



IBM Developer  
SKILLS NETWORK

# Winning Space Race with Data Science

<Name>

<Date>



# Outline

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

# Executive Summary

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- Summary of methodologies
  - Data collection methodology
  - Data wrangling
  - Exploratory data analysis (EDA) using visualization and SQL
  - Interactive visual analytics using Folium and Plotly Dash
  - Predictive analysis using classification models
- Summary of all results
  - Exploratory data analysis results
  - Interactive analytics demo in screenshots
  - Predictive analysis results

# Introduction

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- SpaceX, or Space Exploration Technologies Corp., is an American aerospace manufacturer and space transportation company founded by Elon Musk in 2002.
- One of the best innovation ideas in the world is SpaceX creating a reusable rocket technology.
- The Falcon 9 is a two-stage orbital launch vehicle developed and manufactured by SpaceX. The ability to land and reuse the first stage contributes significantly to reducing the cost of space travel.
- The prediction of Falcon 9 Landing using Data Science.





Section 1

# Methodology

# Methodology

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## 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

---

- Describe how data sets were collected.

Downloading from spacex

```
[6]: spacex_url="https://api.spacexdata.com/v4/launches/past"
[7]: response = requests.get(spacex_url)
```

Converting into dataframe

```
[12]: # Use json_normalize meethod to convert the json result into a dataframe
data=pd.json_normalize(response.json())
```

```
[23]: # Create a data from launch_dict
data=pd.DataFrame(launch_dict)
```

Cleansing data

```
[37]: # Hint data['BoosterVersion']!='Falcon 1'
data_falcon9=data[data['BoosterVersion']!='Falcon 1']
data_falcon9.head()
```

# Data Collection – SpaceX API

- Present your data collection with SpaceX REST calls using key phrases and flowcharts
- <https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/jupyter-labs-spacex-data-collection-api.ipynb>

Start

[Initiate SpaceX API Call]

[Specify API Endpoint (e.g., Launches, Rockets, etc.)]

[Include Parameters (e.g., Date Range, Filters)]

[Make HTTP Request]

[Receive API Response]

[Check Status Code]

[Status Code == 200?]

|-----[Yes]

| v

| [Parse JSON Response]

| v

| [Extract Relevant Data]

| v

| [Data Processing/Analysis] ----> [Display or Save Results]

[Error Handling]

[End]



# Data Wrangling

- Describe how data were processed
- Cleansing all the null values and replacing them with the mean value.
- <https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb>
- <https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/jupyter-labs-web scraping.ipynb>

```
[32]: data_falcon9.isnull().sum()
```

```
[32]: FlightNumber    0  
      Date            0  
      BoosterVersion  0  
      PayloadMass     5  
      Orbit           0  
      LaunchSite      0  
      Outcome         0  
      Flights         0  
      GridFins        0  
      Reused          0  
      Legs            0  
      LandingPad      26  
      Block           0  
      ReusedCount     0  
      Serial          0  
      Longitude       0  
      Latitude        0  
      dtype: int64
```

```
[34]: # Calculate the mean value of PayloadMass column  
mean=data_falcon9['PayloadMass'].mean()  
# Replace the np.nan values with its mean value  
data_falcon9.replace(np.nan,mean)
```

# EDA with Data Visualization

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- Summarize what charts were plotted and why you used those charts
- There are various types of charts and graphs, each suitable for different data types and purposes.
- Bar chart: Show the relationship between individual data points and a categorical variable
- Line chart: Display trends over a continuous interval or time series.
- Pie chart: Represent parts of a whole; each slice represents a proportion of the entire dataset.
- Scatter Plot: Display the relationship between two continuous variables, showcasing individual data points.
- <https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/jupyter-labs-webscraping.ipynb>

# EDA with SQL

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- Using bullet point format, summarize the SQL queries you performed
- SQL (Structured Query Language) is commonly used in conjunction with machine learning tasks, especially when dealing with databases and data preprocessing.
- Some of the SQL commands used in the project apart from the basic DDL and DML commands are:
- Functions like Rank(),GroupBy(),OrderBY(),Min(),Max(),Avg(),Sum()
- [https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-sql-coursera\\_sqlite.ipynb](https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-sql-coursera_sqlite.ipynb)

# Build an Interactive Map with Folium

---

- Summarize what map objects such as markers, circles, lines, etc. you created and added to a folium map
- In the context of creating and adding elements to a Folium map, various map objects such as markers, circles, and lines are commonly utilized.
  - Markers are used to pinpoint specific locations on the map.
  - Circles are employed to represent areas of influence or coverage around a central point.
  - Lines are used to depict paths, routes, or connections between two or more points on the map.

(download was not supporting so I added a pdf below)

- <https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/JupyterLite-Folium.pdf>

# Build a Dashboard with Plotly Dash

---

- Summarize what plots/graphs and interactions you have added to a dashboard
- Explain why you added those plots and interactions
- Add the GitHub URL of your completed Plotly Dash lab, as an external reference and peer-review purpose



# Predictive Analysis (Classification)

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- Summarize how you built, evaluated, improved, and found the best performing classification model
- I have calculated and predicted the output using four different algorithms and compared their accuracy.

(download was not supporting so I added a pdf below)

- <https://github.com/gayathridevi1256/Applied-Data-Science-Capstone/blob/main/JupyterLite-Machine%20learning%20prediction.pdf>

# Results

---

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

Landing Outcome	RANK() OVER(ORDER BY COUNT(LANDING_OUTCOME) ASC)
Precluded (drone ship)	1
Failure (parachute)	2
Uncontrolled (ocean)	2
Controlled (ocean)	4
Success (ground pad)	4
Failure (drone ship)	6
Success (drone ship)	6
No attempt	8

```
Best Algorithm is Tree with a score of 0.875
Best Params is : {'criterion': 'entropy', 'max_depth': 2, 'max_features': 'sqrt', 'min_samples_leaf': 2, 'min_samples_split': 2, 'splitter': 'best'}
```



The background of the slide is an abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks in shades of blue and red, creating a sense of motion or data flow. A faint, light blue grid pattern is also visible, particularly in the lower-left quadrant. The overall effect is high-tech and digital.

