

## Winning Space Race with Data Science

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#### Outline

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- Conclusion
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#### **Executive Summary**

- Summary of methodologies
  - SpaceX Data Collection using SpaceX API
  - SpaceX Data Collection with Web Scraping
  - SpaceX Data Wrangling
  - SpaceX Exploratory Data Analysis using SQL
  - SpaceX EDA DataViz Using Python Pandas and Matplotlib
  - SpaceX Launch Sites Analysis with Folium-Interactive Visual Analytics and Plotly Dash
  - SpaceX Machine Learning Landing Prediction
- Summary of all results
  - EDA results
  - Interactive Visual Analytics and Dashboards
  - Predictive Analysis

#### Introduction

#### Project background and context

SpaceX states that Flacon 9 rocket launches costs about 62 million dollars while other providers costs up to 165 million dollars each, this is because SpaceX can reuse the first stage and saves a lot of money. If it can be determined if the first stage will land, the cost of a launch can be calculated. This information can be used if another company wants to bid against SpaceX for a launch.

#### Problems you want to find answers

This capstone project will predict if the Falcon 9 first stage will successfully land based on data from Falcon 9 rocket launches advertised on its website.



## 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**

How the data sets were collected

Data is first collected using SpaceX API by making a get request.

The SpaceX launch data was requested and parsed using the GET request and the JSON was converted into a Pandas data frame.

## Data Collection – SpaceX API

# Task 1: Request and parse the SpaceX launch data using the GET request To make the requested JSON results more consistent, we will use the following static response object for this project: In [9]: static\_json\_url='https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/datasets/API We should see that the request was successfull with the 200 status response code In [10]: response.status\_code Out[10]: 200 Now we decode the response content as a Json using .json() and turn it into a Pandas dataframe using .json\_normalize() In [12]: # Use json\_normalize meethod to convert the json result into a dataframe data = pd.json\_normalize(response.json())

- Data was collecting using the SpaceX API by making a get request then decoding the response content and turning it into a Pandas dataframe.
- This is the following Github URL of the complete code:

## **Data Collection - Scraping**

- Web scraped to collect Falcon 9 launch record from Wikipedia using BeautifulSoup and request to extract records.
- This is the following Github URL of the complete code:

https://github.com/johnnyzeng-analysis/Coursera-Python-Capstone-Project/blob/main/jupyterlabs-webscraping.ipynb

#### TASK 1: Request the Falcon9 Launch Wiki page from its URL

First, let's perform an HTTP GET method to request the Falcon9 Launch HTML page, as an HTTP response.

```
# use requests.get() method with the provided static_url
# assign the response to a object
response = requests.get(static_url)
```

Create a BeautifulSoup object from the HTML response

```
# Use BeautifulSoup() to create a BeautifulSoup object from a response text content
soup = BeautifulSoup(response.content, 'html.parser')
```

Print the page title to verify if the BeautifulSoup object was created properly

```
# Use soup.title attribute
soup.title
```

<title>List of Falcon 9 and Falcon Heavy launches - Wikipedia</title>

## **Data Wrangling**

- After creating a Pandas data frame from the data, it was filtered, missing data values were replaced with mean value of the column, and an outcome column was created.
- This is the following Github URL of the complete code:

https://github.com/johnny-zeng-analysis/Coursera-Python-Capstone-Project/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb

#### TASK 4: Create a landing outcome label from Outcome column

Using the Outcome, create a list where the element is zero if the corresponding row in Outcome is in the set bad\_outcome; otherwise, it's one. Then assign it to the variable landing\_class:

```
# Landing_class = 0 if bad_outcome
# Landing_class = 1 otherwise

# Create a set of outcomes where the second stage did not land successfully
bad_outcomes = set(landing_outcomes.keys()[[1, 3, 5, 6, 7]])

# Create a new column 'landing_class' based on conditions
df['landing_class'] = np.where(df['Outcome'].isin(bad_outcomes), 0, 1)
```

This variable will represent the classification variable that represents the outcome of each launch. If the value is zero, the first stage did not land successfully; one means the first stage landed Successfully

