

# Winning Space Race with Data Science

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### Outline

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

### **Executive Summary**

#### **Summary of methodologies**

- -Data collection
- -Data wrangling
- -EDA with data visualization
- -EDA with SQL
- -Exploring and preparing data
- -Interactive visual analysis with Folium
- -Building interactive HTML dashboard with Dash Pllotly
- -Predictive analysis(classification methods)

#### Summary of all results

- -EDA for best launching predictions.
- -Machine Learning Prediction shows the best model for predict if the 1st launch will land successfully, mostly in the oceans.

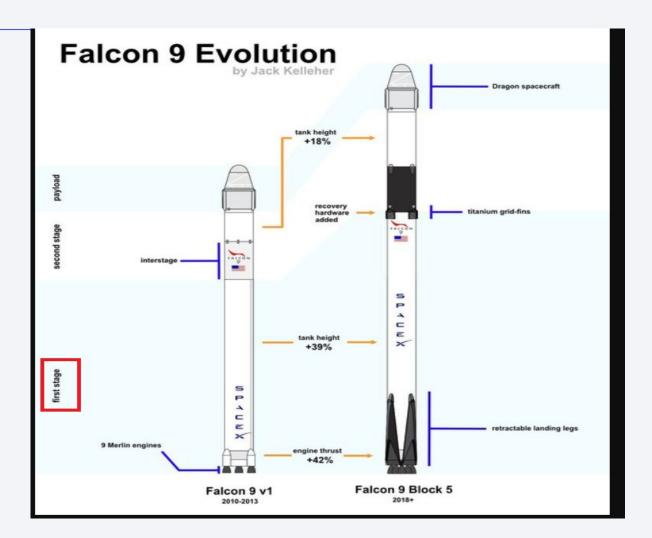
#### Introduction

#### Project background and context

-Space X can save a lot of money by reusing the first stage of each launch,so our job is to adapt and study their data so we can use it in a rival company named space Y.

# Problems you want to find answers

-Check the best launch site for successful launches and predict if the first stage will land correctly so it could be reused ,therefore saving big money for Space Y





### Methodology

#### **Executive Summary**

- Data collection methodology:
  - -Space X API
  - -Web scraping from wikipedia
- Perform data wrangling
  - -Cleaning the data ,eliminate null values and less important columns,add classification column
- 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 Collection**

Describe how data sets were collected.

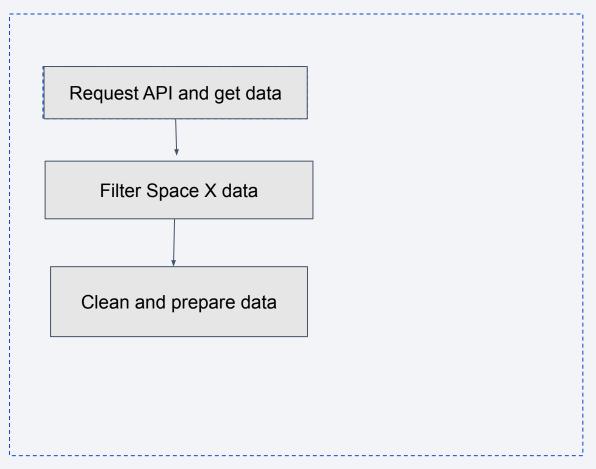
- Data sets were collected via Wikipedia with BeautifulSoup and Space x API,using python programming language with pandas library.

### Data Collection – SpaceX API

-The data was request from Spacex X API, from it we can get all data necessary for the project.

#### -Github link:

https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab1%20Data%20Collection%20API%20(Application%20Programming%20Interface).ipvnb

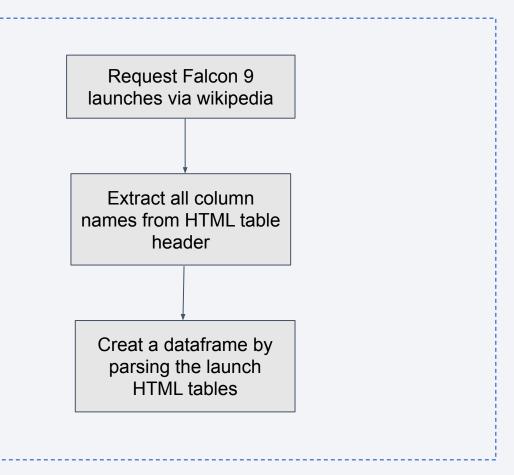


### Data Collection - Scraping

Data was collected via wikipedia with
 BeautifulSoup: <a href="https://en.wikipedia.org/w/index.php?title=List\_of-Falcon\_9\_and\_Falcon\_Heavy\_launches&oldid=1027686922.">https://en.wikipedia.org/w/index.php?title=List\_of-Falcon\_9\_and\_Falcon\_Heavy\_launches&oldid=1027686922.</a>. Then extract Falcon 9 launch records in HTML table, parsing the data and convert to pandas dataframe.