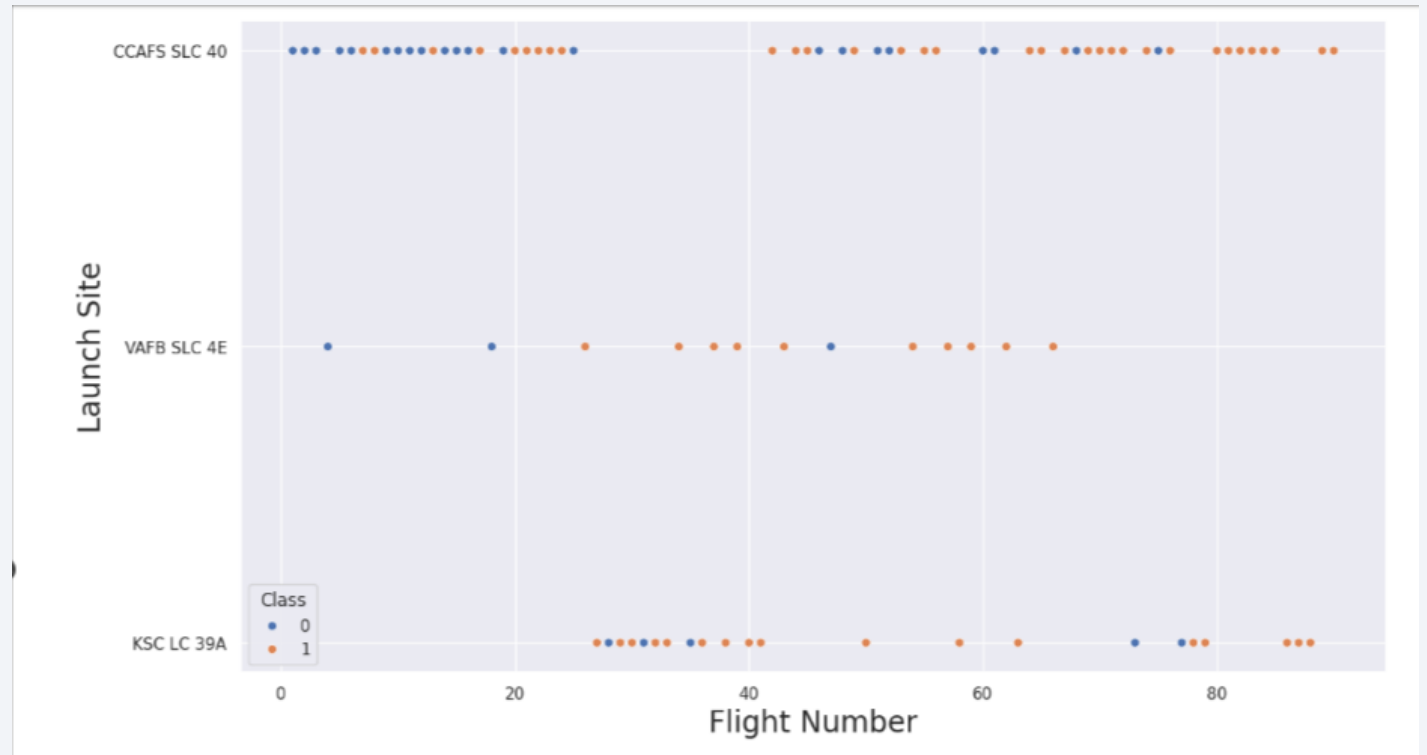
Section 2

# Insights drawn from EDA



# Flight Number vs. Launch Site

- Show a scatter plot of Flight Number vs. Launch Site
- Show the screenshot of the scatter plot with explanations



# Payload vs. Launch Site

- Show a scatter plot of Payload vs. Launch Site
- Show the screenshot of the scatter plot with explanations