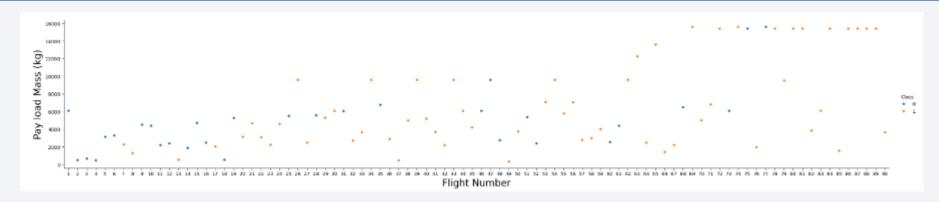
```
#df['Class']=landing_class
#df[['Class']].head(8)

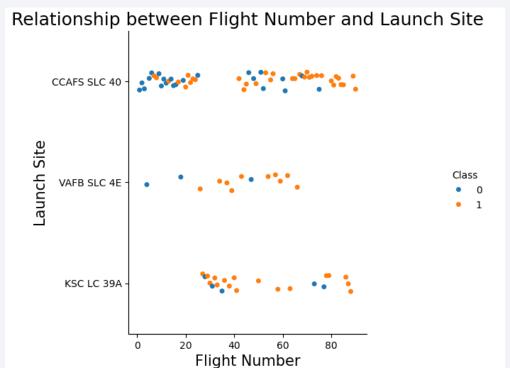
df[['Outcome', 'landing_class']].head(8)
```

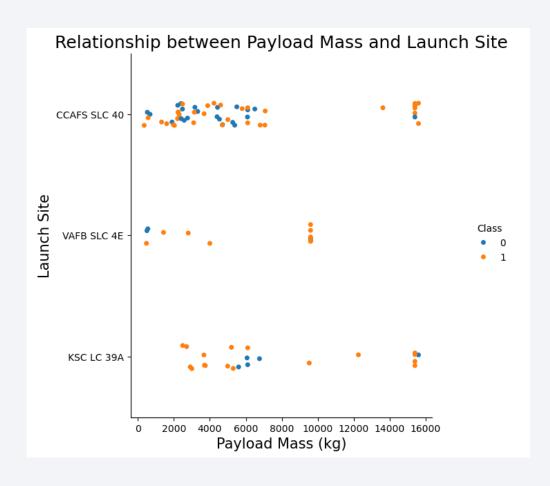
|   | Outcome     | landing_class |
|---|-------------|---------------|
| 0 | None None   | 0             |
| 1 | None None   | 0             |
| 2 | None None   | 0             |
| 3 | False Ocean | 0             |
| 4 | None None   | 0             |
| 5 | None None   | 0             |
| 6 | True Ocean  | 1             |
| 7 | True Ocean  | 1             |
|   |             |               |

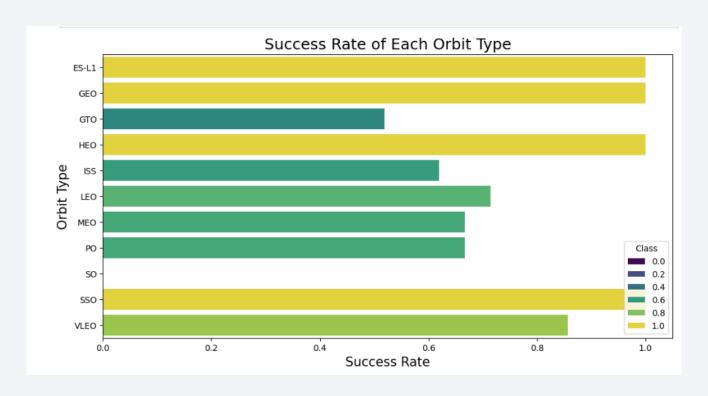
- Data analysis was performed using Pandas and Matplotlib
- Scatter plots, bar charts, and line charts were used to visualize the data
- This is the following Github URL of the complete code:

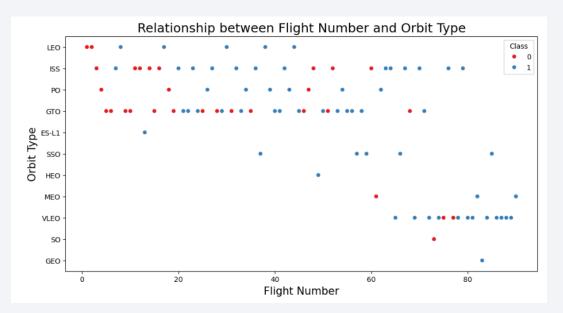
https://github.com/johnny-zeng-analysis/Coursera-Python-Capstone-Project/blob/main/jupyter-labs-eda-dataviz.ipynb.jupyterlite.ipynb

