



IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

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Outline

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

In this project, a rival company to SpaceX (i.e., SpaceY) uses SpaceX Falcon 9 rocket data to determine the rocket first stage landing success and the cost of a launch. A summary for the methodologies and results described in this report is outlined below.

Summary of methodologies

- Data Collection
- Data wrangling
- Exploratory data analysis with data visualization and SQL
- Building an interactive map with Folium
- Building Dashboard with Plotly Dash
- Predictive analysis (classification)

Summary of all results

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



GitHub URL: <https://github.com/ocean2024/Ocean2024/tree/main>

Introduction

This capstone project is part of the IBM Data Science Professional Certificate. The goal of the project is to demonstrate proficiency in data science and machine learning techniques using a real-world data and to summarize the results in a report.

In this project, a rival company to SpaceX (i.e., SpaceY) uses SpaceX Falcon 9 rocket data to **determine the rocket first stage landing success** and also uses the rocket data to **determine the cost of a launch**. Space Y uses the data to bid against SpaceX for a rocket launch. SpaceX advertises Falcon 9 rocket launch cost to be 62 million dollars; whereas other companies rocket launch is more than 165 million dollars.

Throughout the project Python Jupyter note books are used to perform the data collection and analysis. These Jupyter note books and the final pdf report are saved in my GitHub repository webpage.

The major parts of this report include data collection methodology, data wrangling, exploratory data analysis (EDA), interactive data visualization, machine learning (ML) classification model development, and model evaluation. Finally, the accuracy of different ML algorithms are compared in predicting the future landing of the Falcon 9 first stage rocket.



Section 1

Methodology

FALCON 9

OVERVIEW

HEIGHT	70 m / 229.6 ft
DIAMETER	3.7 m / 12 ft
MASS	549,054 kg / 1,207,920 lb
PAYLOAD TO LEO	22,800 kg / 50,265 lb
PAYLOAD TO GTO	8,300 kg / 18,300 lb
PAYLOAD TO MARS	4,020 kg / 8,860 lb



Methodology

Data used in this project was collected from SpaceX Rest API and from Wikipedia launch table. The wrangling of the collected data included cleaning, preparation for visualization and information extraction for usage in ML predictive models such as logistic regression, support vector machine (SVM), decision tree, and K-nearest neighbors (KNN).

In addition, exploratory data analysis (EDA) was performed using visualization and SQL. Lastly, Folium and Plotly Dash Python libraries were use in data representation and in the interactive visual analytics of the data.

- Finally, predictive analysis was performed using classification models for predicting whether first stage of Falcon 9 rocket will land successfully using Skikit-learn and the accuracy of the model was determined.

Data Collection: Overview

Data collection and visualization major steps:

Step # 1: Collect Data from SpaceX API and Convert data to .json file

Step# 2: Scrap and filter data to include Falcon 9 data, assign data to dataframe and dictionary, and export data to a csv file

Step 3: Plot and visualize the data

Portion of generated output data file: dataset_part1.csv

A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
FlightNum	Date	BoosterVer	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	LandingPad	Block	ReusedCore	Serial	Longitude	Latitude
1	6/4/2010	Falcon 9	6123.548	LEO	CCSFS SLC	None Non	1	FALSE	FALSE	FALSE		1	0	B0003	-80.5774	28.56186
2	5/22/2012	Falcon 9	525	LEO	CCSFS SLC	None Non	1	FALSE	FALSE	FALSE		1	0	B0005	-80.5774	28.56186
3	3/1/2013	Falcon 9	677	ISS	CCSFS SLC	None Non	1	FALSE	FALSE	FALSE		1	0	B0007	-80.5774	28.56186
4	9/29/2013	Falcon 9	500	PO	VAFB SLC	False Ocean	1	FALSE	FALSE	FALSE		1	0	B1003	-120.611	34.63209
5	12/3/2013	Falcon 9	3170	GTO	CCSFS SLC	None Non	1	FALSE	FALSE	FALSE		1	0	B1004	-80.5774	28.56186
6	1/6/2014	Falcon 9	3325	GTO	CCSFS SLC	None Non	1	FALSE	FALSE	FALSE		1	0	B1005	-80.5774	28.56186
7	4/18/2014	Falcon 9	2296	ISS	CCSFS SLC	True Ocean	1	FALSE	FALSE	TRUE		1	0	B1006	-80.5774	28.56186
8	7/14/2014	Falcon 9	1316	LEO	CCSFS SLC	True Ocean	1	FALSE	FALSE	TRUE		1	0	B1007	-80.5774	28.56186
9	8/5/2014	Falcon 9	4535	GTO	CCSFS SLC	None Non	1	FALSE	FALSE	FALSE		1	0	B1008	-80.5774	28.56186
10	9/7/2014	Falcon 9	4428	GTO	CCSFS SLC	None Non	1	FALSE	FALSE	FALSE		1	0	B1011	-80.5774	28.56186