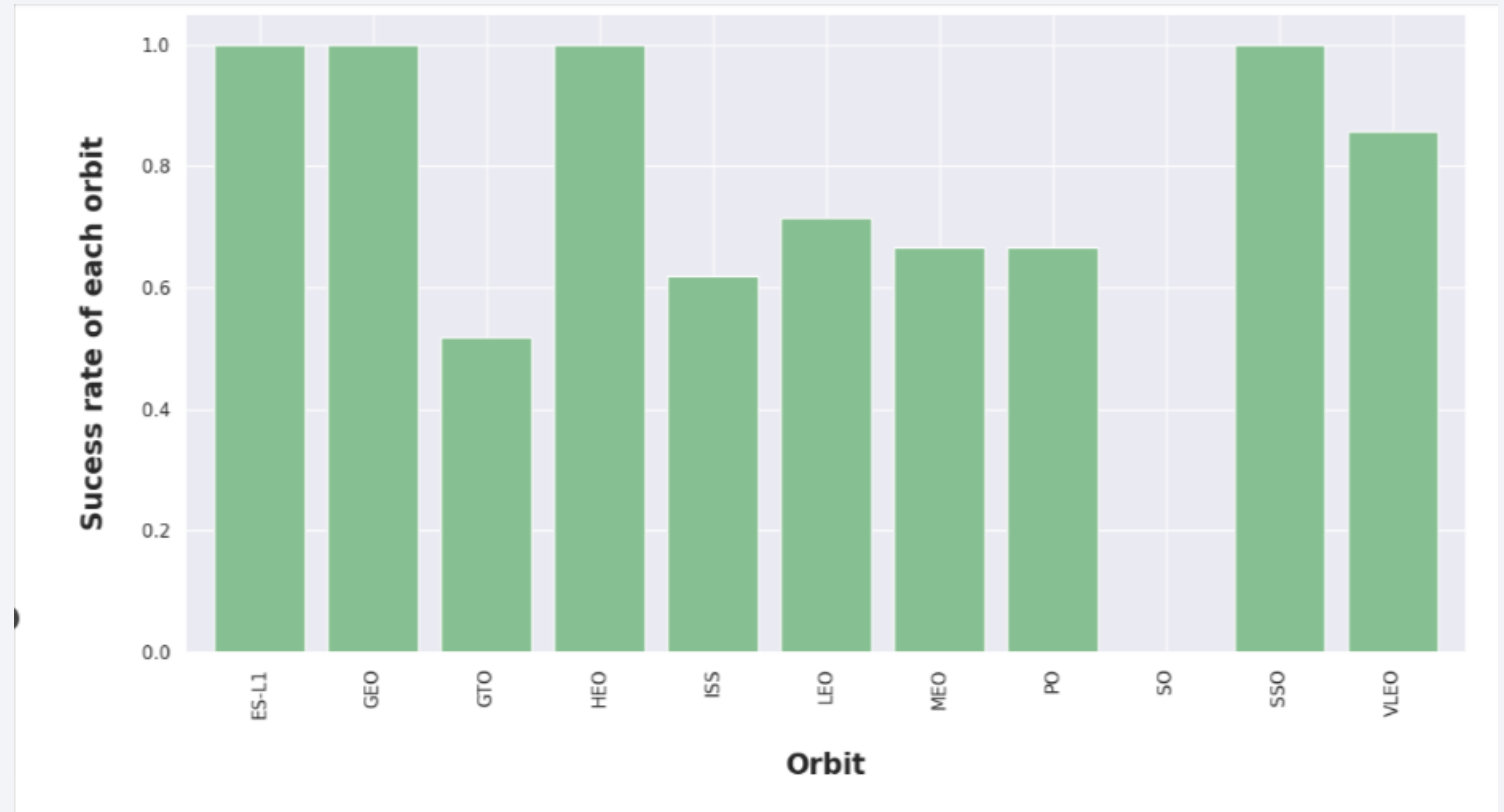




# Success Rate vs. Orbit Type

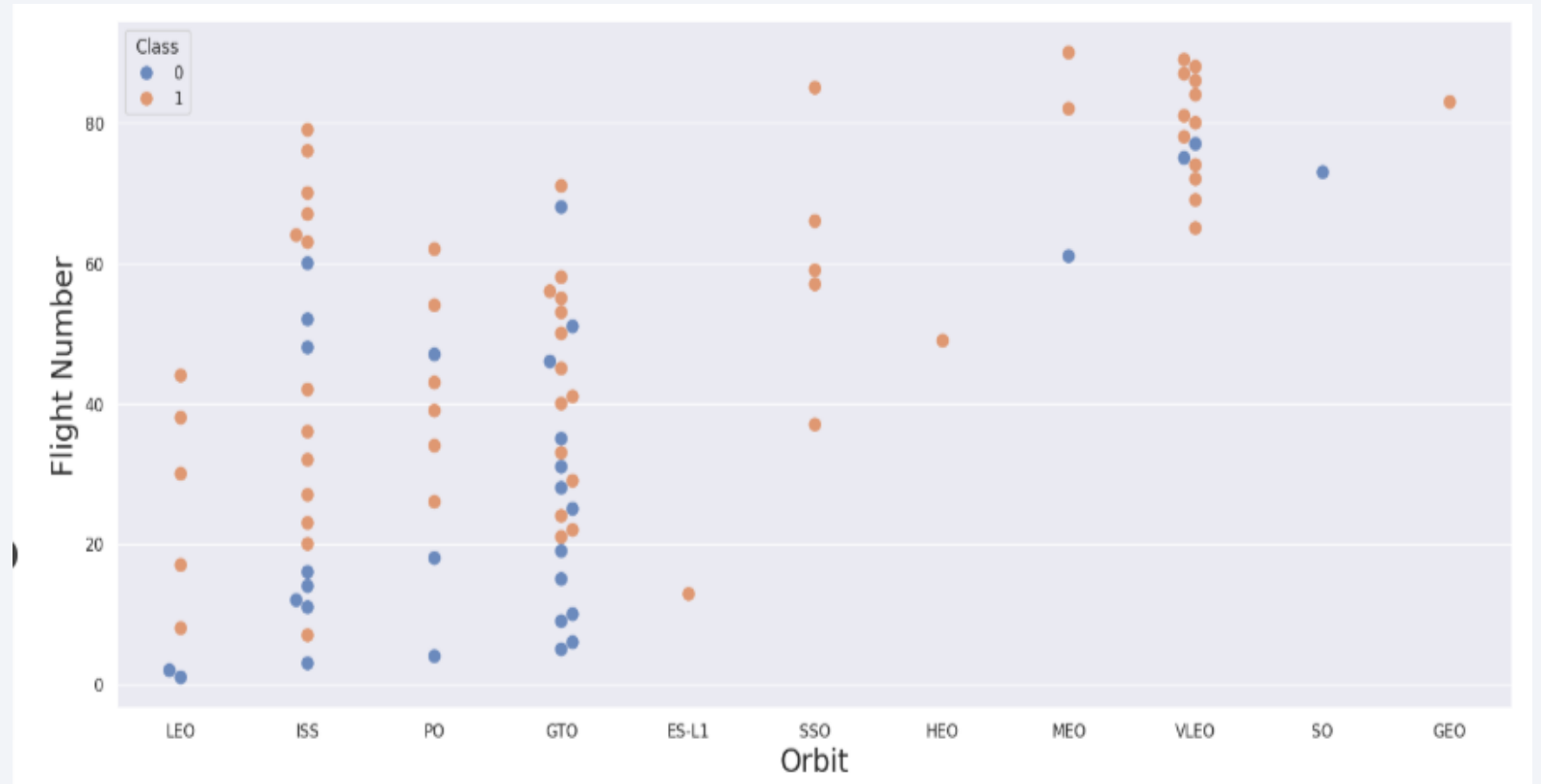
---

- Show a bar chart for the success rate of each orbit type
- Show the screenshot of the scatter plot with explanations



