

# Winning Space Race with Data Science

Shibil PM 27-03-2023



## **Outline**

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

## **Executive Summary**

Summary of methodologies

Collecting data through API

Collecting data through web scraping

Data wrangling

Exploratory analysis using SQL

Exploratory analysis using visualization

Interactive visual analytics with folium and dash

Machine learning prediction

Summary of all results

Exploratory analysis result

Predictive analytics result

### Introduction

Project background and context

To evaluate Space X, for Space Y to compete.

Problems you want to find answers

What factors determine if the rocket will land successfully?

The best way to estimate the total cost for launches?

Where is the best place to make lauches?



## Methodology

#### **Executive Summary**

- Data collection methodology:
  - Space X API (<a href="https://api.spacexdata.com/v4/rockets/">https://api.spacexdata.com/v4/rockets/</a>)
  - Web Scrapping (Wikipedia)
- Perform data wrangling
  - Collected data and cleaned properly through 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
  - The data normalized and split to training and test sets. Using these sets evaluated accuracy and score for four different models

## **Data Collection**

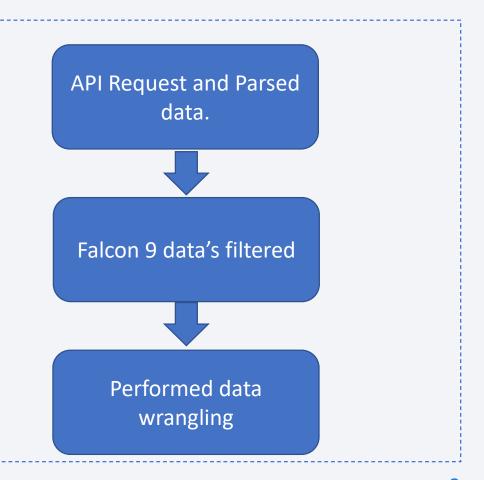
- Data collected from Space X API.
- Data collected using Web Scrapping Wikipedia.

## Data Collection - SpaceX API

 Data collected through get request and it was in json format after it stored in to a pandas data frame using .json\_normalize. And performed data wrangling.

#### • Github:

https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/jupyter-labs-spacex-data-collection-api%20(1).ipynb

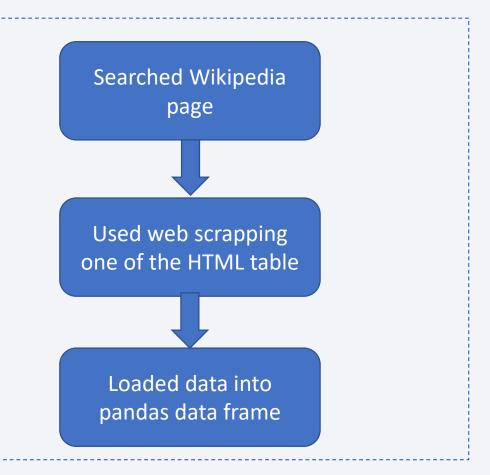


## Data Collection - Scraping

 Data collected through web scrapping. Utilized one of the Wikipedia pages and data scrapped from one of the html table.

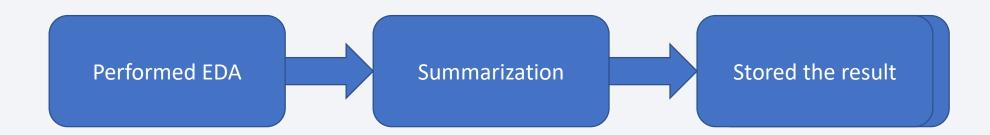
#### • Github:

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



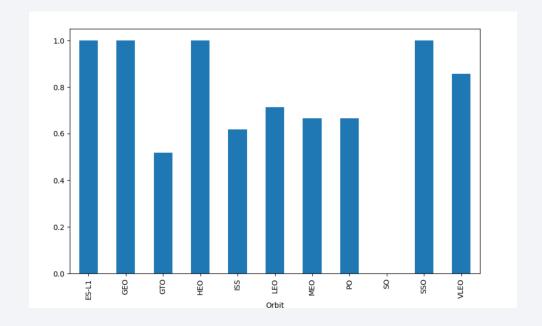
## **Data Wrangling**

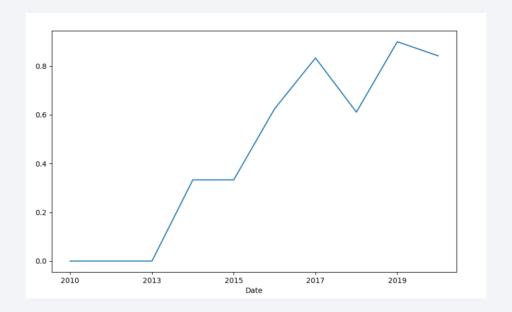
- Performed exploratory data analysis and determined the training labels.
- Calculated the number of launches at each site, and the number and occurrence of each orbits
- Created landing outcome label from outcome column and exported the results to csv.
- Github: <a href="https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/spacex-Data-20wrangling.ipynb">https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/spacex-Data-20wrangling.ipynb</a>



## **EDA** with Data Visualization

- Explored the data by visualizing the relationship between flight number and launch Site, payload and launch site, success rate of each orbit type, flight number and orbit type, the launch success yearly Add the
- Github: <a href="https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-dataviz.ipynb.ipynb">https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-dataviz.ipynb.ipynb</a>





## **EDA** with SQL

- Names of the unique launch sites in the space mission
- 5 records where launch sites begin with the string 'CCA'
- The 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 acheived.
- Names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- Total number of successful and failure mission outcomes
- Names of the booster\_versions which have carried the maximum payload mass. Use a subquery
- 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

12

• Github: <a href="https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/sql-coursera%20(1).ipynb">https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/sql-coursera%20(1).ipynb</a>

## Build an Interactive Map with Folium

- Marked all launch sites, and added map objects such as markers, circles, lines to mark the success or failure of launches for each site on the folium map.
- Assigned the feature launch outcomes (failure or success) to class 0 and 1.
- Using marker clusters, we identified which launch sites have relatively high success rate.
- Github: <a href="https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/jupyter launch site location.ipynb">https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/jupyter launch site location.ipynb</a>

## Build a Dashboard with Plotly Dash

Graphs and plots used:

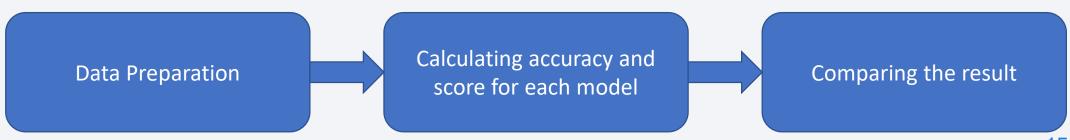
Percentage of launches by site (Pie Chart)

Payload range(scatter plot)

- This combination allowed to quickly analyze the relation between payload and launch sites
- It helped to to find where is the best place to launch according to payloads
- Github: <a href="https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/dashboard.py">https://github.com/ShibilPM/Applied-Data-Science-Capstone/blob/main/dashboard.py</a>

## Predictive Analysis (Classification)

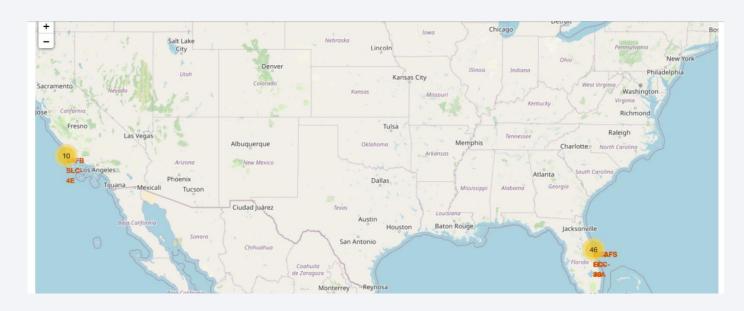
- Loaded the data using numpy and pandas, transformed the data, split our data into training and testing.
- Built different machine learning models and tune different hyperparameters using GridSearchCV.
- Found the best performing classification model.
- Github: <a href="https://github.com/ShibilPM/Applied-Data-Science-">https://github.com/ShibilPM/Applied-Data-Science-</a>
   Capstone/blob/main/SpaceX Machine Learning Prediction Part 5.ipynb



#### Results

- Exploratory data analysis results:
  - -SpaceX uses 4 different launch sites.
  - -Two booster versions failed at landing drone ships in 2015.
  - -Many Falcon 9 booster version were successful at landing in drone ships having payload above average.
  - -The number of landing outcomes became as better as years passed.