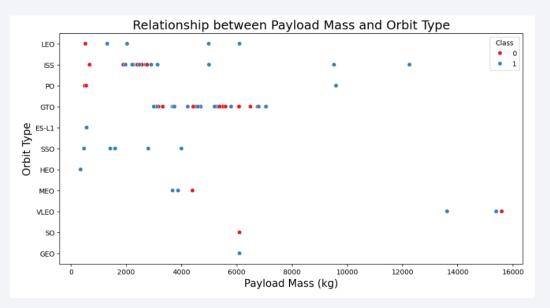


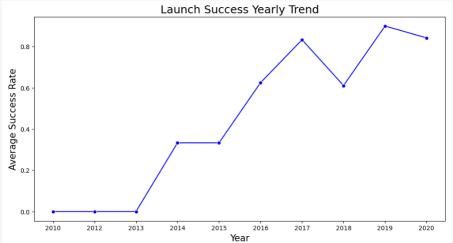












#### EDA with SQL

- SQL was used to perform various queries such as, names of unique launch sites, launch sites beginning with specific strings, total payload mass carried by boosters, average payload mass carried by booster version F9, dates of successful landing outcomes, and many more.
- This is the following Github URL of the complete code:

https://github.com/johnny-zeng-analysis/Coursera-Python-Capstone-Project/blob/main/jupyter-labs-eda-sql-coursera\_sqllite.ipynb

#### Build an Interactive Map with Folium

- Folium map with marked launch sites with markers, circles was created.
- This is the following Github URL of the complete code:

https://github.com/johnny-zeng-analysis/Coursera-Python-Capstone-Project/blob/main/lab jupyter launch site location.jupyterlite.ipynb

#### Build a Dashboard with Plotly Dash

- An interactive dashboard application was created using Plotly dash which includes:
  - Launch Site Drop-down
  - Callback function to render success-pie-chart based on site dropdown selected
  - Range slider to select payload
  - Callback function to render success-payload-scatter-chart scatter plot

This is the following Github URL of the complete code:

https://github.com/johnny-zeng-analysis/Coursera-Python-Capstone-Project/blob/main/labs jupyter dashboard application plotly dash.py

#### Predictive Analysis (Classification)

- 1- Loading the data as a Pandas Dataframe was the first step before performing exploratory Data analysis.
- 2- I then proceeded to create a Numpy array from the column class in data
- 3- Then I standardized the dataset (x) by transforming it using preprocessing. Standard Scaler() from Sklearn
- 4- The data was then split into training and testing sets used for analysis

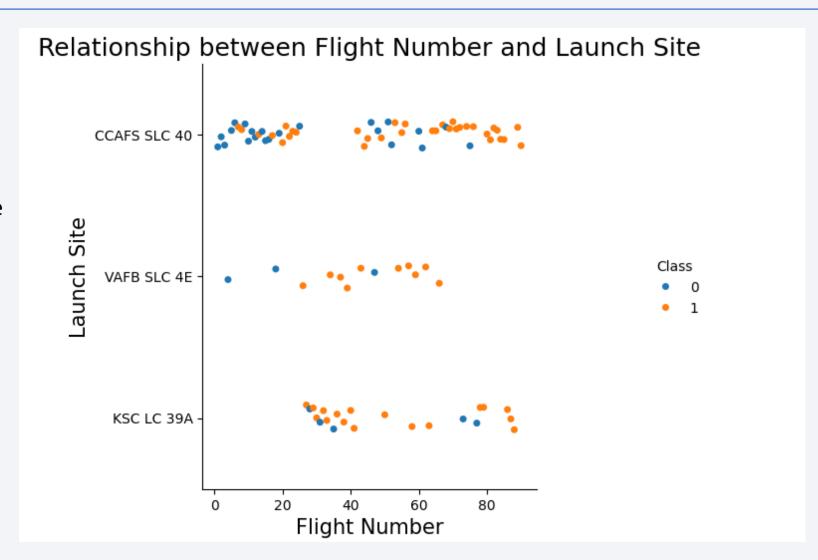
#### Results

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



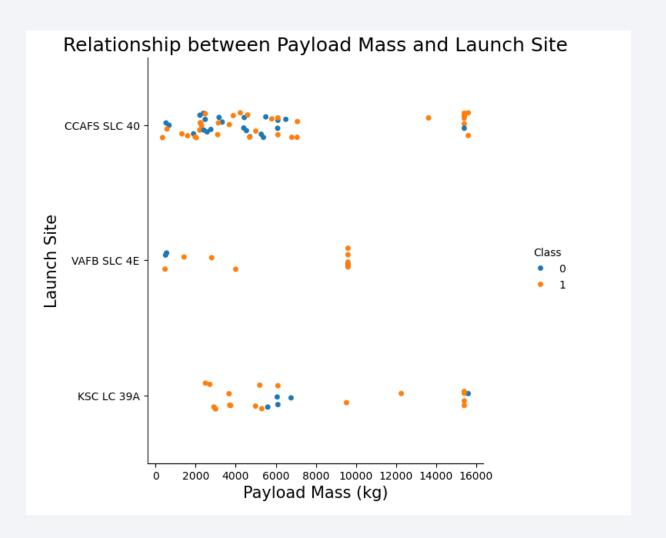
### Flight Number vs. Launch Site

The graphs shows as the flight number increases in each of the launch sites, the success rate also increases