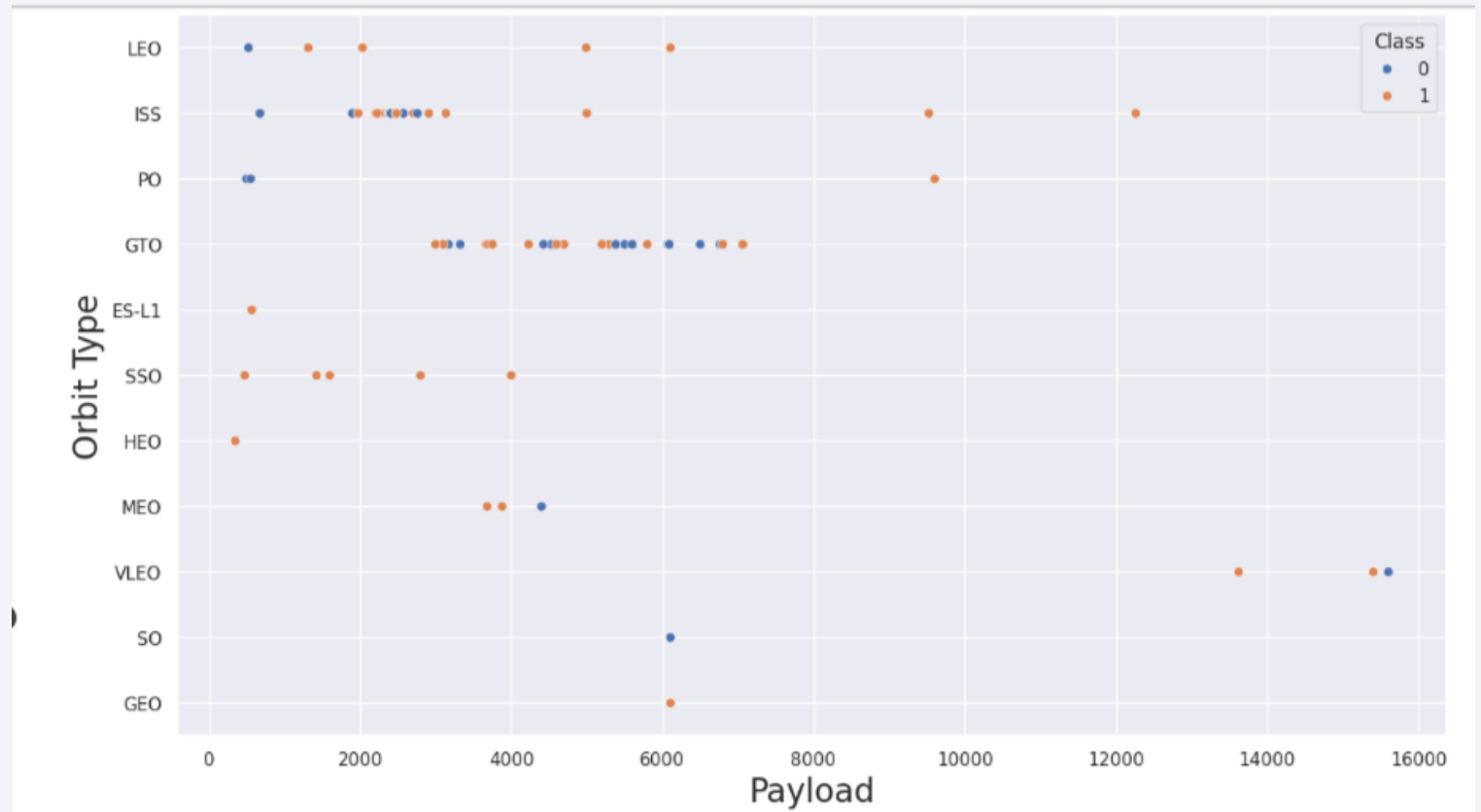
# Flight Number vs. Orbit Type

- Show a scatter point of Flight number vs. Orbit type
- Show the screenshot of the scatter plot with explanations



# Payload vs. Orbit Type

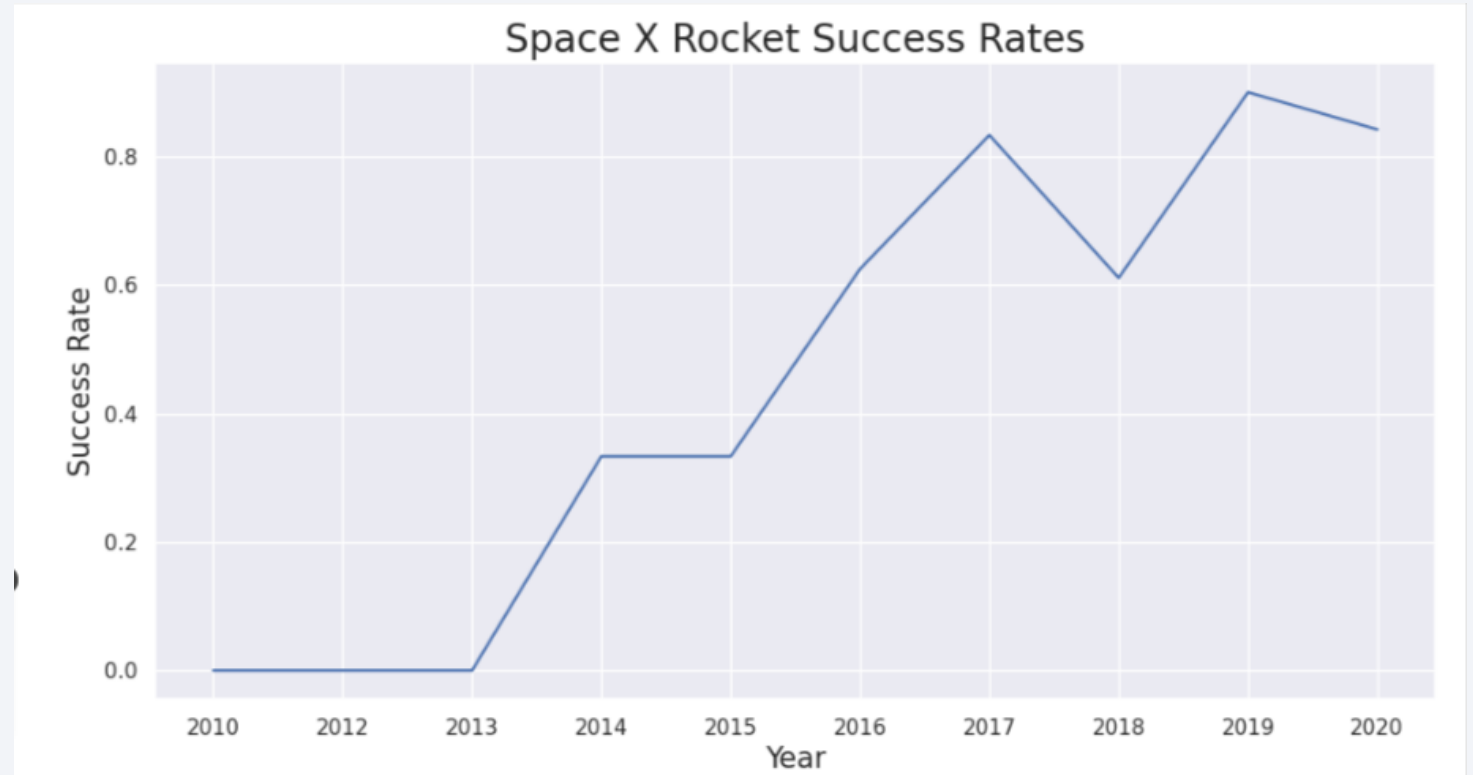
- Show a scatter point of payload vs. orbit type
- Show the screenshot of the scatter plot with explanations