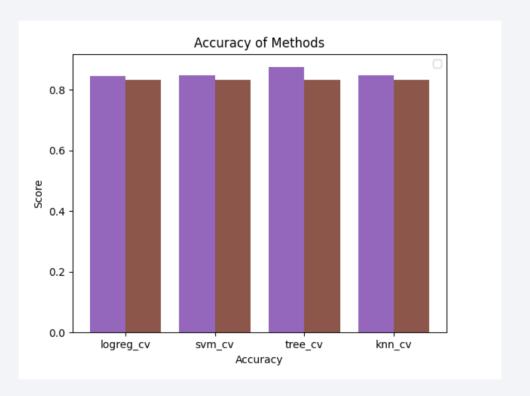
## Results



Most launches happend at east cost launch sites.

## Results

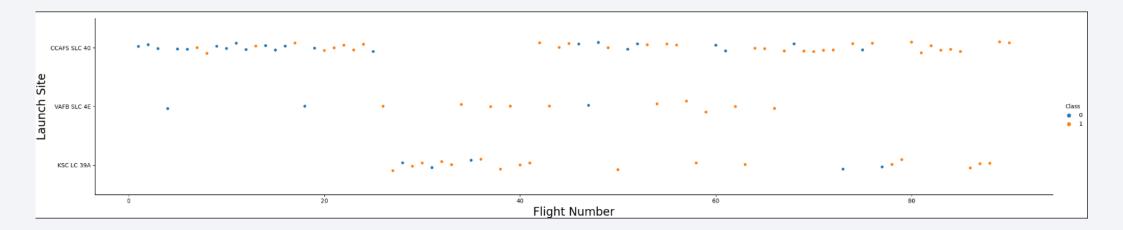
• Decision Tree Classifier is the best model having accuracy over 87.5%.





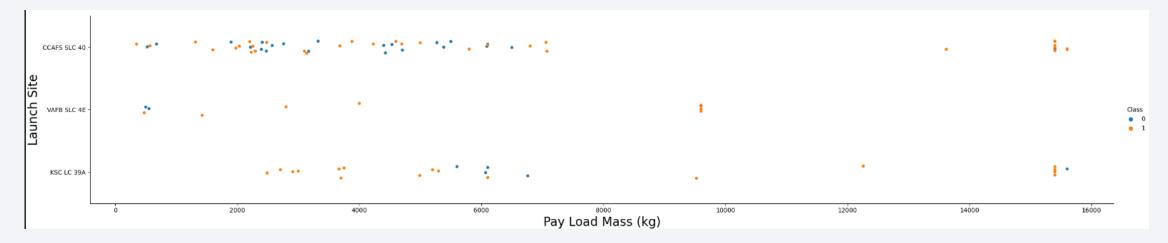
## Flight Number vs. Launch Site

- Flight Number vs. Launch Site
- -Found that the larger the flight amount at a launch site, the greater the success rate at a launch site.



# Payload vs. Launch Site

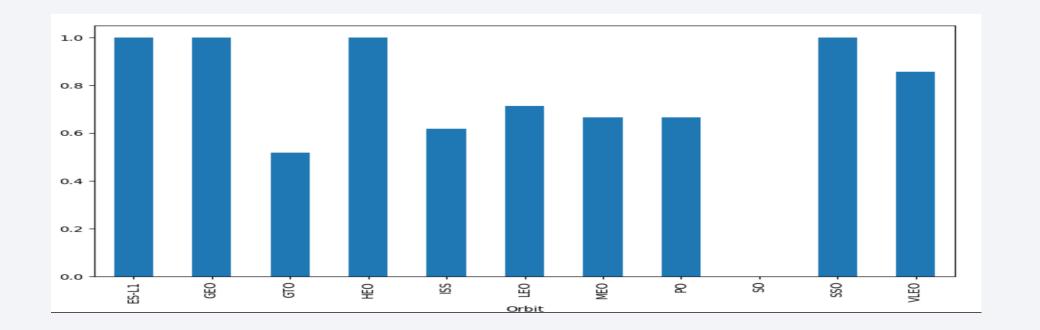
- Show a scatter plot of Payload vs. Launch Site
  - -Payload over 9000kg have excellent success rate.