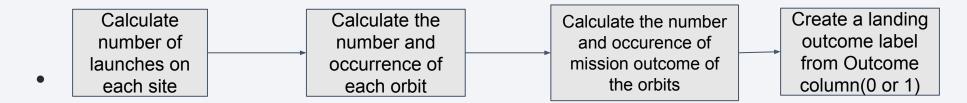
#### -Github link:

https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab2%20Data%20Collection%20with%20Web%20Scraping.ipynb



### **Data Wrangling**

-Convert the successful launches in 1 and 0 for unsuccessful and add to our dataframe for further analysis.



#### • Github link::

https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab3%20Data%20W rangling.ipynb

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

#### Charts:

The scatter plot was used because of the relationship between 2 variables (0 and 1 for class), the bar plot was used between orbit group and successful rate(fixed value per orbit) and with variables changing ver time we can use a line plot

For payload mass and Flight number for each site was plotted a scatter plot, for orbit type vs successful rate was plotted a bar plot and for successful rate per date was plotted a line plot.

#### Github Link:

https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab5%20Graphs%20Exploring%20and%20Preparing%20Data.ipynb

#### **EDA** with SQL

#### • SQL queries :

- 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 acheived.
- 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 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 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 Link:

### Build an Interactive Map with Folium

- folium.Map:Creattion of a map with center location to be NASA Johnson Space Center at Houston, Texas.
- folium.Circle:highlighted circle area with a text label on a specific coordinate
- folium.Marker:to give more information of certain points

#### Github link:

Images-https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab6%20images.png

Lab-https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab6%20Interactive%20Visual%20Analytics%20with%20Folium.ipynb

### Build a Dashboard with Plotly Dash

#### Charts:

- Pie Chart
- Scatter Chart

Pie chart is useful because we can visualize by parts and colours the percentage os successful launches by launch site, is more interactive, easier and faster to analyze.

SCatter charts are good to analyze large data with numerical variables in this case 0 and 1.

• Github Link:

https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab7 %20spacex\_dash.py

# Predictive Analysis (Classification)

• First we standardize the data then split in train and test data to be evaluate by 4 methods o classification (LOG Regression,SVM,KNN and Decision Tree),check which one has the best accuracy

Standardize the data

X\_train, X\_test, Y\_train, Y\_test Split data

GridSearchCV and apply each method to find confusion matrix

Analyze the results with a chart to check the best accuracy in the 4 methods

Github link:

https://github.com/Brunoamaral0/Coursera-capstone-project-data-science/blob/main/Lab8 %20Machine%20Learning%20Prediction.ipynb

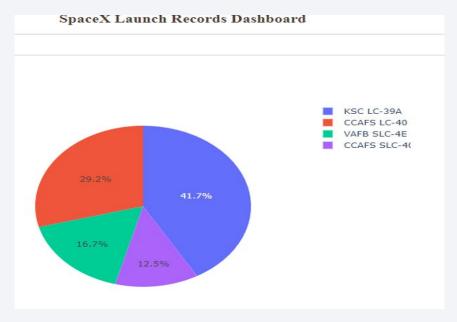
#### Results

#### Exploratory data analysis results

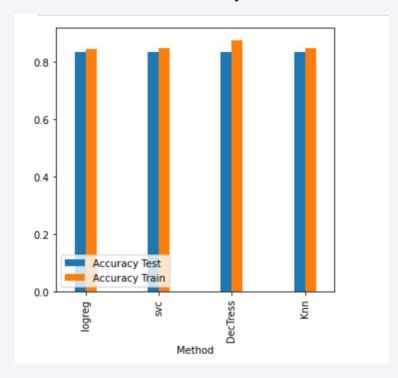
- -CCAFS LC-40, has a success rate of 60 %, while KSC LC-39A and VAFB SLC 4E has a success rate of 77%.(in flight number x lunch site y graph)
- VAFB-SLC launchsite there are no rockets launched for heavy payload mass(greater than 10000).
- **¬**Orbits type with best successful rate are ES-L1,GEO,HEO,SSO
- -Success rate since 2013 kept increase, heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISSing till 2020
- -Success rate of launches since 2013 kept increasing till 2020

