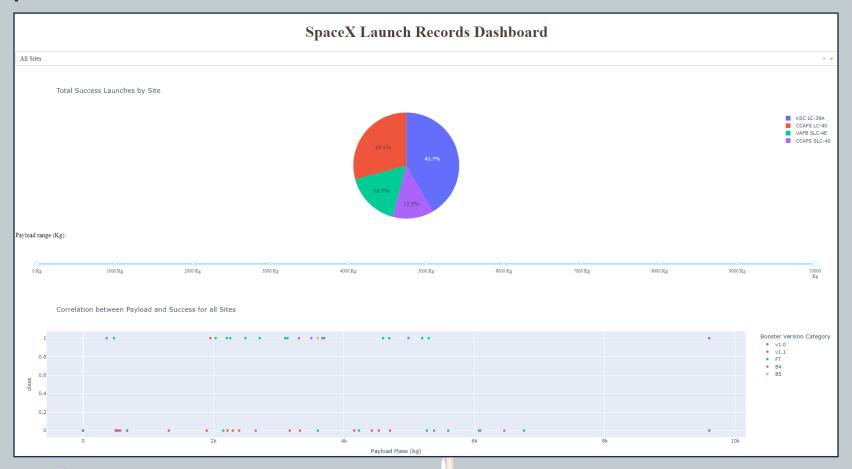
# Build a Dashboard with Plotly Dash

# The dashboard has the following components:

Dropdown menu	Allow the user to choose between the different launch sites or view the data for all of them.
Pie Chart	Shows the total success and the total failure for the launch site chosen on the dropdown menu.
Rangeslider	Allows a user to select a payload mass in a fixed range.
Scatterplot	Shows the relationship between Success vs. Payload Mass.

# Build a Dashboard with Plotly Dash

# Dashboard printscreen:





# Predictive Analysis (Classification)

3. Split the data into 2. Standardize the 1. Create a NumPy 4. Create a data with training and testing array from the GridSearchCV object sets with the StandardScaler, then column "Class" in with cv = 10 to find fitting it and train\_test\_split data. the best parameters. transforming it. function. 6. Calculate the 8. Find the method 5. Apply GridSearchCV on 7. Examine the that performs the accuracy on the test best by examining the data using the confusion matrix for LogReg, SVM, Decision Tree, and method .score() for all models. Jaccard\_score and KNN models. all models. F1 score metrics.

# Results

### **EXPLORATORY DATA ANALYSIS RESULTS:**

- Lighter payloads perform better when compared to heavier ones.
- The lunch success is increasing with the number of years of experience, demonstrating a positive trend over time.
- The Launch Complex 39A at Kennedy Space Center has the highest number of successful launches of the 4 sites analyzed.
- GEO, HEO, SSO and ES L1 are the orbit types with the highest rate of successful launches.

### PREDICTIVE ANALYSIS RESULTS:

```
Accuracy of GridSeachCV: 0.8625
Accuracy of SVM score: 0.7777
Accuracy of Decision Tree: 0.90357
Accuracy of KNN model: 0.8767

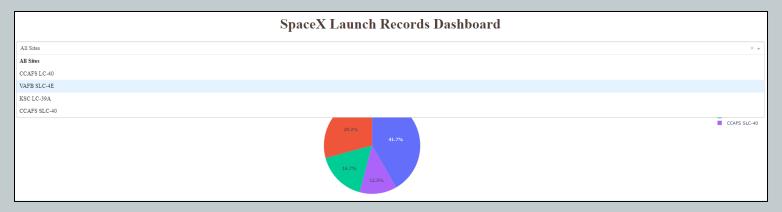
Best Parameters are: Decision Tree with a score of 0.90357

Best Parameters are: {'criterion': 'entropy', 'max_depth': 16, 'max_features': 'sqrt', 'min_samples_leaf': 2, 'min_samples_split': 2, 'splitter': 'random'}
```