# Success Rate vs. Orbit Type

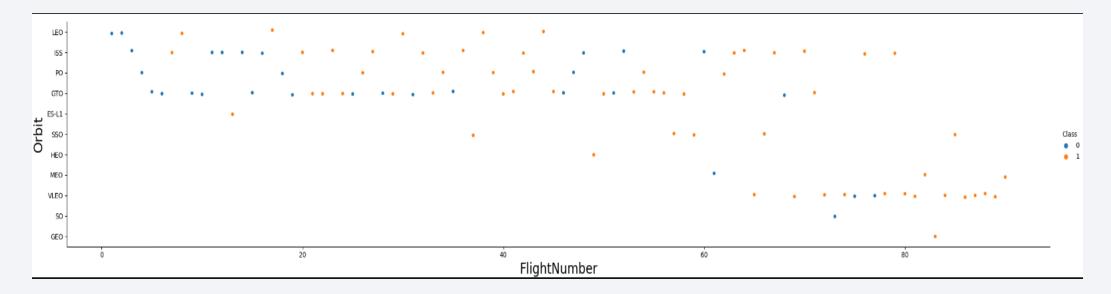
success rate of each orbit type

-ES-L1, GEO, SSO and HEO have most success rates.



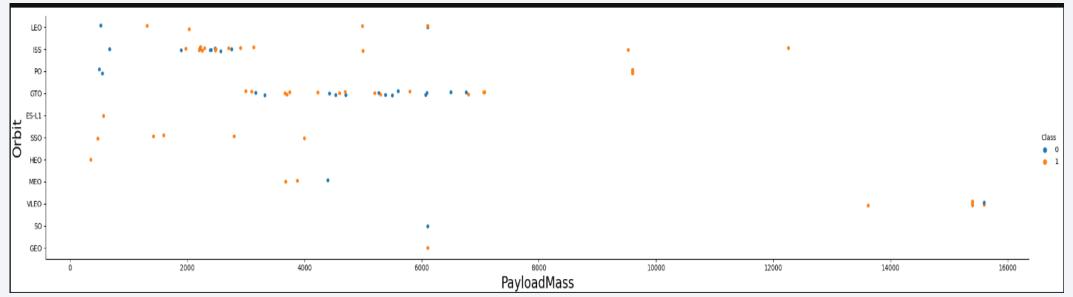
# Flight Number vs. Orbit Type

- Flight number vs. Orbit type
  - -Success rate improves over time for all orbits.



# Payload vs. Orbit Type

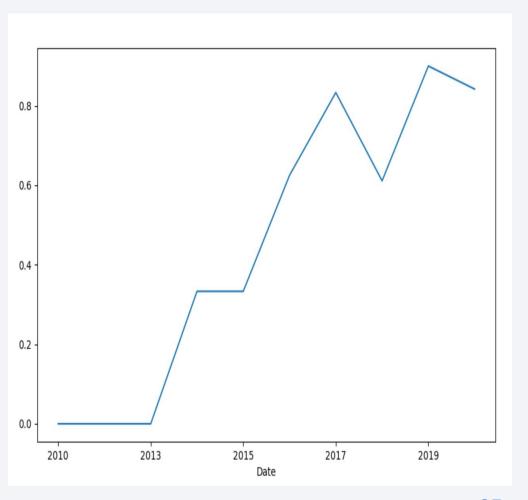
• payload vs. orbit type



## Launch Success Yearly Trend

yearly average success rate

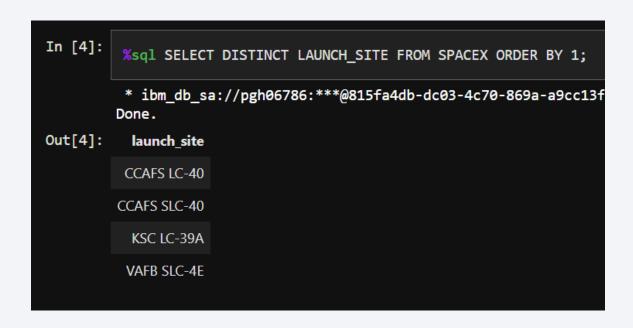
-Success rate since 2013 kept on increasing till 2020.