GitHub URL: https://github.com/ocean2024/Ocean2024/blob/main/dataset_part_1.csv

Data Collection: SpaceX API

Request response from SpaceX API using get request and convert data to .json file

Use custom functions to clean data

Clean data and assign data to dictionary and data frame

Filter data to include only Falcon 9 launches and export data to a csv file: **dataset_part1**

```
] 1 static_json_url='https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/datasets/API_call_spacex_api.json'

] 1 # convert the json result into a dataframe using json_normalize method
  2 data=pd.json_normalize(response.json())

1 # Lets take a subset of our dataframe keeping only the features we want and the flight number, and date_utc.
2 data = data[['rocket', 'payloads', 'launchpad', 'cores', 'flight_number', 'date_utc']]
3 #
4 # We will remove rows with multiple cores because those are falcon rockets with 2 extra rocket boosters and rows that have multiple payloads in a single rocket.
5 data = data[data['cores'].map(len)==1]
6 data = data[data['payloads'].map(len)==1]
7
8 # Since payloads and cores are lists of size 1 we will also extract the single value in the list and replace the feature.
9 data['cores'] = data['cores'].map(lambda x : x[0])
10 data['payloads'] = data['payloads'].map(lambda x : x[0])
11
12 # We also want to convert the date_utc to a datetime datatype and then extracting the date leaving the time
13 data['date'] = pd.to_datetime(data['date_utc']).dt.date
14
15 # Using the date we will restrict the dates of the launches
16 data = data[data['date'] <= datetime.date(2020, 11, 13)]
```

Portion of output

```
[ ] 1 data_falcon9.head()
```

	FlightNumber	Date	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	LandingPad	Block	ReusedCount	Serial	Longitude	Latitude
4	1	2010-06-04	Falcon 9	6123.547647	LEO	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B0003	-80.577366	28.561857
5	2	2012-05-22	Falcon 9	525.000000	LEO	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B0005	-80.577366	28.561857
6	3	2013-03-01	Falcon 9	677.000000	ISS	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B0007	-80.577366	28.561857
7	4	2013-09-29	Falcon 9	500.000000	PO	VAFB SLC 4E	False Ocean	1	False	False	False	None	1.0	0	B1003	-120.610829	34.632093
8	5	2013-12-03	Falcon 9	3170.000000	GTO	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B1004	-80.577366	28.561857

GitHub URL:

https://github.com/ocean2024/Ocean2024/blob/main/IBM%20DS_Capstone%20Project_Data%20Collection.ipynb

Data Collection: Scraping

Step 1: Perform HTTP get to request Falcon 9 HTML page and create BeautifulSoup object from HTML

Step 2: Extract all column/variable names from the HTML table header

Step3: Create a data frame by parsing the launch HTML tables

Step 4: export data into CSV file (spacex_web_scraped.csv)

GitHub URL:

https://github.com/ocean2024/Ocean2024/blob/main/IBM_DS%20Capstone_Project_Web_Data_Scraping_.ipynb

```
[5] 1 # use requests.get() method with the provided static_url
    2 # assign the response to a object
    3 response = requests.get(static_url)
```

Create a BeautifulSoup object from the HTML response

```
[6] 1 # Use BeautifulSoup() to create a BeautifulSoup object from a response text content
    2 soup = BeautifulSoup(response.content, 'html.parser')
```