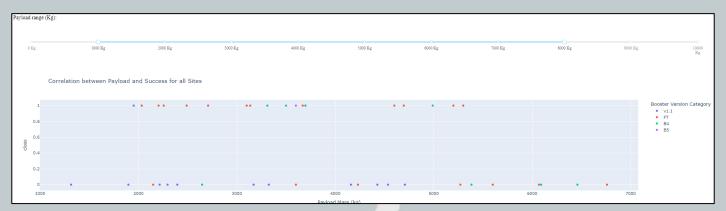
# Results

### **INTERACTIVE ANALYTICS DEMO IN SCREENSHOTS**

Dropdown menu to select launch sites in order to show in the pie chart the total success launches:

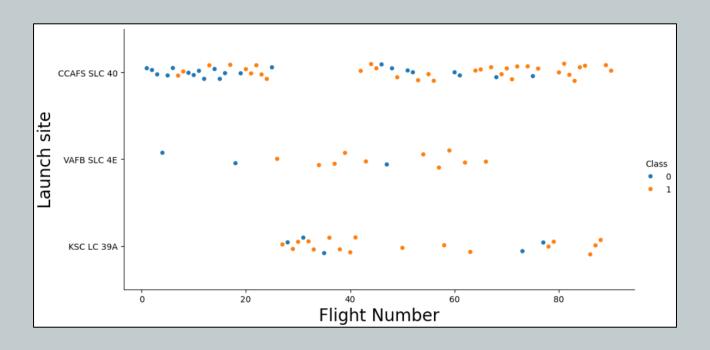


Scatterplot representing correlation with payload and success for all sites, with range slider to select payload:



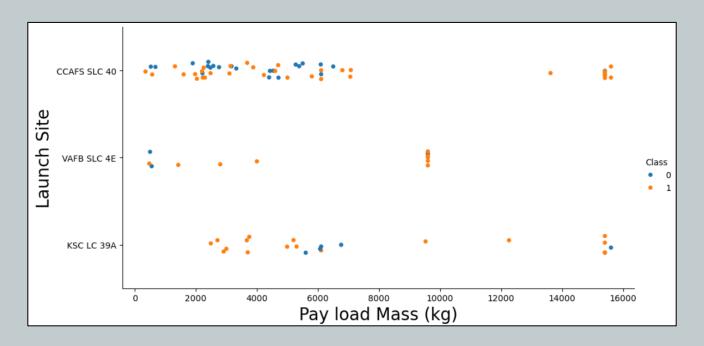


# Flight Number vs. Launch Site



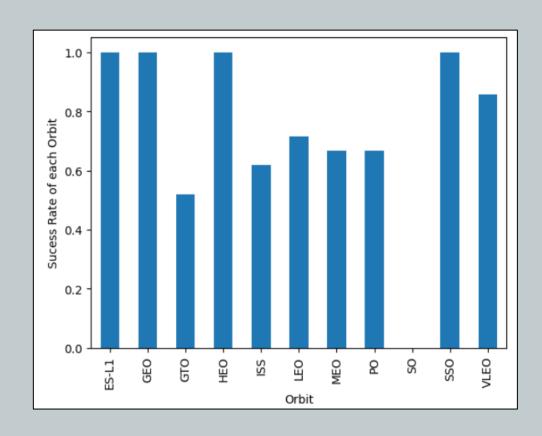
- The earliest flights all failed while the latest flights all succeeded.
- The CCAFS SLC 40 launch site has about a half of all launches.
- VAFB SLC 4E and KSC LC 39A have higher success rates.
- It can be assumed that each new launch has a higher rate of success.