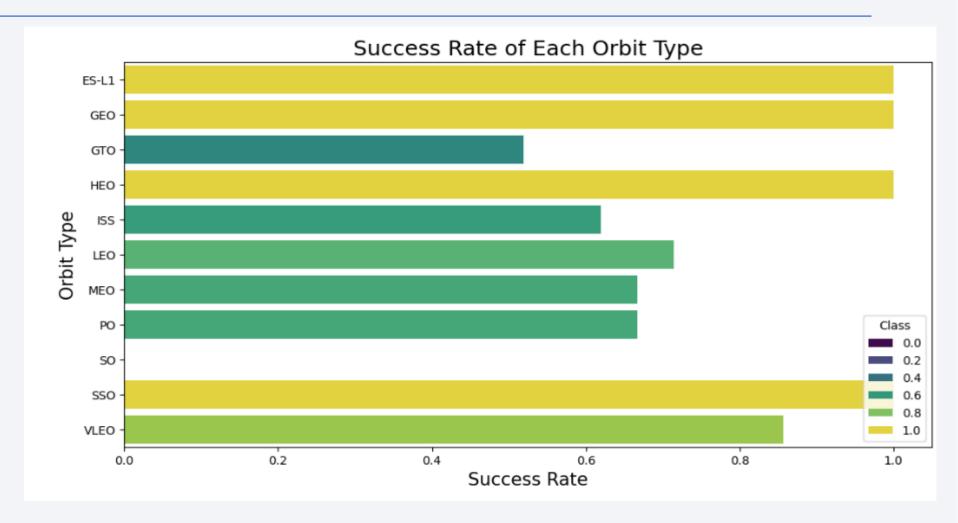
#### Payload vs. Launch Site

 This chart shows that VAFB SLC 4E launchsite has no rockets launched with a payload mass greater than 10,000 kg



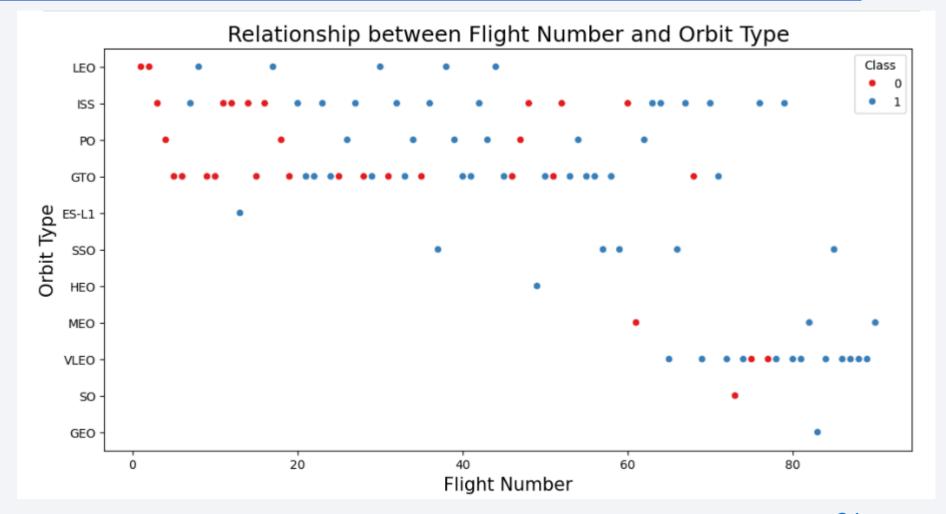
### Success Rate vs. Orbit Type

This chart shows 4
 orbit types has a
 success rate of
 100% with one
 having 0% success
 rate.