# Launch Success Yearly Trend

---

- Show a line chart of yearly average success rate
- Show the screenshot of the scatter plot with explanations



# All Launch Site Names

---

- Find the names of the unique launch sites
- Present your query result with a short explanation here
- DISTINCT keyword is used to return unique values.

```
[8]: %sql SELECT DISTINCT LAUNCH_SITE AS "Launch_Sites" FROM SPACEXTBL;
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
[8]: Launch_Sites
```

```
CCAFS LC-40
```

```
VAFB SLC-4E
```

```
KSC LC-39A
```

```
CCAFS SLC-40
```



# Launch Site Names Begin with 'CCA'

---

- Find 5 records where launch sites begin with `CCA`
- Present your query result with a short explanation here
- We use LIKE operator to find launch\_site starting with CCA and we use LIMIT to reduce the output to 5 records

```
[10]: %sql SELECT LAUNCH_SITE FROM SPACEXTBL WHERE LAUNCH_SITE LIKE 'CCA%' LIMIT 5;
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
[10]: Launch_Site
```

```
CCAFS LC-40
```

```
CCAFS LC-40
```

```
CCAFS LC-40
```

```
CCAFS LC-40
```

```
CCAFS LC-40
```

# Total Payload Mass

---

- Calculate the total payload carried by boosters from NASA
- Present your query result with a short explanation here
- SUM function is used to calculate the total.

## Task 3

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

```
[30]: %sql SELECT SUM(PAYLOAD_MASS_KG_) FROM SPACEXTBL WHERE Customer = 'NASA (CRS)';
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
[30]: SUM(PAYLOAD_MASS_KG_)
```

```
45596
```

# Average Payload Mass by F9 v1.1

---

- Calculate the average payload mass carried by booster version F9 v1.1
- Present your query result with a short explanation here
- AVG function is used to calculate average.

## Task 4

Display average payload mass carried by booster version F9 v1.1

```
[31]: %sql SELECT AVG(PAYLOAD_MASS_KG_) FROM SPACEXTBL WHERE Booster_Version='F9 v1.1';
```

```
* sqlite:///my_data1.db
```

Done.

```
[31]: AVG(PAYLOAD_MASS_KG_)
```

```
2928.4
```

# First Successful Ground Landing Date

---

- Find the dates of the first successful landing outcome on ground pad
- Present your query result with a short explanation here
- MIN function returns the minimum of values.

## Task 5

List the date when the first succesful landing outcome in ground pad was acheived.

*Hint: Use min function*

```
[34]: %sql SELECT MIN(Date) FROM SPACEXTBL WHERE Landing_Outcome='Success (ground pad)';
```

```
* sqlite:///my_data1.db  
Done.
```

```
[34]: MIN(Date)
```

```
2015-12-22
```

# 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
- Present your query result with a short explanation here

- BETWEEN funtions

gives values in between  
a range.

There are some additional

-l rows in the output.

## Task 6

List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

```
[38]: %sql SELECT Booster_Version FROM SPACEXTBL WHERE PAYLOAD_MASS__KG_ BETWEEN 4000 AND 6000;
```

```
* sqlite:///my_data1.db  
Done.
```

```
[38]: Booster_Version
```

```
F9 v1.1
```

```
F9 v1.1 B1011
```

```
F9 v1.1 B1014
```

```
F9 v1.1 B1016
```

```
F9 FT B1020
```

```
F9 FT B1022
```

```
F9 FT B1026
```

```
F9 FT B1030
```

```
F9 FT B1021.2
```



# Total Number of Successful and Failure Mission Outcomes

---

- Calculate the total number of successful and failure mission outcomes
- Present your query result with a short explanation here
- COUNT function

returns the number of

Items and LIKE function

Compares the similarity.

## Task 7

List the total number of successful and failure mission outcomes

```
[50]: %sql SELECT COUNT(*) AS MISSION_SUCCESS FROM SPACEXTBL WHERE Mission_Outcome LIKE 'Success%';
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
[50]: MISSION_SUCCESS
```

```
100
```

```
[49]: %sql SELECT COUNT(*) AS MISSION_FAIL FROM SPACEXTBL WHERE Mission_Outcome LIKE 'Fail%';
```

```
* sqlite:///my_data1.db
```

```
Done.
```

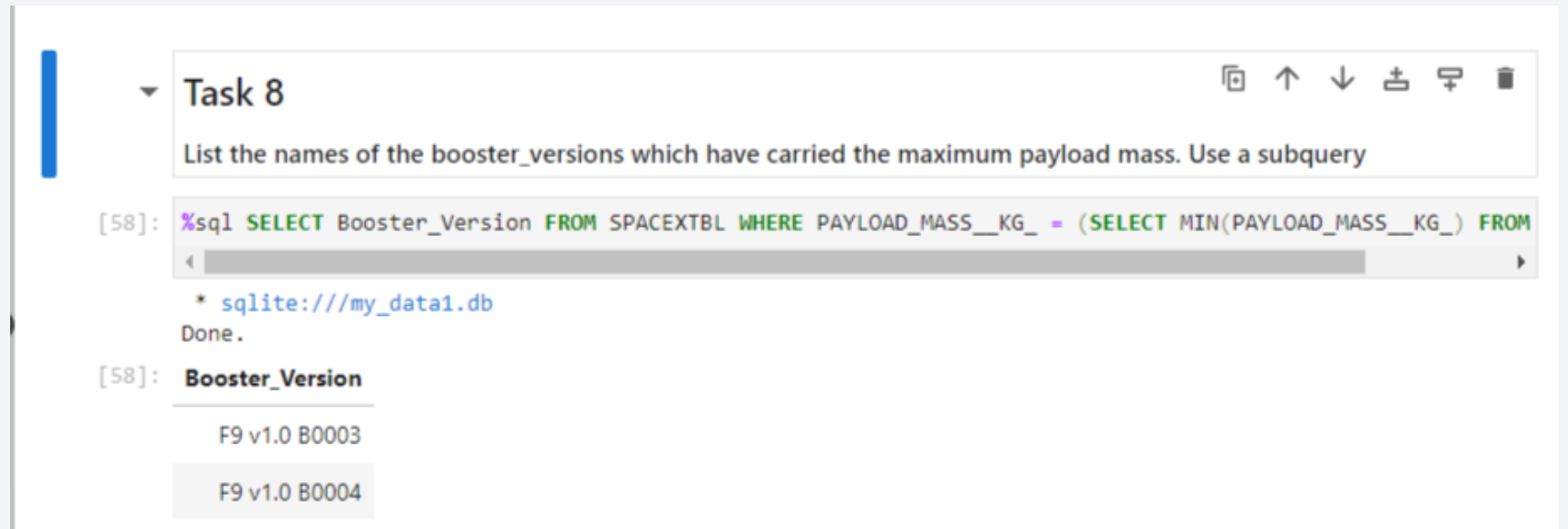
```
[49]: MISSION_FAIL
```

```
1
```

# Boosters Carried Maximum Payload

---

- List the names of the booster which have carried the maximum payload mass
- Present your query result with a short explanation here
- SubQuery is used in this case.