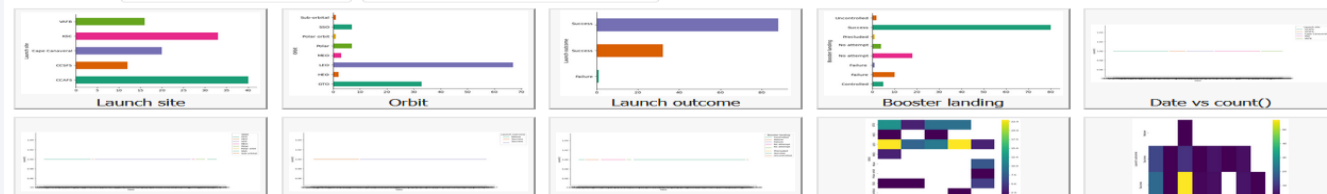
Print the page title to verify if the BeautifulSoup object was created properly

```
[7] 1 # Use soup.title attribute
    2 soup.title
```

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

	Flight No.	Launch site	Payload	Payload mass	Orbit	Customer	Launch outcome	Version Booster	landing	Date	Time
0	1	CCAFS	Dragon Spacecraft Qualification Unit	0	LEO	[[SpaceX], \n	Success\n	F9 v1.0B0003.1	Failure	4 June 2010	18:45
1	2	CCAFS	Dragon	0	LEO	[[.mw-parser-output .plainlist ol,.mw-parser-o...	Success	F9 v1.0B0004.1	Failure	8 December 2010	15:43
2	3	CCAFS	Dragon	525 kg	LEO	[[NASA], (, [COTS], \n	Success	F9 v1.0B0005.1	No attempt\n	22 May 2012	07:44
3	4	CCAFS	SpaceX CRS-1	4,700 kg	LEO	[[NASA], (, [CRS], \n	Success\n	F9 v1.0B0006.1	No attempt	8 October 2012	00:35
4	5	CCAFS	SpaceX CRS-2	4,877 kg	LEO	[[NASA], (, [CRS], \n	Success\n	F9 v1.0B0007.1	No attempt\n	1 March 2013	15:10

Next steps: [Generate code with df](#) [View recommended plots](#)



Portion of Output

Data Collection: Data Wrangling

Step 1: Load data from dataset_part1.csv file and calculate the number of launches on each site

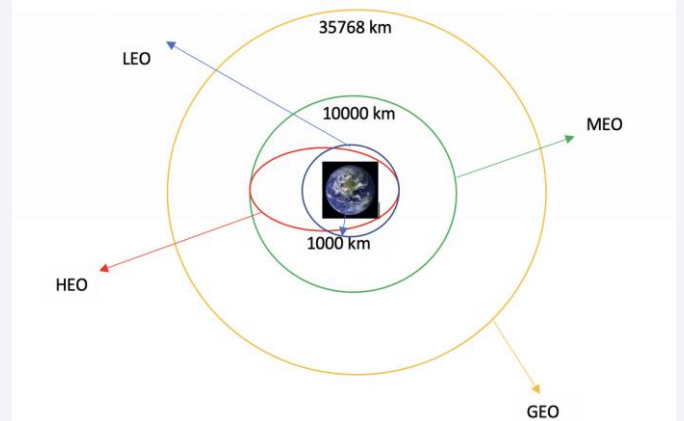
Step 2: Calculate the number and the occurrence of each orbit

Step 3: Calculate the number and occurrence of mission outcome of the orbits

Step 4: Create a landing outcome label from outcome column and export data into **dataset_part2.csv** file

```
[5] 1 # Apply value_counts() on column LaunchSite
    2 df['LaunchSite'].value_counts()
```

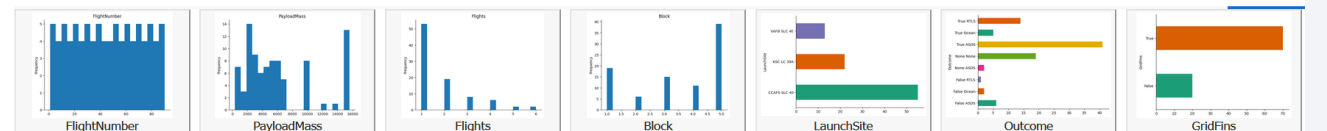
```
LaunchSite
CCAFS SLC 40    55
KSC LC 39A      22
VAFB SLC 4E     13
Name: count, dtype: int64
```



```
[12] 1 df.head(5)
```