# Payload vs. Launch Site



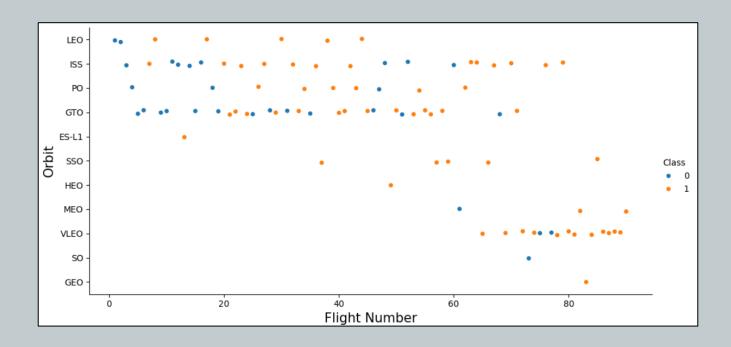
- For every launch site, the higher the payload mass, the higher the success rate.
- Most of the launches with payload mass over 7000 kg were successful.
- KSC LC 39A has a 100% success rate for payload mass under 5500 kg.

# Success Rate vs. Orbit Type



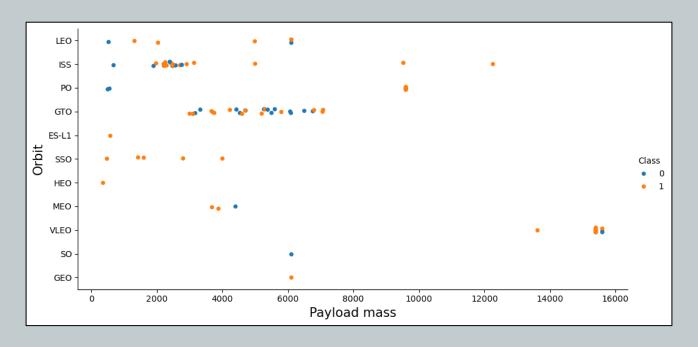
- ES-L1, GEO, HEO and SSO have 100% success rate.
- SO as a 0% success rate.
- GTO, ISS, LEO, MEO and PO have a success rate between 50% an 85%.

# Flight Number vs. Orbit Type



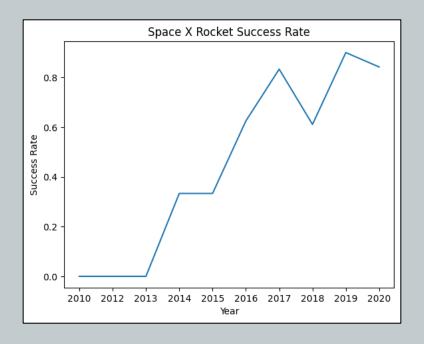
- The success rate increases with the number of flights for the LEO orbit.
- For some orbits, like GTO, there is no relation between the success rate and the number of flights.
- High success rates from orbits SSO or HEO may be due to the knowledge acquired from previous launches for other orbits.

# Payload vs. Orbit Type



- Heavy payloads show a negative influence on GTO orbits and a positive influence on GTO and Polar LEO (ISS) orbits.
- Lower payloads for a GTO orbit improves the success of a launch.

# Launch Success Yearly Trend



### **OBSERVATIONS:**

• After the first 3 years of unsuccessful launches, the success rate has been increasing up to 2020.

# All Launch Site Names

# **QUERY:**

# Task 1 Display the names of the unique launch sites in the space mission [13]: %sql select distinct launch\_site from SPACEXTABLE;

### **RESULT:**



# **EXPLANATION:**

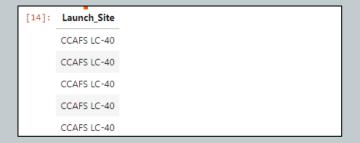
Use of DISTINCT in the query allows to remove duplicate LAUNCH\_SITE entries.

# Launch Site Names Begin with 'CCA'

# **QUERY:**



### **RESULT:**



### **EXPLANATION:**

- The WHERE clause followed by LIKE clause filters launch sites that contain the substring CCA.
- LIMIT 5 shows 5 records from filtering.