The screenshot shows a Jupyter Notebook interface. At the top, there is a section titled "Task 8" with a description: "List the names of the booster\_versions which have carried the maximum payload mass. Use a subquery". Below this, a code cell [58]: contains a SQL query: `%sql SELECT Booster_Version FROM SPACEXTBL WHERE PAYLOAD_MASS_KG_ = (SELECT MIN(PAYLOAD_MASS_KG_) FROM`. The query is executed, and the output shows the database connection: `* sqlite:///my_data1.db` and `Done.`. Below the output, the results of the query are displayed as a table with one column, **Booster\_Version**. The table contains two rows: `F9 v1.0 B0003` and `F9 v1.0 B0004`.

```
[58]: %sql SELECT Booster_Version FROM SPACEXTBL WHERE PAYLOAD_MASS_KG_ = (SELECT MIN(PAYLOAD_MASS_KG_) FROM
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
[58]: Booster_Version
```

F9 v1.0 B0003
F9 v1.0 B0004

# 2015 Launch Records

---

- List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- Present your query result with a short explanation here
- Substr gives the substring .

## 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: SQLite 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.**

```
[71]: %sql SELECT substr(Date, 6,2),BOOSTER_VERSION,LAUNCH_SITE,Landing_Outcome FROM SPACEXTBL WHERE Landing_Outcome = 'Failure (drone ship)'
```

```
* sqlite:///my_data1.db
```

```
Done.
```

```
[71]:
```

	substr(Date, 6,2)	Booster_Version	Launch_Site	Landing_Outcome
--	-------------------	-----------------	-------------	-----------------

	01	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
--	----	---------------	-------------	----------------------

	04	F9 v1.1 B1015	CCAFS LC-40	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
- Present your query result with a short explanation here
- The **ORDER BY** keyword sorts the records in ascending **order by** default.

## 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.

```
[83]: %sql SELECT LANDING_OUTCOME as "Landing Outcome" , RANK() OVER(ORDER BY COUNT(LANDING_OUTCOME) ASC) FROM
```

```
* sqlite:///my_data1.db  
Done.
```

```
[83]:
```

Landing Outcome	RANK() OVER(ORDER BY COUNT(LANDING_OUTCOME) ASC)
Precluded (drone ship)	1
Failure (parachute)	2
Uncontrolled (ocean)	2
Controlled (ocean)	4
Success (ground pad)	4
Failure (drone ship)	6
Success (drone ship)	6
No attempt	8

A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The image is a composite of a dark blue sky with stars and a view of the Earth's surface from space. The Earth's surface is mostly dark blue, with a thin layer of white clouds. A bright, glowing arc of city lights is visible along the horizon, indicating a coastal area. The text "Section 3" is overlaid on the left side of the image.

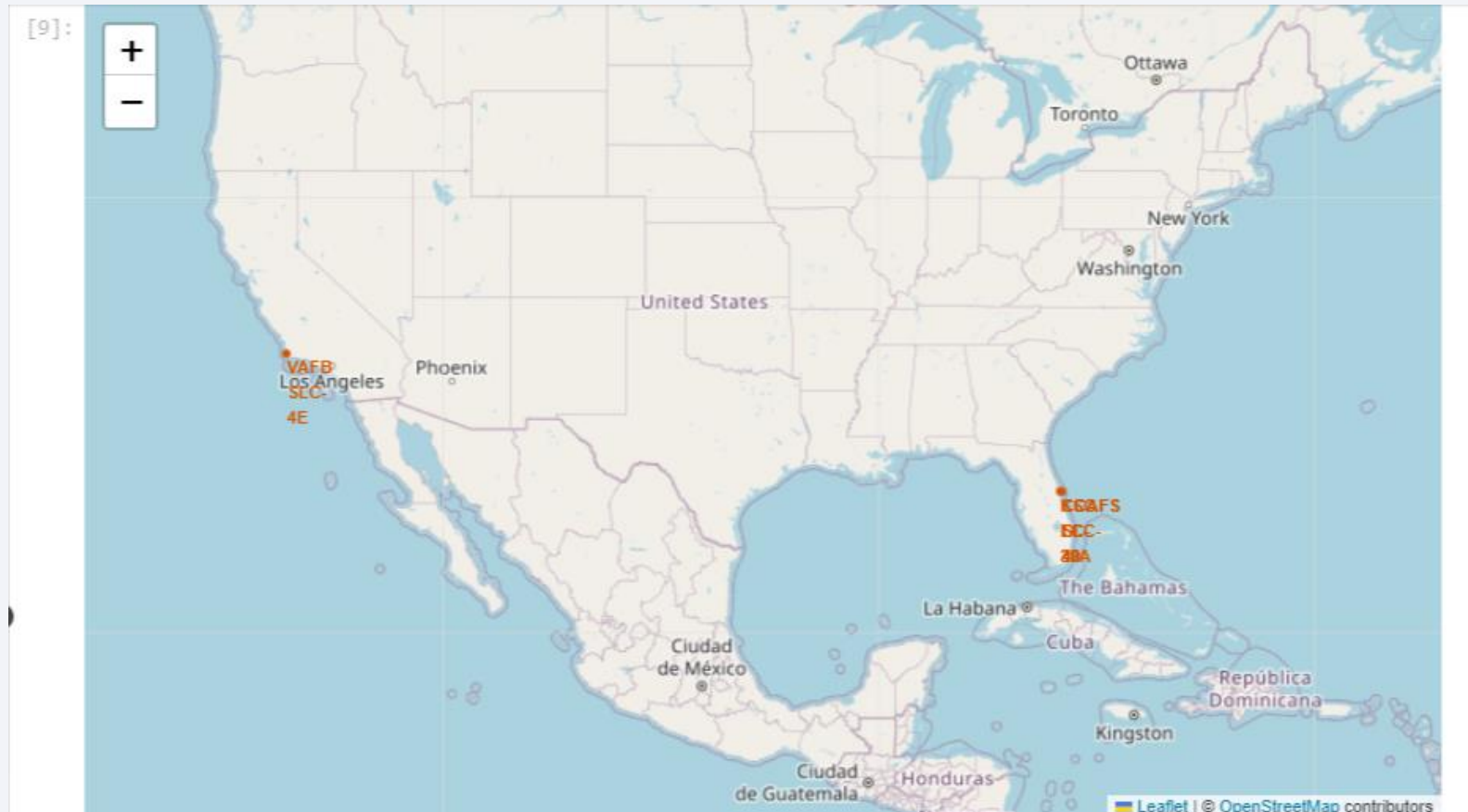
Section 3

# Launch Sites Proximities Analysis

# Folium Map Of Launch Sites Locations

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The generated folium map that includes all launch sites' location markers on a global map