### All Launch Site Names

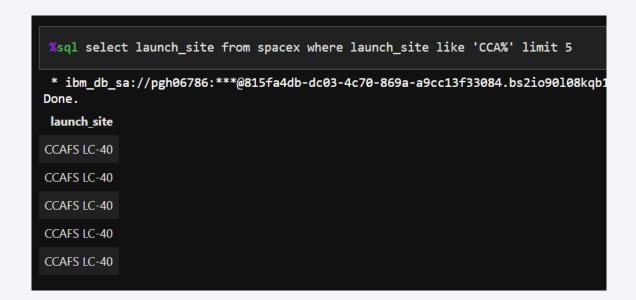
Find the names of the unique launch sites

Found the launch site names without any duplicates using Distinct



## Launch Site Names Begin with 'CCA'

- Find 5 records where launch sites begin with `CCA`
  - Fetched data using like keyword



## **Total Payload Mass**

- Calculate the total payload carried by boosters from NASA.
  - -Found the total using sum() function

## Average Payload Mass by F9 v1.1

- Calculate the average payload mass carried by booster version F9 v1.1
  - Found the average using AVG() function.

## First Successful Ground Landing Date

Find the dates of the first successful landing outcome on 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
  - Used between keyword to select the range.

```
In [15]: %sql SELECT DISTINCT BOOSTER_VERSION FROM SPACEX WHERE PAYLOAD_MASS_KG_ BETWEEN 4000 AND 6000 AND LANDING_OUTCOME = 'Success (drone ship)'

* ibm_db_sa://pgh06786:***@815fa4db-dc03-4c70-869a-a9cc13f33084.bs2io90108kqb1od8lcg.databases.appdomain.cloud:30367/bludb
Done.

Out[15]: booster_version

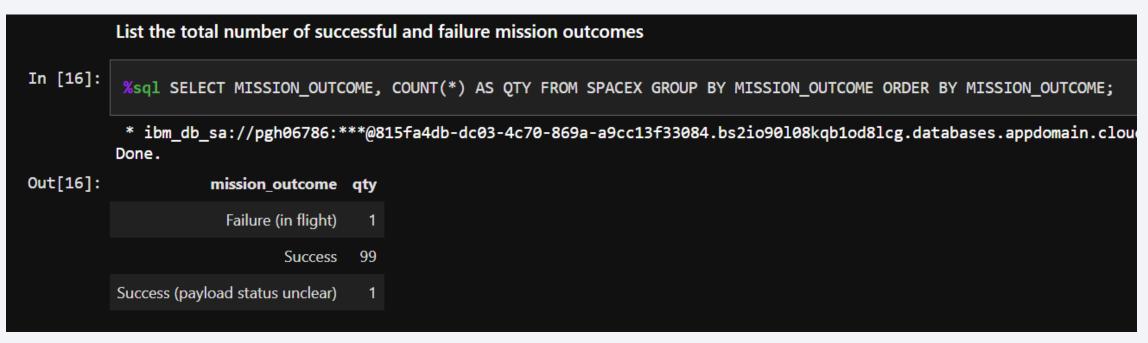
F9 FT B1021.2

F9 FT B1022

F9 FT B1026
```