# **Total Payload Mass**

# **QUERY:**

# Task 3 Display the total payload mass carried by boosters launched by NASA (CRS) \*\*Sal SELECT SUM(PAYLOAD\_MASS\_\_KG\_) FROM SPACEXTBL WHERE Customer = 'NASA (CRS)';

### **RESULT:**

```
SUM(PAYLOAD_MASS__KG_)
45596
```

### **EXPLANATION:**

• The query sums all the payload masses where the customer is NASA (CRS).

# Average Payload Mass by F9 v1.1

## **QUERY:**

# Task 4 Display average payload mass carried by booster version F9 v1.1 \*\*\*sql SELECT AVG(PAYLOAD MASS KG ) FROM SPACEXTBL WHERE Booster Version LIKE 'F9 v1.0%';

### **RESULT:**

AVG(PAYLOAD\_MASS\_KG\_)

340.4

# **EXPLANATION:**

• The query returns the average of all payload masses where the booster version contains the substring F9 v1.1.

# First Successful Ground Landing Date

## **QUERY:**

# Task 5 List the date when the first succesful landing outcome in ground pad was acheived. Hint:Use min function \*\*\*Sql SELECT MIN(Date) FROM SPACEXTBL WHERE Landing Outcome = 'Success (ground pad)';

### **RESULT:**

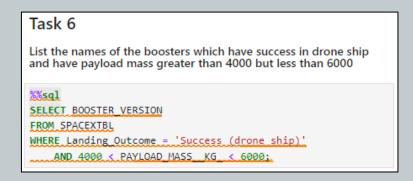
MIN(Date) 2015-12-22

### **EXPLANATION:**

• The WHERE clause filters the dataset in order to keep only records where landing was successful. Then, with the MIN function, the record with the earliest date is selected.

# Successful Drone Ship Landing with Payload between 4000 and 6000

# **QUERY:**



### **RESULT:**

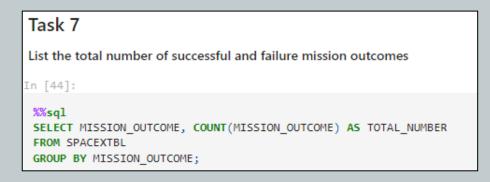
Booster_Version	
F9 FT B1021.1	F9 FT B1036.1
F9 FT B1022	F9 FT B1038.1
F9 FT B1023.1	F9 B4 B1041.1
F9 FT B1026	F9 FT B1031.2
F9 FT B1029.1	F9 B4 B1042.1
F9 FT B1021.2	F9 B4 B1045.1
F9 FT B1029.2	F9 B5 B1046.1

### **EXPLANATION:**

 The query returns the booster versions where landing was successful, and payload mass is between 4000 and 6000 kg. The WHERE and AND clauses filter the dataset.

# Total Number of Successful and Failure Mission Outcomes

## **QUERY:**



### **RESULT:**

mission_outcome	total_number
Failure (in flight)	1
Success	99
Success (payload status unclear)	1

### **EXPLANATION:**

• The different values of the Mission Outcome column were counted and presented on a table as total number of occurrences.

# **Boosters Carried Maximum Payload**

## **QUERY:**

# 

### **RESULT:**

Booster_Version	
F9 B5 B1048.4	F9 B5 B1049.5
F9 B5 B1049.4	F9 B5 B1060.2
F9 B5 B1051.3	F9 B5 B1058.3
F9 B5 B1056.4	F9 B5 B1051.6
F9 B5 B1048.5	F9 B5 B1060.3
F9 B5 B1051.4	F9 B5 B1049.7

### **EXPLANATION:**

• A subquery was used to filter data by returning only the heaviest payload mass with MAX function. The main query uses subquery results and returns unique booster version with the heaviest payload mass.