#### Dash results (KSC and CCAFS have highest success launch rate)

Note:dataframe used in dahs is different from the one used EDA

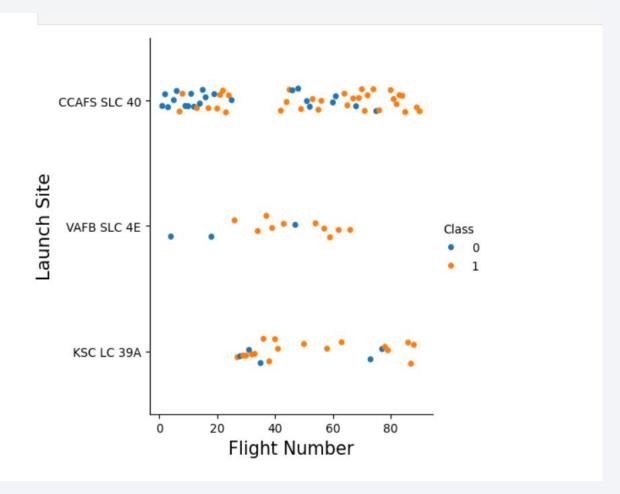


#### Predictive analysis results





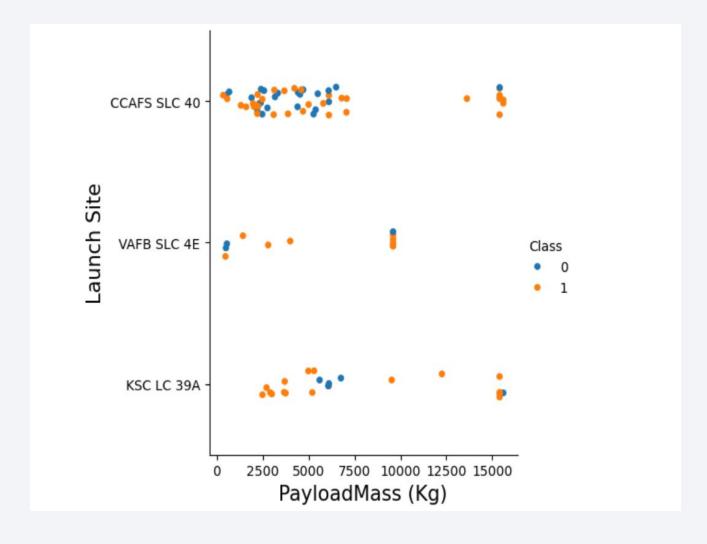
### Flight Number vs. Launch Site



In the graphs we can verify that the launch sites with highest rate of success landings are KSC and VAFB with 15 successful and 5 unsuccessful ,rate of 15/20=75% for KSC and 10 successful and 3 unsuccessful for VAFB rate =77%.

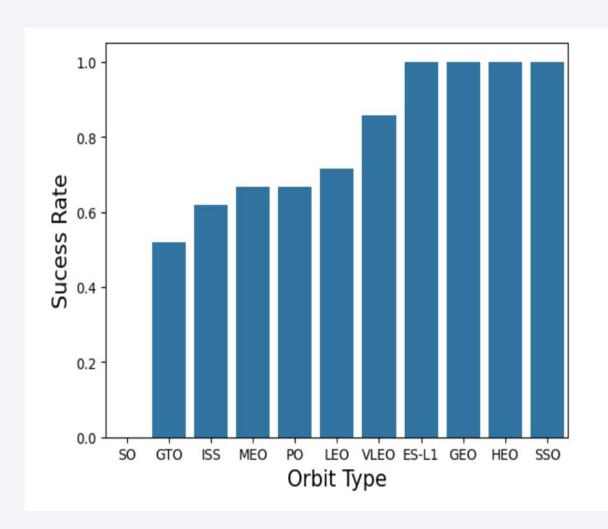
CCAFS has only 60 % with 33 successful and 22 unsuccessful rate =60%.

### Payload vs. Launch Site



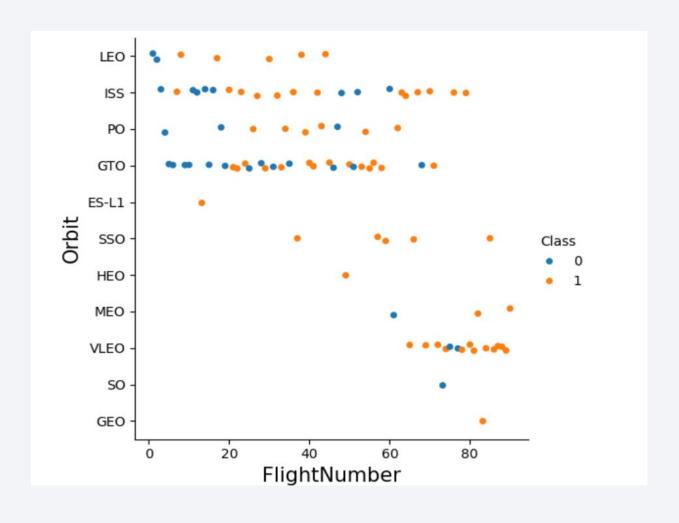
- CCAFS has good success rate with payloads higher than 10 tons with 5 successful and 1 unsuccessful rate =83,3%.
- VAFB has no record for payloads above 10 tons while under 10 tons the results are inconclusive due the lack of launches.
- KSC has all launches successful for payloads under or equal 5 tons, while between 5 tons and 7,5 tons the launches were all unsuccessful, above 10 tons there were only one successful launch to 5 successful, but due to lack of launches we can't say that if we use payload with more than 10 tons the launch has high rate for success.