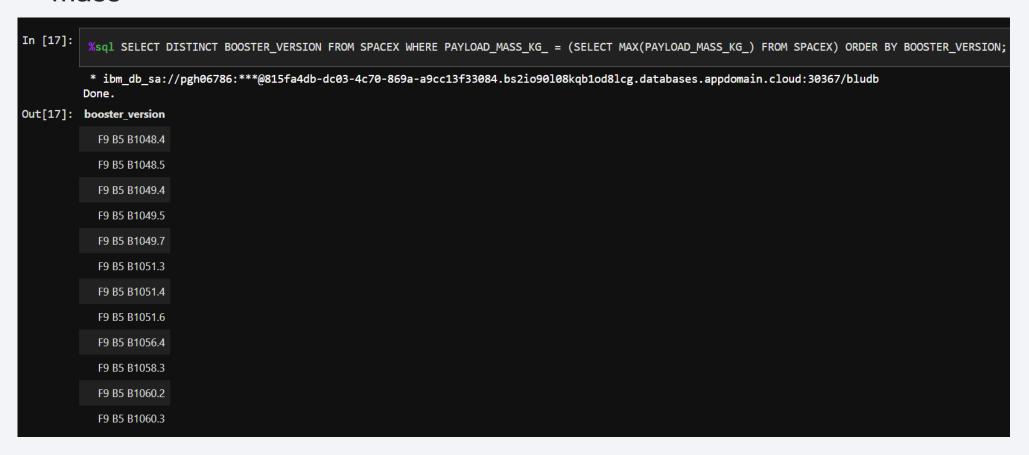
#### Total Number of Successful and Failure Mission Outcomes

Calculate the total number of successful and failure mission outcome.



## **Boosters Carried Maximum Payload**

List the names of the booster which have carried 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

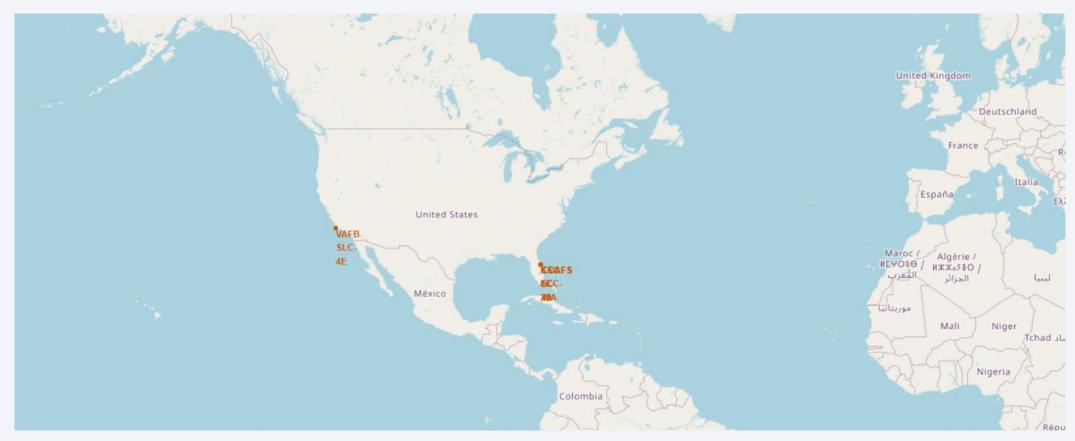
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

 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

In [23]:	sql SELECT LANDIN	IG_OU	COME, COUNT(*) AS QTY FROM SPACEX WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20' GROUP BY LANDING_OUTCOME ORDER BY	QTY DE
	* ibm_db_sa://pgh Done.	10678	6:***@815fa4db-dc03-4c70-869a-a9cc13f33084.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:30367/bludb	
Out[23]:	landing_outcome	qty		
	No attempt	10		
	Failure (drone ship)	5		
	Success (drone ship)	5		
	Controlled (ocean)	3		
	Success (ground pad)	3		
	Failure (parachute)	2		
	Uncontrolled (ocean)	2		
	Precluded (drone ship)	1		



#### **Launch Sites**



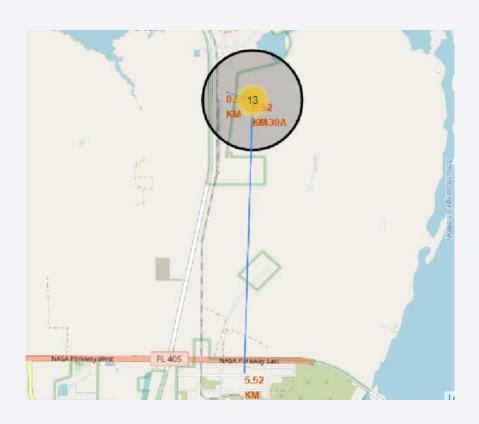
SpaceX launch sites are in united states ameria coasts.