# 2015 Launch Records

# **QUERY:**

# Task 9 List the records which will display the month names, failure landing\_outcomes in drone ship ,booster versions, launch\_site for the months in year 2015. Note: SQLLite does not support monthnames. So you need to use substr(Date, 6,2) as month to get the months and substr(Date,0,5)='2015' for year. \*\*SQL SELECT\_Landing\_Outcome\_, BOOSTER\_VERSION\_, LAUNCH\_SITE FROM\_SPACEXTBL WHERE\_Landing\_Outcome\_= 'Failure\_(drone\_ship)' AND\_year(DATE) = 2015;

### **RESULT:**

landing_outcome	booster_version	launch_site
Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40

### **EXPLANATION:**

• The query returns landing outcome, booster version and launch site where landing was unsuccessful, and landing date was the year 2015.

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

# **QUERY:**

```
SELECT Landing Outcome, COUNT(Landing Outcome) AS TOTAL_NUMBER FROM SPACEXTBL
WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20'
GROUP BY Landing_Outcome
ORDER BY TOTAL_NUMBER DESC
```

### **RESULT:**

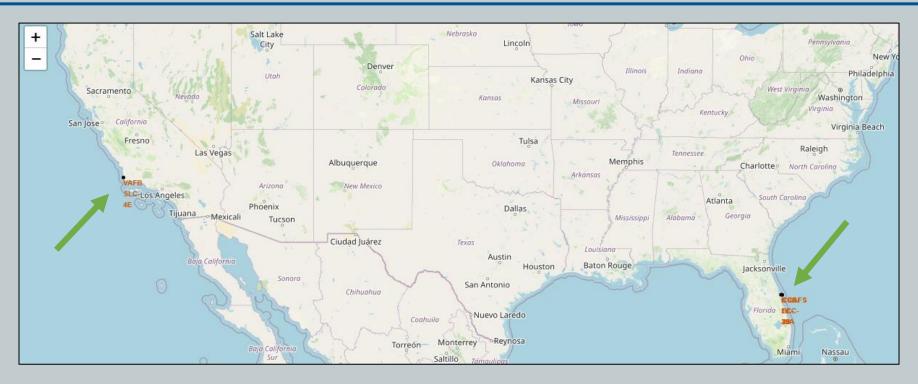
Landing_Outcome	TOTAL_NUMBER
No attempt	10
Success (drone ship)	5
Failure (drone ship)	5
Success (ground pad)	3
Controlled (ocean)	3
Uncontrolled (ocean)	2
Failure (parachute)	2
Precluded (drone ship)	1

### **EXPLANATION:**

• The query counts all the landing outcomes instances thar occurred between 2010-06-04 and 2017-03-20 and sets up the results in descending order.

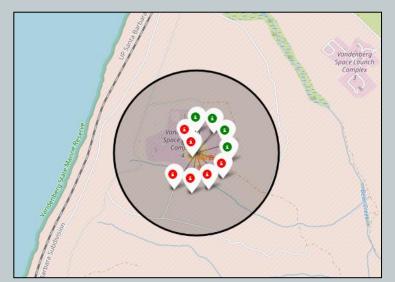


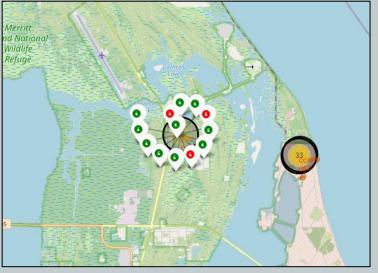
### **Ground Stations**

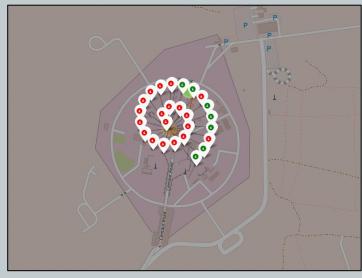


- The map shows the launch locations on the US map, both on the east and west coast.
- The ocean proximity reduces risk to human populations in case of launch failures and also provides a safer trajectory for the rocket stages that fall back to Earth after separation.