### Success Rate vs. Orbit Type



- The orbits with highest success rate were ES-L1,GEO,HEO,SSO .ls important to emphasize that ES-L1,GEO,HEO only made one launch so very inconclusive, while SSO did 5 launches all succeed.
- VLEO still the best orbit for prediction next launch due to high rate,85% and 14 launches
- All the other orbits are inconclusive very few launches.

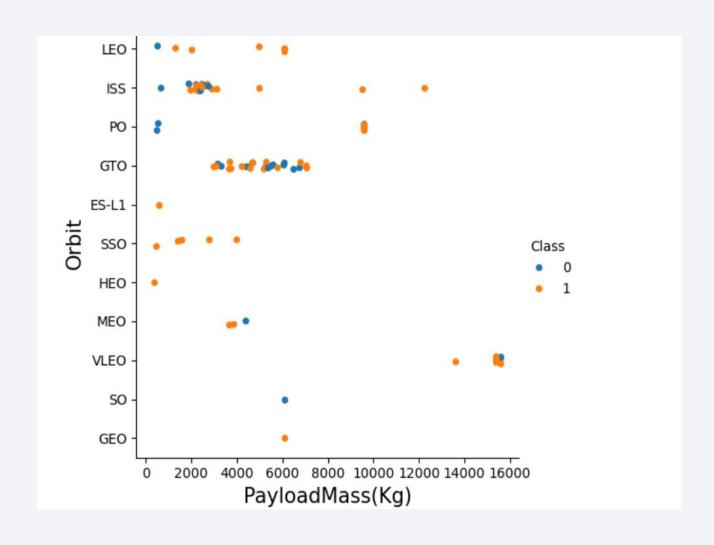
# Flight Number vs. Orbit Type



- SSO did 5 launches all succeed,100% success rate.
- VLEO is the best orbit for prediction next launch due to high rate,85% and 14 launches

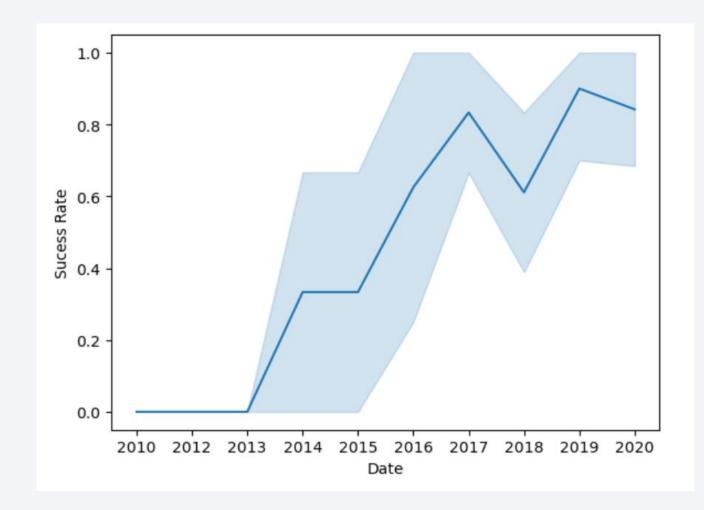
No launches were made with SO orbit

# Payload vs. Orbit Type



- SSO orbit had payloads mass all under or equal 4 tons.
- VLEO orbit had all launches higher than
   12 tons and less than 16 tons
- GTO orbit many launches between payload masses of 2 and 8 tons

# Launch Success Yearly Trend



- Success rate has grown over time since 2013.
- In the period of 2016 and 2018, success decreased by 20 %.

#### All Launch Site Names

#### Launch\_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

The mission has 4 launch sites in our dataframe.