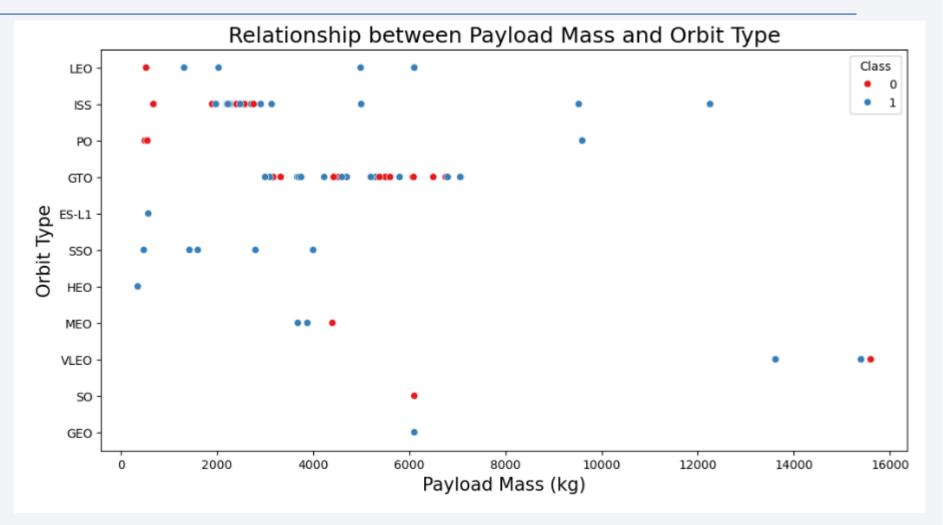
### Flight Number vs. Orbit Type

 This chart shows that LEO has more success with more flights whereas GTO seems to have no pattern between flight numbers and success rate.



### Payload vs. Orbit Type

 This chart shows heavier payload mass have a high chance of successfully landing.



### Launch Success Yearly Trend

• This chart shows after 2013 success rate of launches has gone up drastically every year except for years 2017-2018 and 2019-2020.

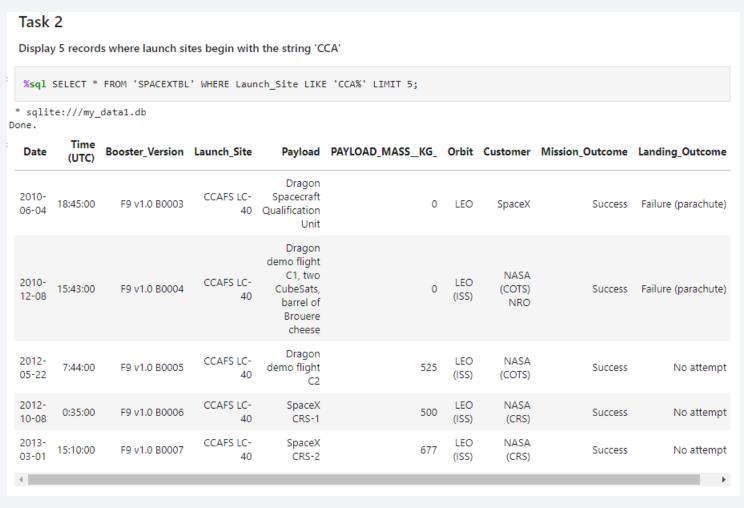


#### All Launch Site Names

- Find the names of the unique launch sites
- Used SELECT DISTINCT to return unique launch sites

## Launch Site Names Begin with 'CCA'

- Find 5 records where launch sites begin with `CCA`
- Used LIKE % to display records with CCA string



### **Total Payload Mass**

```
Task 3

Display the total payload mass carried by boosters launched by NASA (CRS)

**sql SELECT SUM(PAYLOAD_MASS__KG_) as "Total Payload Mass(Kgs)", Customer FROM 'SPACEXTBL' WHERE Customer = 'NASA (CRS)';

* sqlite:///my_datal.db
Done.

**Total Payload Mass(Kgs) Customer

45596 NASA (CRS)
```

- Calculate the total payload carried by boosters from NASA
- Used SUM to return and display total sum of PAYLOAD\_MASS\_\_KG\_ column

### Average Payload Mass by F9 v1.1

## Task 4 Display average payload mass carried by booster version F9 v1.1 \*\*sql SELECT AVG(PAYLOAD\_MASS\_\_KG\_) as "Payload Mass Kgs", Customer, Booster\_Version FROM 'SPACEXTBL' WHERE Booster\_Version I \* sqlite:///my\_datal.db Done. Payload Mass Kgs Customer Booster\_Version 2534.666666666665 MDA F9 v1.1 B1003

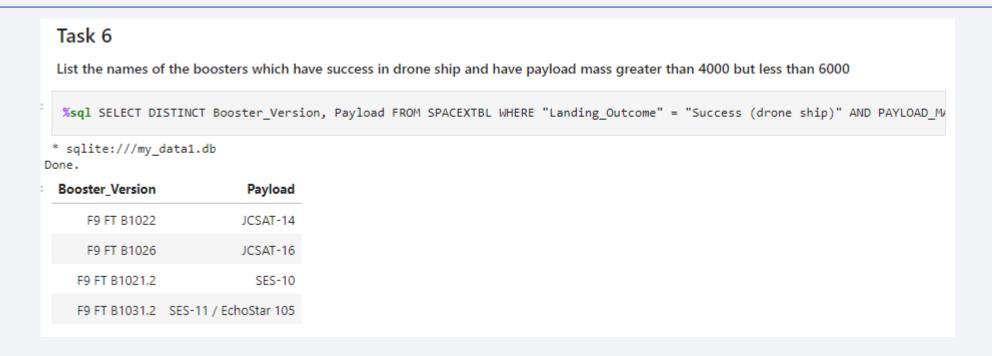
- Calculate the average payload mass carried by booster version F9 v1.1
- Used AVG() to return average payload mass carried by F9 v1.1