## Color-labeled launch

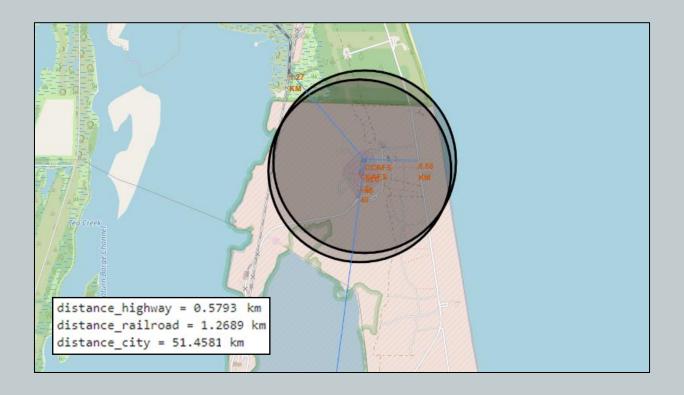






- From the color-labeled markers it's easy to identify which launch sites have relatively high success rates.
- Launch Site KSC LC-39A has a very high success rate.

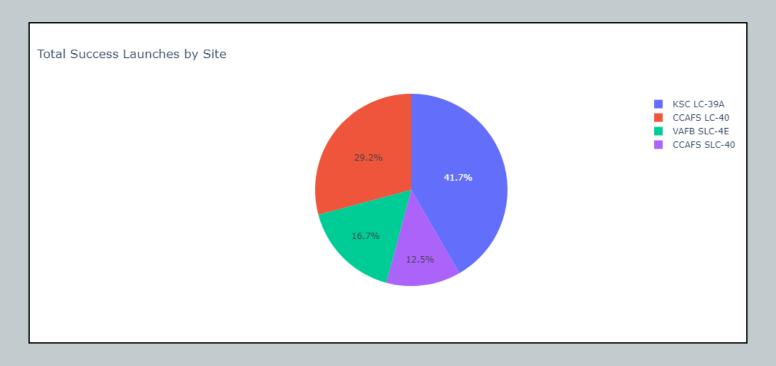
### Distances from CCAFS-SLC4



- The map shows CCAFS-SLC4 benefits from being in the vicinity of the coastline, a higway and a railroad.
- The nearest city is located about 51km away.

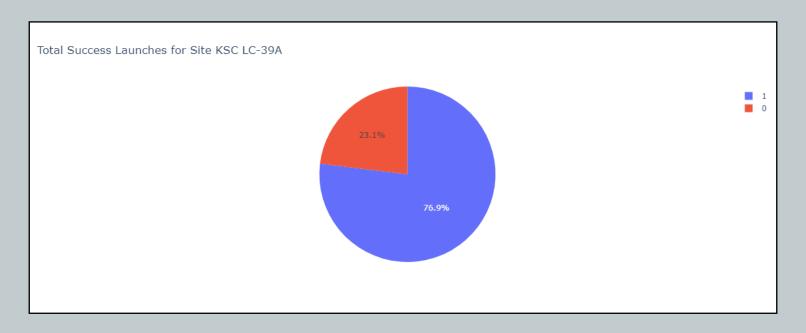


## Dashboard – Total success by Site



- The pie chart shows the total successful launches count for all sites.
- The KSC LC-39A launch site has the best success rate of launches with 41,7%.