# Launch Site Names Begin with 'CCA'

| Date           | Time<br>(UTC) | Booster_Version | Launch_Site     | Payload   | PAYLOAD_MASS_KG_ | Orbit        | Customer              | Mission_Outcome | Landing_Outcome     |
|----------------|---------------|-----------------|-----------------|---|------------------|--------------|-----------------------|-----------------|---------------------|
| 2010-<br>06-04 | 18:45:00      | F9 v1.0 B0003   | CCAFS LC-<br>40 | Dragon<br>Spacecraft<br>Qualification<br>Unit                                   | 0                | LEO          | SpaceX                | Success         | Failure (parachute) |
| 2010-<br>12-08 | 15:43:00      | F9 v1.0 B0004   | CCAFS LC-<br>40 | Dragon<br>demo flight<br>C1, two<br>CubeSats,<br>barrel of<br>Brouere<br>cheese | 0                | LEO<br>(ISS) | NASA<br>(COTS)<br>NRO | Success         | Failure (parachute) |
| 2012-<br>05-22 | 7:44:00       | F9 v1.0 B0005   | CCAFS LC-<br>40 | Dragon<br>demo flight<br>C2   | 525              | LEO<br>(ISS) | NASA<br>(COTS)        | Success         | No attempt          |
| 2012-<br>10-08 | 0:35:00       | F9 v1.0 B0006   | CCAFS LC-       | SpaceX<br>CRS-1   | 500              | LEO<br>(ISS) | NASA<br>(CRS)         | Success         | No attempt          |
| 2013-<br>03-01 | 15:10:00      | F9 v1.0 B0007   | CCAFS LC-       | SpaceX<br>CRS-2   | 677              | LEO<br>(ISS) | NASA<br>(CRS)         | Success         | No attempt          |
| 4              |               |                 |                 |   |                  |              |                       |                 | <b>*</b>            |

- Exemple with 5 launch sites started by "CCA" in our dataframe and their labels.

### **Total Payload Mass**

```
%sql SELECT sum(PAYLOAD_MASS__KG_) as sum from SPACEXTABLE where Customer like '%NASA (CRS)%'
 * sqlite:///my_data1.db
Done.
  sum
  48213
 row 76 contain [NASA (CRS), Kacific 1] ,if is only NASA (CRS) the code must be
  %sql SELECT sum(PAYLOAD_MASS__KG_) as sum from SPACEXTABLE where Customer ='NASA (CRS)'
 * sqlite:///my_data1.db
Done.
  sum
  45596
```

# Average Payload Mass by F9 v1.1



- Average payload mass by F9 v1.1 under 3 tons.
- Average of 2928 kg

# First Successful Ground Landing Date

min (Date)

2015-12-22

- Date where first successful landing outcome in ground pad was achieved.

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



 Only 4 boosters landing in drone ship successful with weights between 4 and 6 tons.

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



- 98 successful outcomes

- 3 outcomes were not 100 % successful

# **Boosters Carried Maximum Payload**

#### 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

- All boosters recorded that carry the maximum payload mass

### 2015 Launch Records

| Month Name | Booster_Version | Landing_Outcome      | Launch_Site |
|------------|-----------------|----------------------|-------------|
| 1          | F9 v1.1 B1012   | Failure (drone ship) | CCAFS LC-40 |
| 4          | F9 v1.1 B1015   | Failure (drone ship) | CCAFS LC-40 |

- Failure landing outcomes in drone ship in the year 2015 in months of january and april for boosters B1012 and B1015.

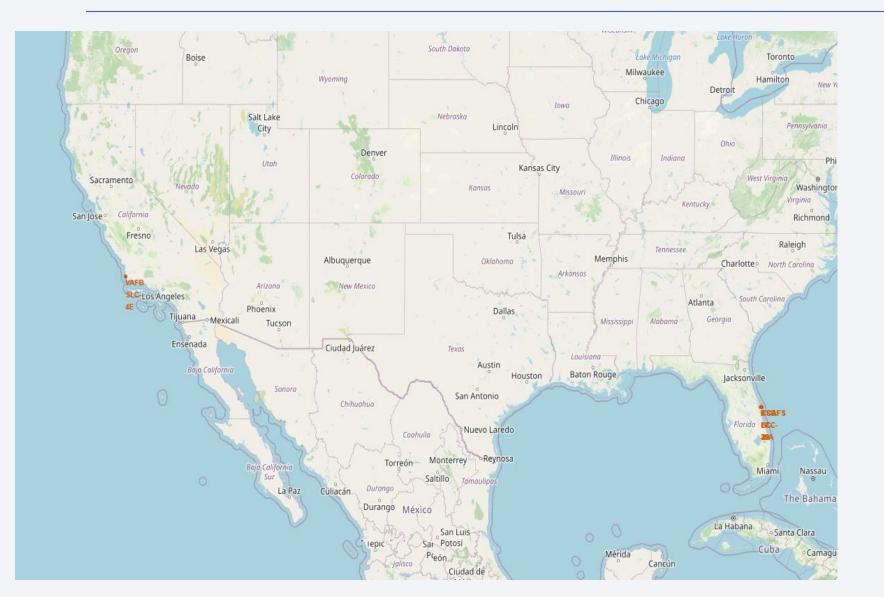
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

| number | Landing_Outcome      |  |  |
|--------|----------------------|--|--|
| 9      | Success (ground pad) |  |  |
| 5      | Failure (drone ship) |  |  |

- 9 launches successfully landing on ground
- 5 failures in landing via drone ship