	FlightNumber	Date	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	LandingPad	Block	ReusedCount	Serial	Longitude	Latitude
0	1	2010-06-04	Falcon 9	6104.959412	LEO	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B0003	-80.577366	28.561857
1	2	2012-05-22	Falcon 9	525.000000	LEO	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B0005	-80.577366	28.561857
2	3	2013-03-01	Falcon 9	677.000000	ISS	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B0007	-80.577366	28.561857
3	4	2013-09-29	Falcon 9	500.000000	PO	VAFB SLC 4E	False Ocean	1	False	False	False	NaN	1.0	0	B1003	-120.610829	34.632093
4	5	2013-12-03	Falcon 9	3170.000000	GTO	CCAFS SLC 40	None None	1	False	False	False	NaN	1.0	0	B1004	-80.577366	28.561857

Next steps: [Generate code with df](#) [View recommended plots](#)

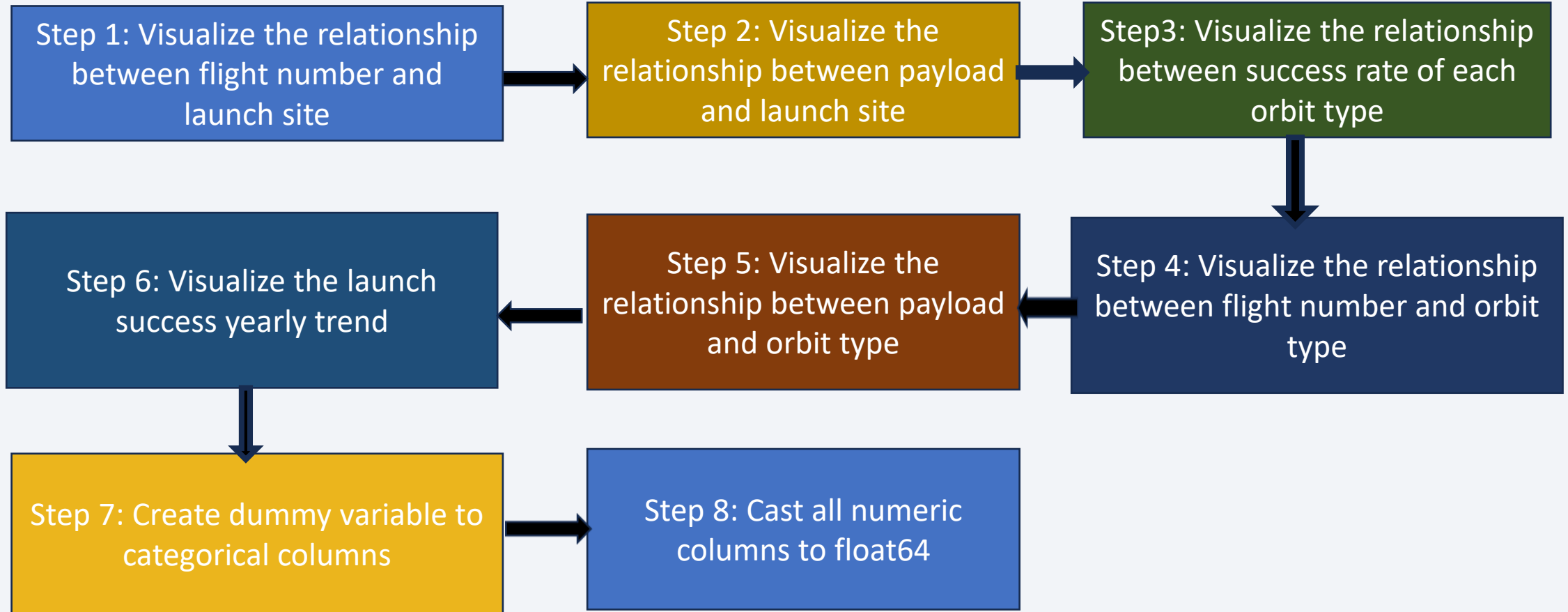


GitHub URL:

https://github.com/ocean2024/Ocean2024/blob/main/IBM_DS_Capstone_Project_Data_Wrangling.ipynb