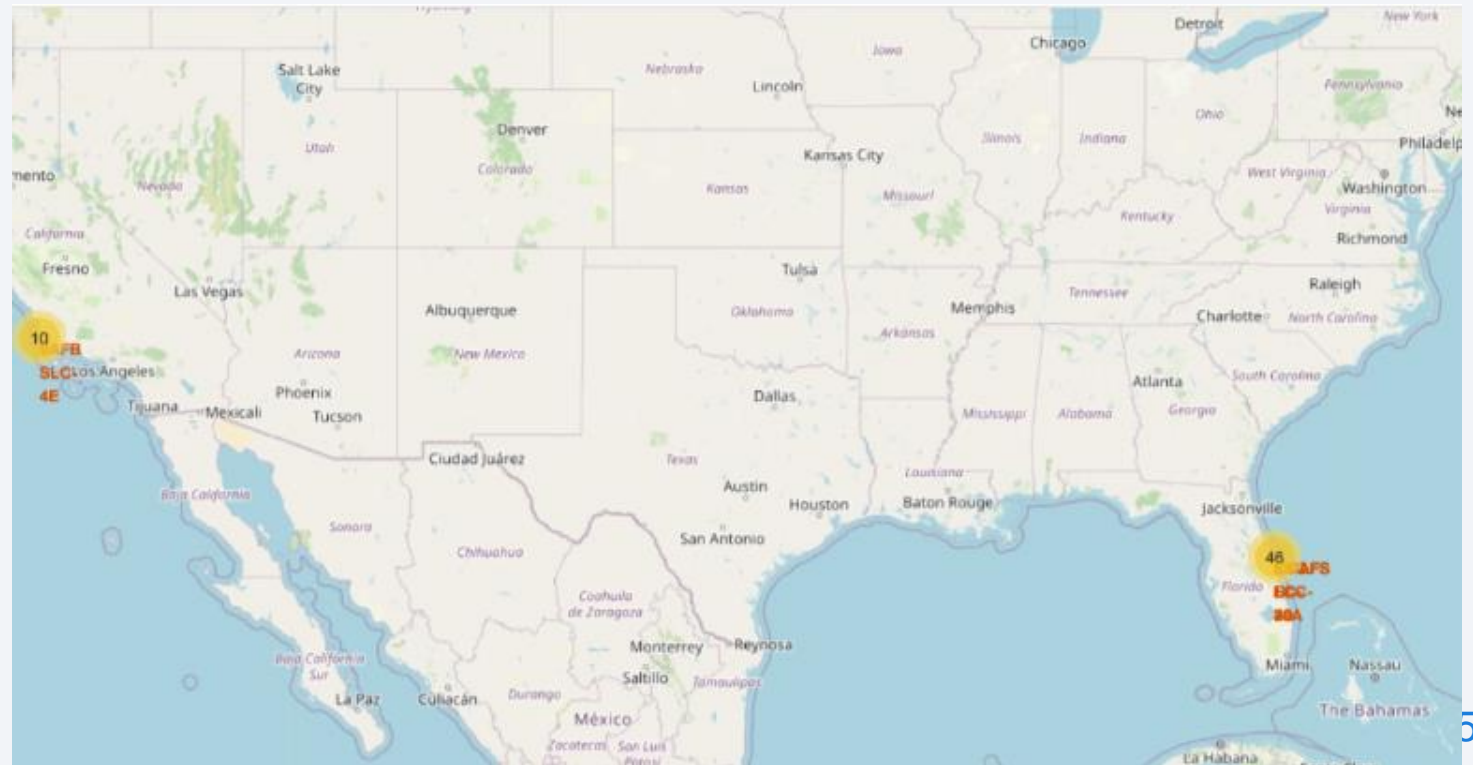




# Folium Map - Color Labeled Launch Outcomes

---

Explore the folium map and make a proper screenshot to show the color-labeled launch outcomes on the map



# Folium Map – Launch Site Proximities

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The generated folium map of a selected launch site to its proximities such as railway, highway, coastline, with distance calculated and displayed







Section 4

# Build a Dashboard with Plotly Dash

# <Dashboard Screenshot 1>

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- Replace <Dashboard screenshot 1> title with an appropriate title
- Show the screenshot of launch success count for all sites, in a piechart
- Explain the important elements and findings on the screenshot

# <Dashboard Screenshot 2>

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- Replace <Dashboard screenshot 2> title with an appropriate title
- Show the screenshot of the piechart for the launch site with highest launch success ratio
- Explain the important elements and findings on the screenshot

## <Dashboard Screenshot 3>

---

- Replace <Dashboard screenshot 3> title with an appropriate title
- Show screenshots of Payload vs. Launch Outcome scatter plot for all sites, with different payload selected in the range slider
- Explain the important elements and findings on the screenshot, such as which payload range or booster version have the largest success rate, etc.