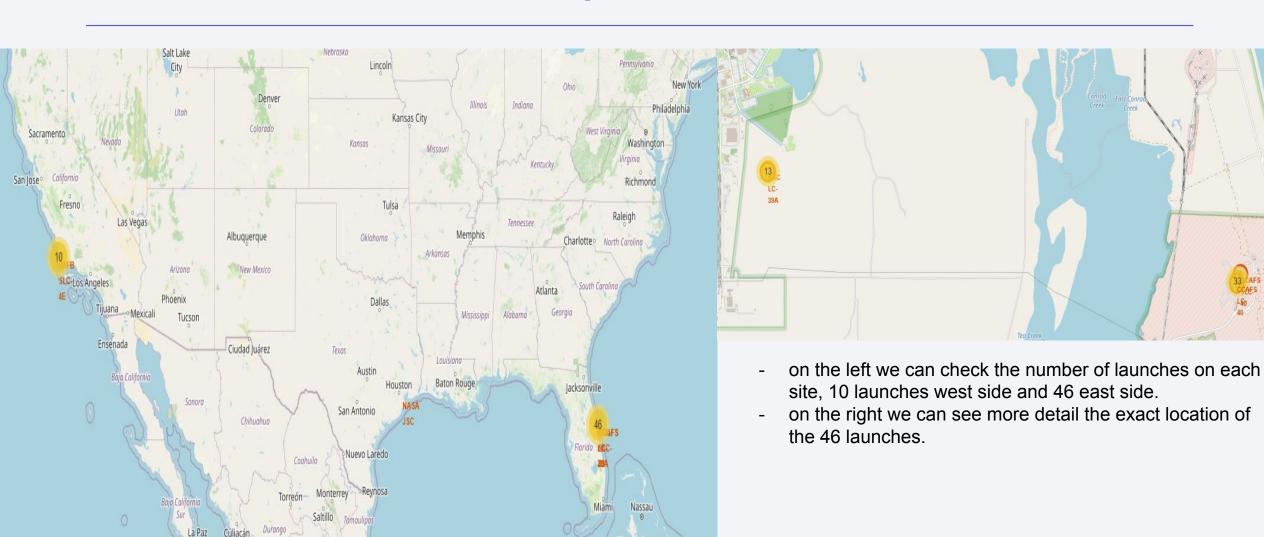


### Launch sites



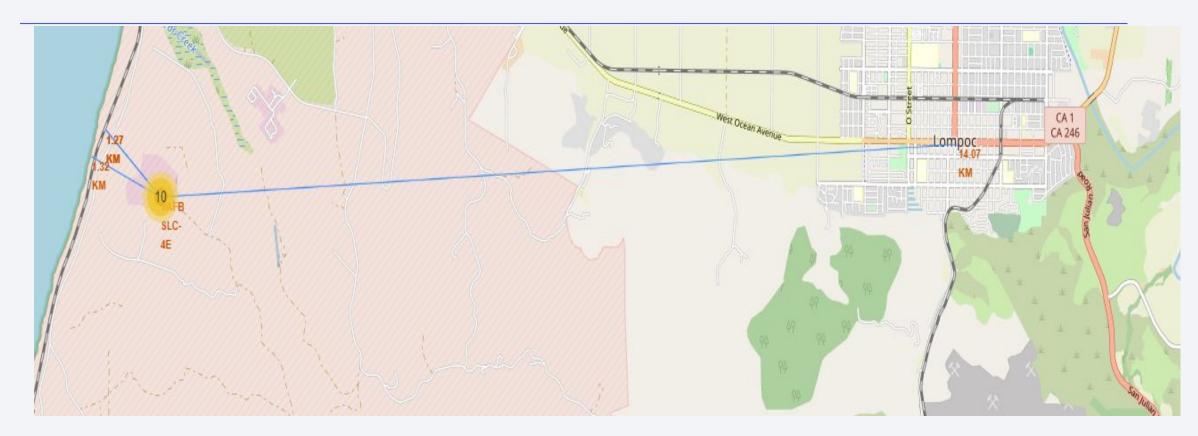
- All launch sites made on the coast.
- Far from cities.