### First Successful Ground Landing Date

# 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)"; \* sqlite://my\_data1.db Done. MIN(DATE) 2015-12-22

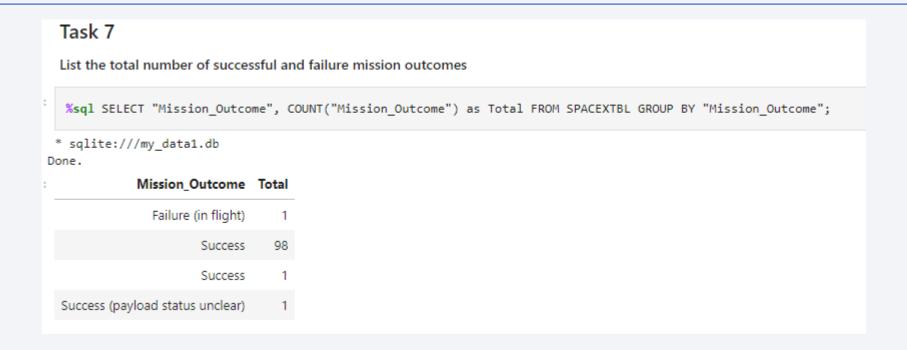
- Find the dates of the first successful landing outcome on ground pad
- Used MIN() to display first date of successful landing on ground pad

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



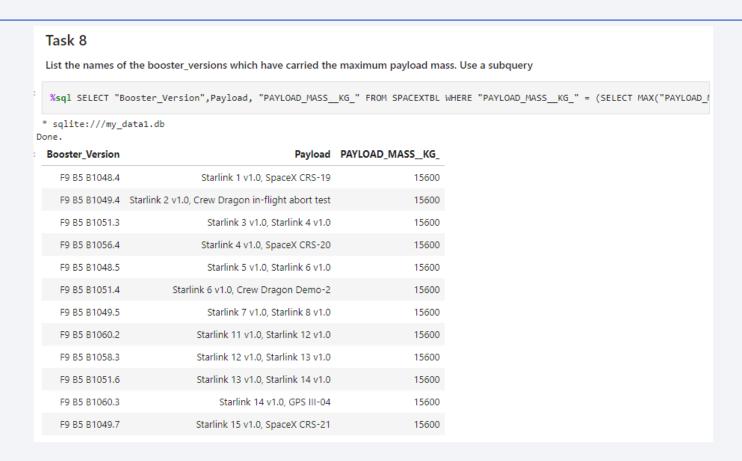
- List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000
- Used Select Distinct to return unique names of boosters between 4000 and 6000 kg mass that were success in drone ship

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



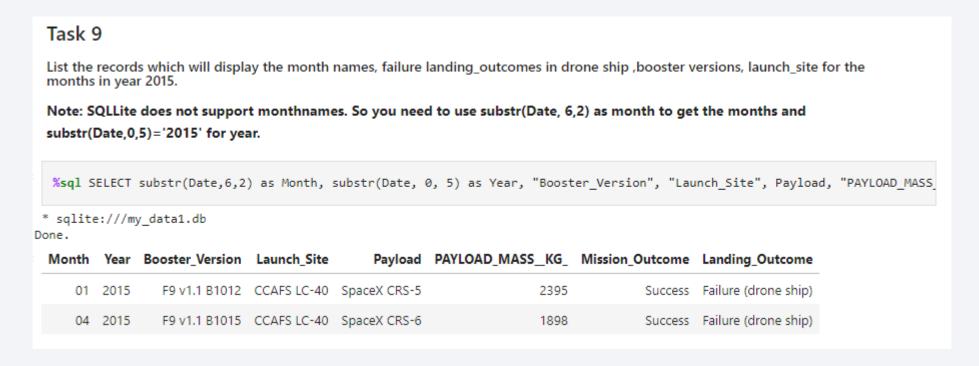
- Calculate the total number of successful and failure mission outcomes
- Used COUNT() with GROUP BY to return total number of missions outcomes

### **Boosters Carried Maximum Payload**



- List the names of the booster which have carried the maximum payload mass
- Used subquery to return Max payload

#### 2015 Launch Records



- List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- Used substr() to extract months in year 2015 where landing\_outcome were failure (drone ship)