# Outcomes by sites



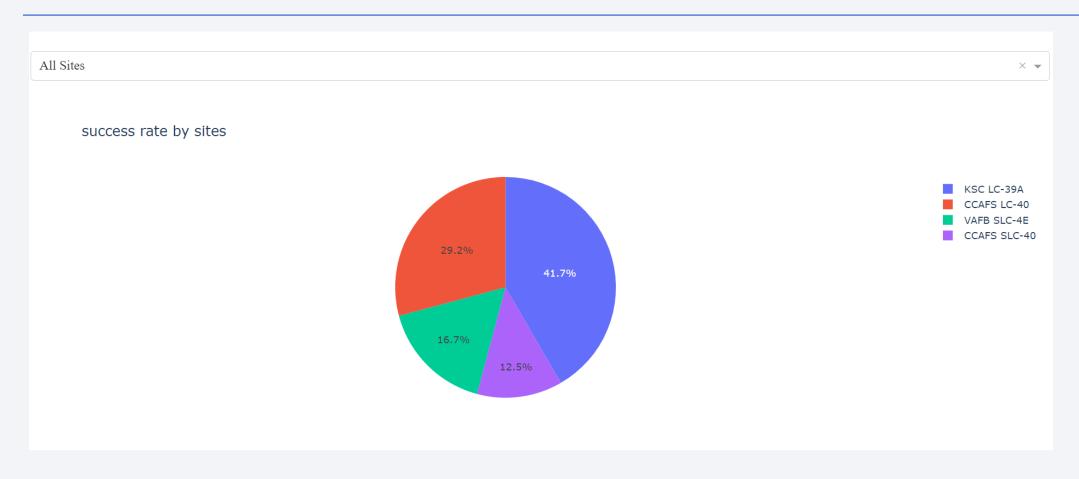
• Green is success and red is failure

### Distance to landmarks



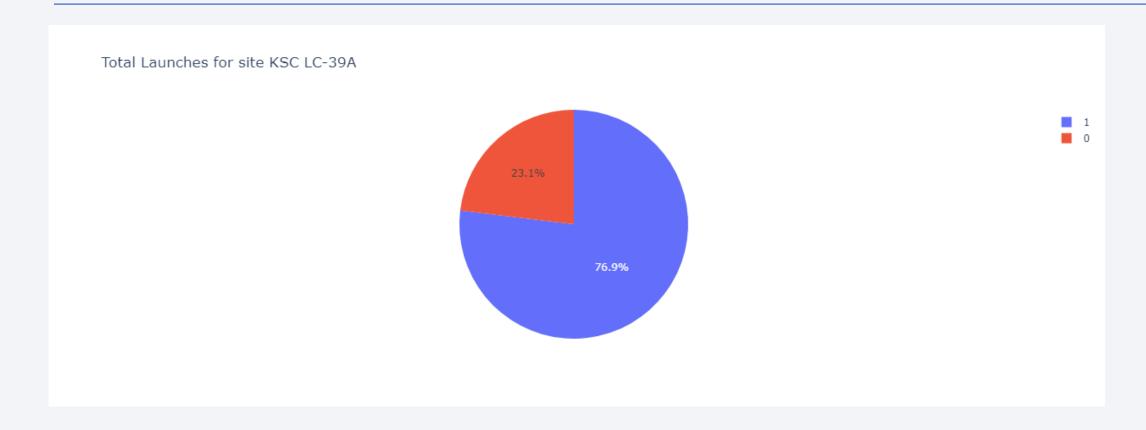


# Success rate for each launching site



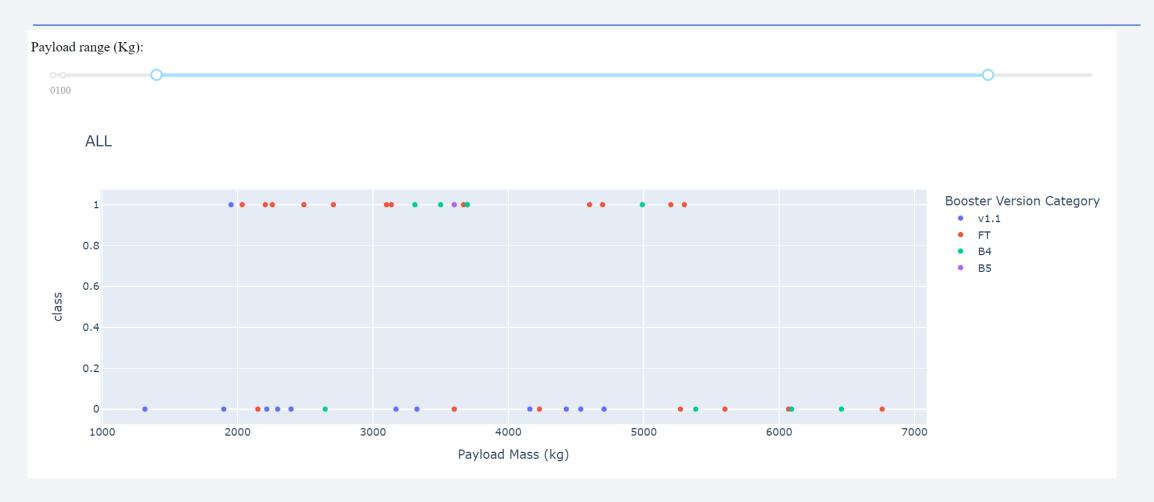
KSC LC-39A have most success rate.

# Launch site with highest success ratio



76.9% Success and 23.1% Failure.

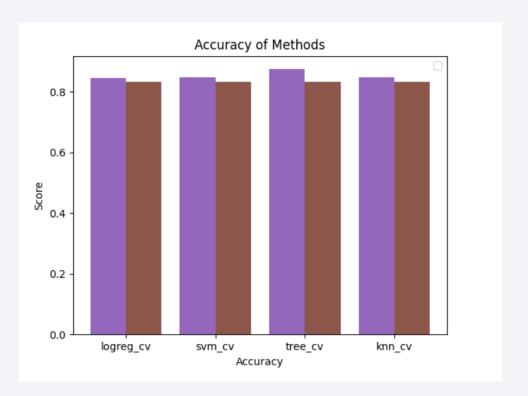