### Dashboard – Total success launches for KSC LC-39A

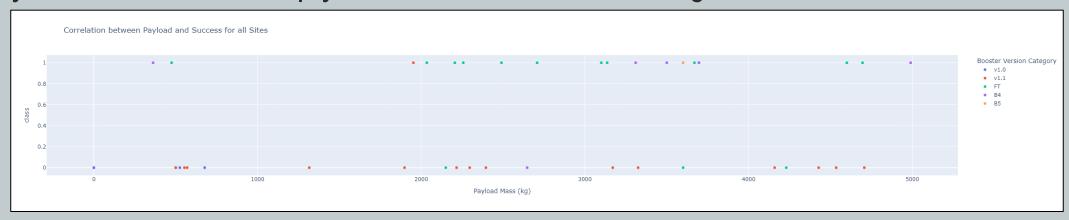


### **OBSERVATIONS:**

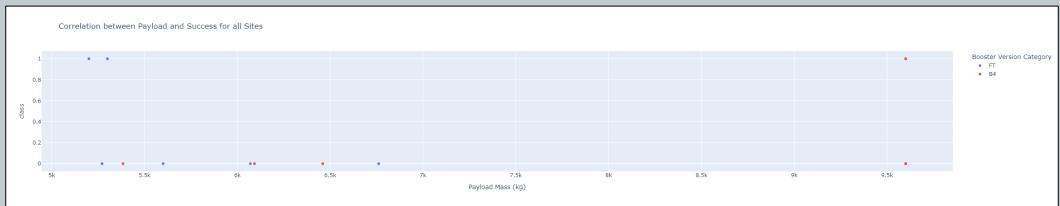
• The KSC LC-39A launch site has a 76.9% success rate and a 23.1% failure for the total launches performed at the site.

## Dashboard – Payload Mass vs Outcome

### Payload Mass vs. Outcome for payload masses between 0 and 5000kg:

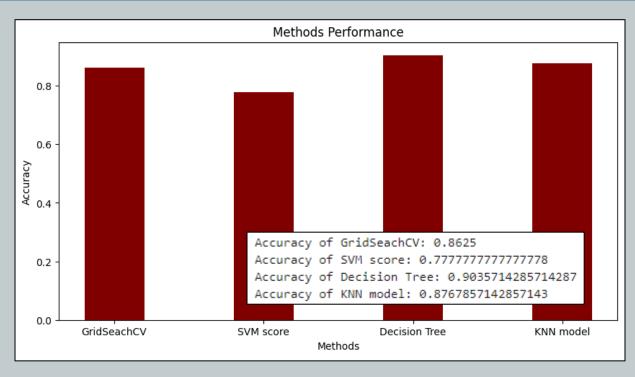


### Payload Mass vs. Outcome for payload masses between 5000 and 5000kg:



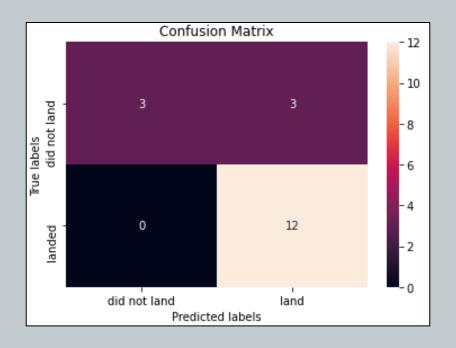


## Classification Accuracy



- The Decision Tree has the highest accuracy score with a value of 90,4%.
- SVM has the lowest accuracy score, with a value of 77,8%.

## Confusion Matrix – Decision Tree Model



- The confusion matrix predicts 12 true positives, 3 false positives, 3 true positive, and 0 false negative.
- The false positives may indicate the model is over-predicting the positive class.

### Conclusions

- A successful mission is explained relies on a set of factors: launch site, orbit and especially the number of previous launches. It's observable that the more recent the data, the most success launches are.
- The orbits with the best success rates are GEO, HEO, SSO, ES-L1.
- Low weight payloads show a better performance than heavy weighted payloads. When analysing the
  different orbits, it's also possible to conclude that the payload can be a factor for the success of the
  mission.
- Launch site KSC LC-39A has the best success rate when compared to all the sites analysed.
- The Decision Tree Algorithm proved to be best model, with the highest train accuracy.

# Appendix

**GitHub Complete Repository** 



## **THANK YOU**