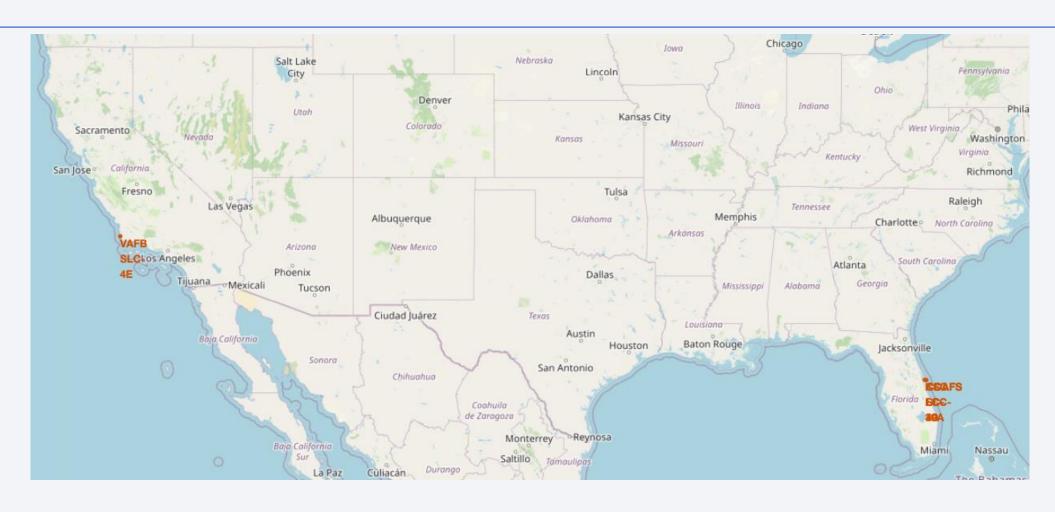
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

 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

#### Task 10 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 \* FROM SPACEXTBL WHERE "Landing Outcome" LIKE 'Success%' AND (Date BETWEEN '2010-06-04' AND '2017-03-20') ORDER \* sqlite:///my\_data1.db Done. Date **Booster Version Launch Site** Payload PAYLOAD MASS KG Customer Mission Outcome Landing Outco (UTC) LEO 2017-SpaceX Success (gro 2490 14:39:00 F9 FT B1031.1 KSC LC-39A NASA (CRS) Success 02-19 CRS-10 (ISS) VAFB SLC-Iridium Iridium Success (dr 2017-Polar 7:54:00 F9 FT B1029.1 9600 Success 01-14 NEXT 1 LEO Communications CCAFS LC-2016-SKY Perfect JSAT Success (dr 5:26:00 F9 FT B1026 GTO JCSAT-16 4600 Success 08-14 40 Group 2016-CCAFS LC-SpaceX LEO Success (gro 4:45:00 F9 FT B1025.1 2257 NASA (CRS) Success 07-18 40 CRS-9 (ISS) CCAFS LC-2016-Success (dr 21:39:00 F9 FT B1023.1 Thaicom 8 3100 GTO Thaicom Success 05-27 40 2016-CCAFS LC-SKY Perfect JSAT Success (dr 5:21:00 JCSAT-14 GTO F9 FT B1022 4696 Success 05-06 40 Group 2016-CCAFS LC-SpaceX LEO Success (dr 20:43:00 F9 FT B1021.1 3136 NASA (CRS) Success 04-08 CRS-8 OG2 Mission 2 CCAFS LC-2015-11 Success (gro 1:29:00 F9 FT B1019 2034 LEO Orbcomm Success 12-22 Orbcomm-OG2 satellites