# Number of launches per site



The Bahamas

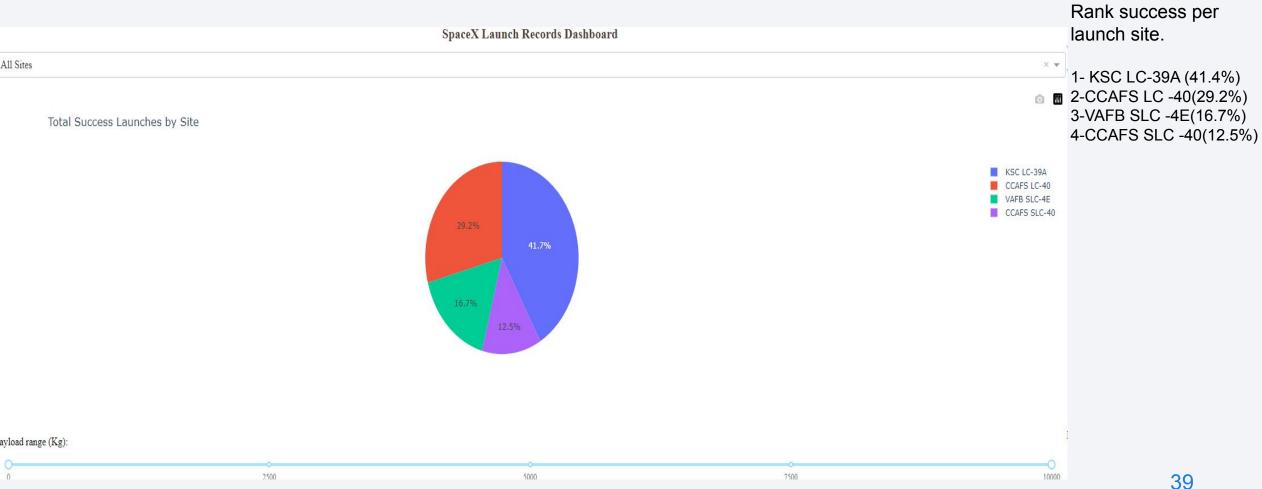
# Nearby distances (Highway, Railway, Coastline)



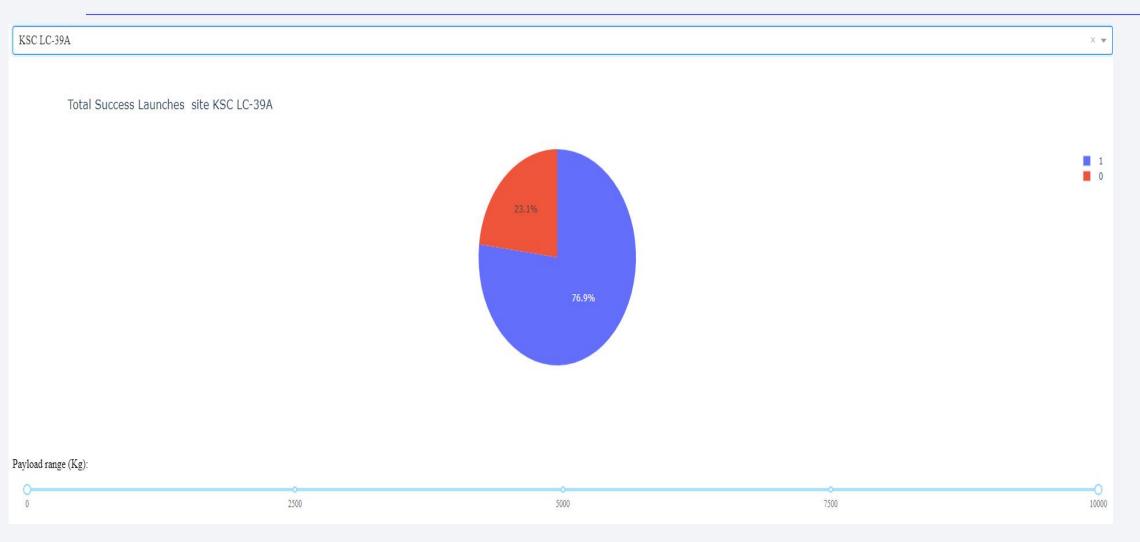
- Nearby distances in the launches sites from the west site map location
- 1 km and 270m from the nearest railway,1km and 320m from the nearest coastline and 14 km and 70 m from nearest highway