# Payload vs class



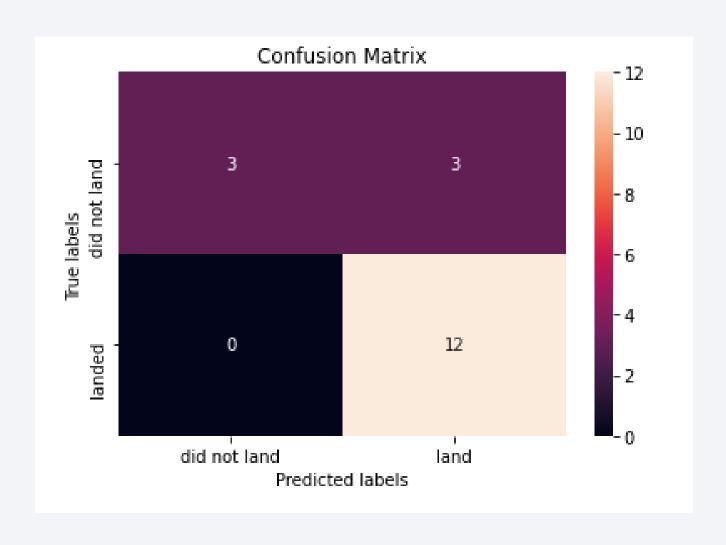


## **Classification Accuracy**

• Decision Tree Classifier is the best model having accuracy over 87.5%.



## **Confusion Matrix**



#### Conclusions

- Launch success rate started to increase in 2013 till 2020.
- Launches above 7000kg are less risky
- KSC LC-39A had the most successful launches of any sites.
- The Decision tree classifier is the best machine learning algorithm for this task.