#### Markers of launch sites



The launch sites are located in the United States as shown by the markers

#### Launch outcomes for each launch site with markers

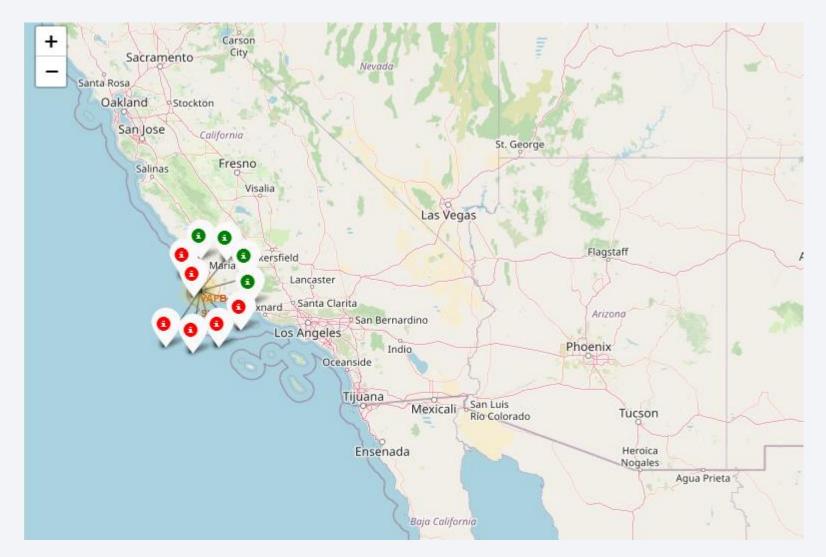




• This shows the success of the launches from each site

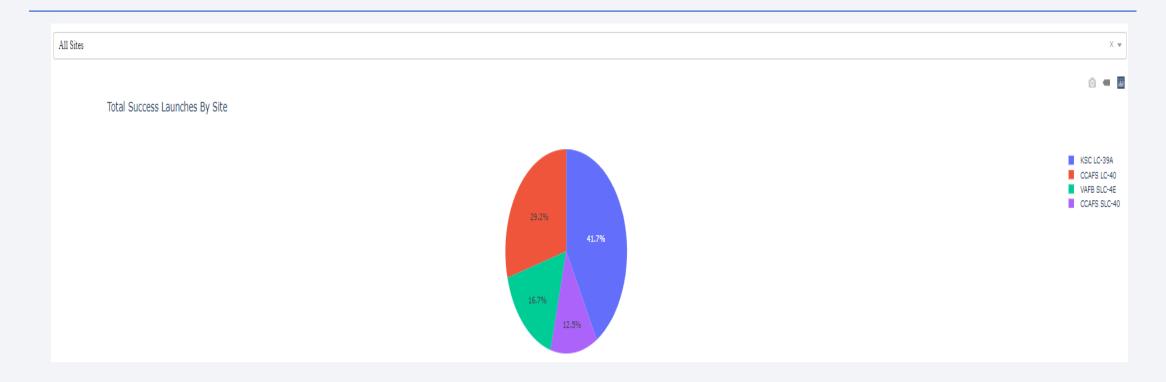
#### Launch outcomes for California site

 This shows the success of the launch outcomes on the California launch site





### Pie Chart Total Successes by Site



• This shows the total successes for all 4 launch sites in a pie chart



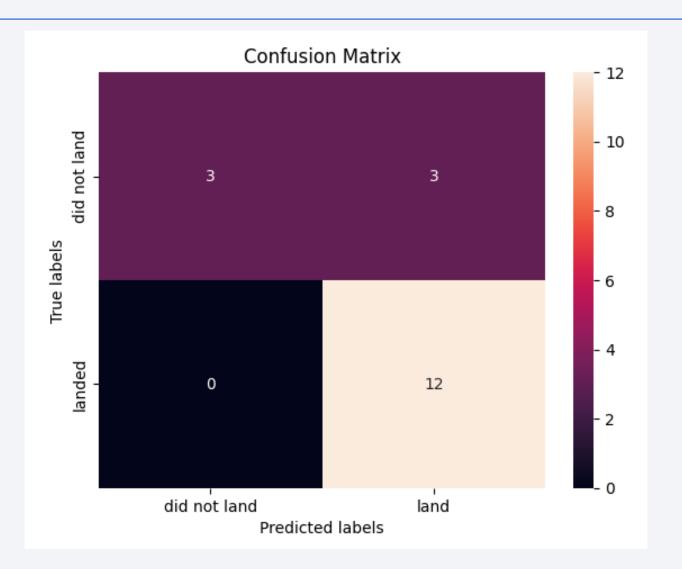
## **Classification Accuracy**

 All methods have the same accuracy of 0.833333

```
Find the method performs best:
Report = pd.DataFrame({'Method' : ['Test Data Accuracy']})
knn_accuracy=knn_cv.score(X_test, Y_test)
Decision_tree_accuracy=tree_cv.score(X_test, Y_test)
SVM_accuracy=svm_cv.score(X_test, Y_test)
Logistic_Regression=logreg_cv.score(X_test, Y_test)
Report['Logistic_Reg'] = [Logistic_Regression]
Report['SVM'] = [SVM accuracy]
Report['Decision Tree'] = [Decision tree accuracy]
Report['KNN'] = [knn accuracy]
Report.transpose()
                            0
    Method Test Data Accuracy
Logistic_Reg
                     0.833333
       SVM
                     0.833333
Decision Tree
                     0.833333
       KNN
                     0.833333
```

#### **Confusion Matrix**

 All 4 of the classification models had the same confusion matrixes



#### Conclusions

- Different launch sites have different success rates
- For 3 of the 4 launch sites, when number of flights increases, so does the success rate of the launches.
- For VAFB-SLC, they do not have launches of payload mass greater than 10000 kg
- GEO, SSO, HEO, and ES-L1 orbits have 100% success rates.
- GTO orbit seems to have no correlation between flight number and success rate