# Successful launches per site



# Launch site with highest sucess



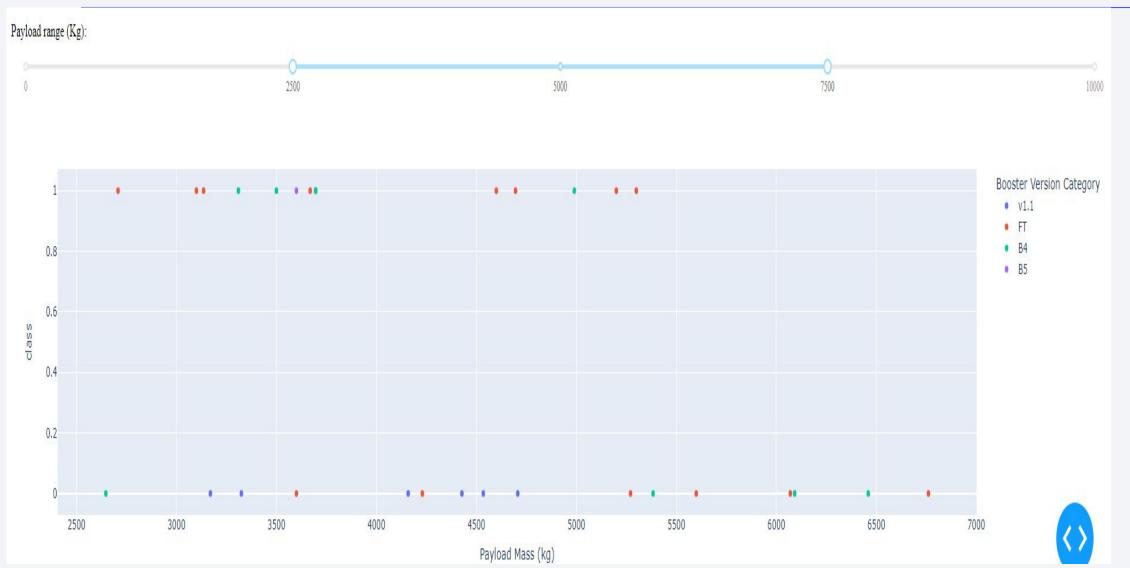
- Success

:77 %

- Unsuccess

: 23.1%

# Payload vs Launch Outcome

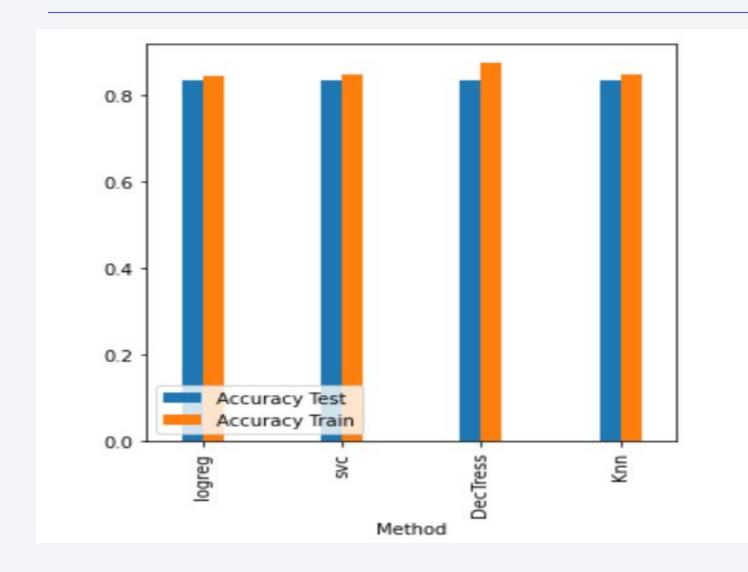


Payload mass: 2.5 tons to 7.5 tons.

Highest success with FT boosters version ,followed by B4.



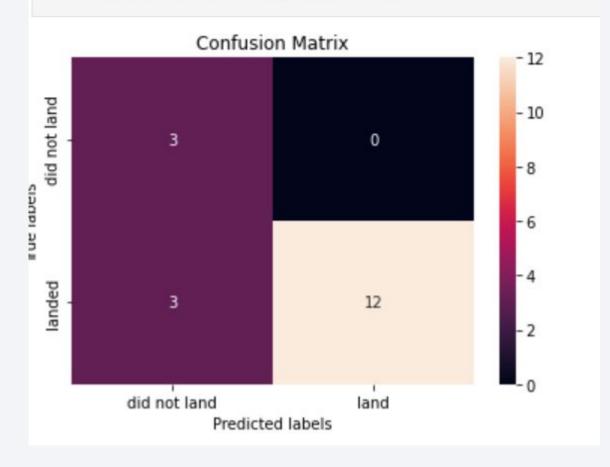
# **Classification Accuracy**



 WE can verify that the best method for training model is the Decision tree method while at same time is the worst for testing model. The other 3 models are very similar in both test and train data.

#### **Confusion Matrix**

yhat=tree\_cv.predict(X\_test)
plot\_confusion\_matrix(yhat,Y\_test)



- 12 true positives and 3 false negatives
- 3 false positives
- 3 launches bad predicted in total of 18

#### Conclusions

- After analyze all the work done from the data collected, we can say that the best launch site was KSC LC 39A nearby coastlines , relative to the weight of the payloads ,under 5 tons and above 10 tons the success of landing and reuse the 1st stage is higher, success rate has grown over time since 2013 due the quality of the spaceships and the quality and treatment of the data, we can use the 4 machine learning prediction methods due to their proximity of accuracy ,but decision tree one has slightly higher accuracy in train test, so must be better for future launches to our company Space Y so we can save a lot of money in rocket launches field.



### **Appendix**

- In the folium maps lab the images are attach in png file, because when uploaded on github they don't appear in the jupyter notebook file.
- In the dash plotly dash the images are attach in png file, when can only upload the code of the html
  dashboard on github
- From slide 24 to 33 the queries were obtained via SQL programming language.