Section 5

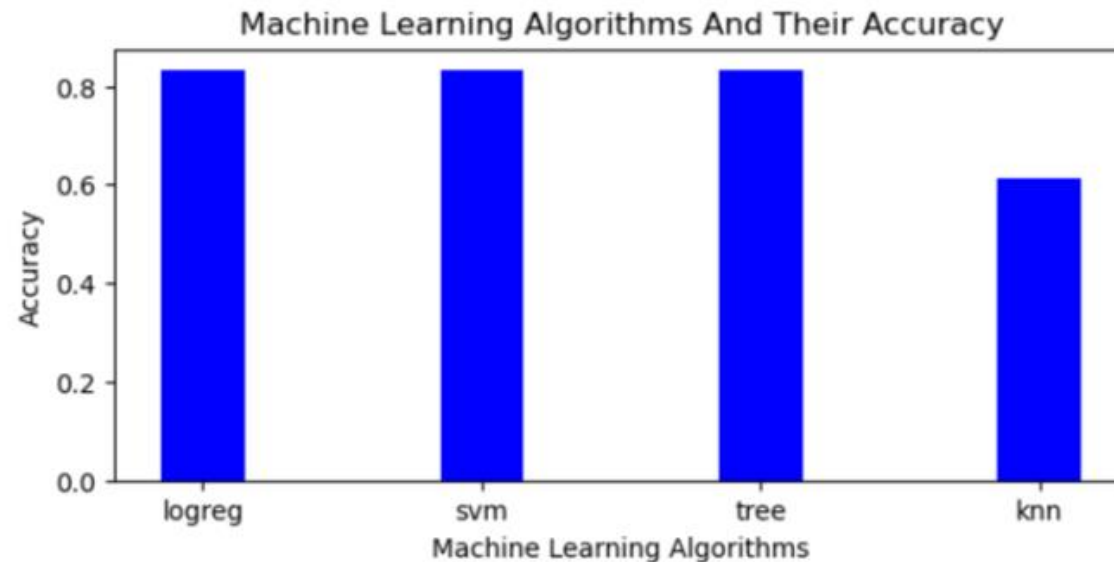
# Predictive Analysis (Classification)



# Classification Accuracy

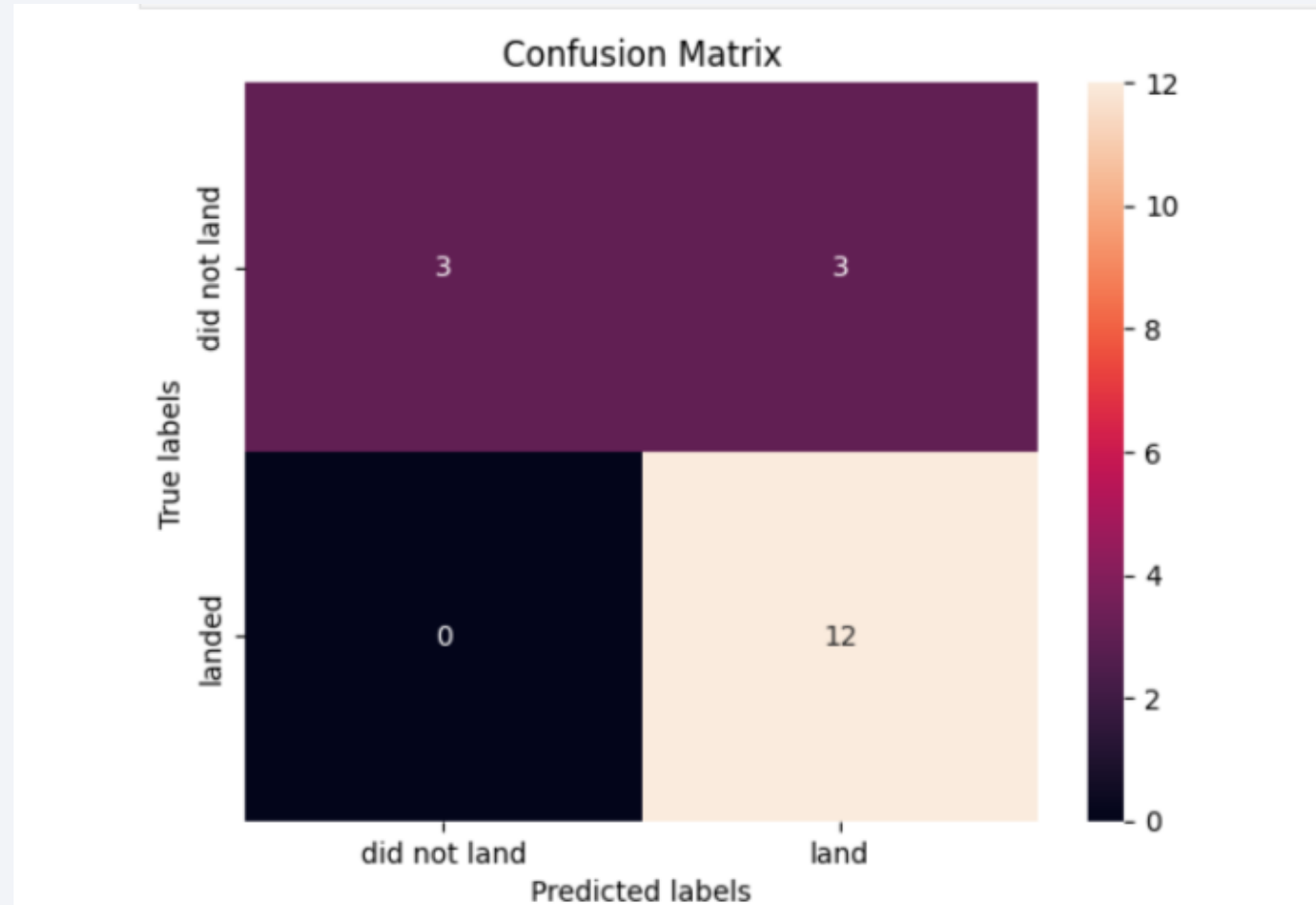
- Visualize the built model accuracy for all built classification models, in a bar chart
- Best Algorithm is Tree with a 0.875.

```
fig = plt.figure(figsize = (7, 5))
plt.bar(['logreg', 'svm', 'tree', 'knn'], [logreg_cv, svm_cv, tree_cv, knn_cv], color = 'blue',
        width = 0.3)
plt.xlabel("Machine Learning Algorithms")
plt.ylabel("Accuracy")
plt.title("Machine Learning Algorithms And Their Accuracy")
plt.show()
```



# Confusion Matrix

- Show the confusion matrix of the best performing model with an explanation



# Conclusions

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- In conclusion, our machine learning project successfully developed a predictive model for Falcon 9 landings, contributing valuable insights into the factors influencing mission success. The model's predictions and accuracy make it a valuable tool for enhancing the efficiency in prediction of Falcon 9 Landing.



# Appendix

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- Include any relevant assets like Python code snippets, SQL queries, charts, Notebook outputs, or data sets that you may have created during this project
- <https://github.com/gayathridevi1256/Applied-Data-Science-Capstone>

Thank you!