EDA with Data Visualization

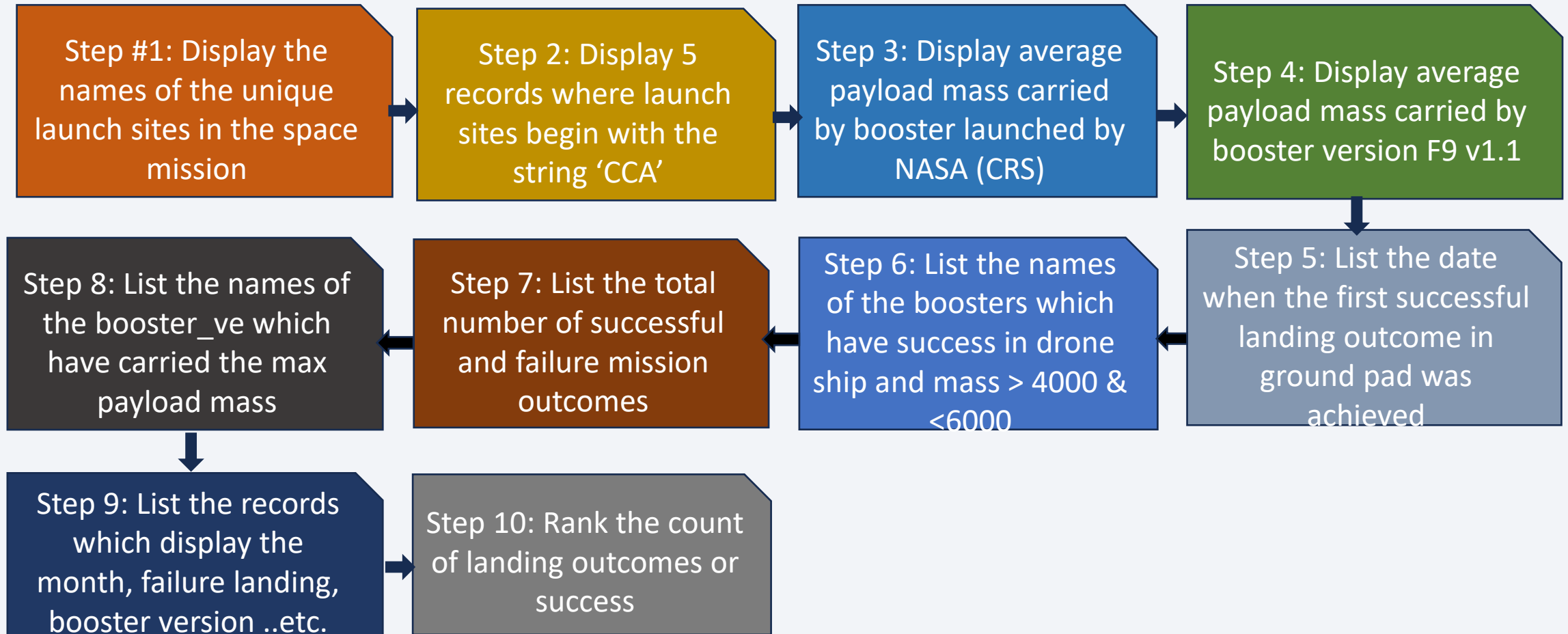
Use Matplotlib and Seaborn for data visualization



GitHub URL:

https://github.com/ocean2024/Ocean2024/blob/main/IBM_DS_Capstone_Project_Data%20Visualization_with_EDA.ipynb

EDA with SQL



GitHub URL:

https://github.com/ocean2024/Ocean2024/blob/main/IBM_DAS_Capstone_Project_SQL.ipynb

Build an Interactive Map with Folium

Step 1: Mark all launch sites on a map created using Folium by adding markers* with circle, popup label and text label to each site using its longitude and latitude coordinates to show the geographical location approximately to the equator

Step 2: Mark the success/failed launches for each site on the map using colored markers

Step 3: Calculate the distance between a launch site to its proximities

* Explanation:

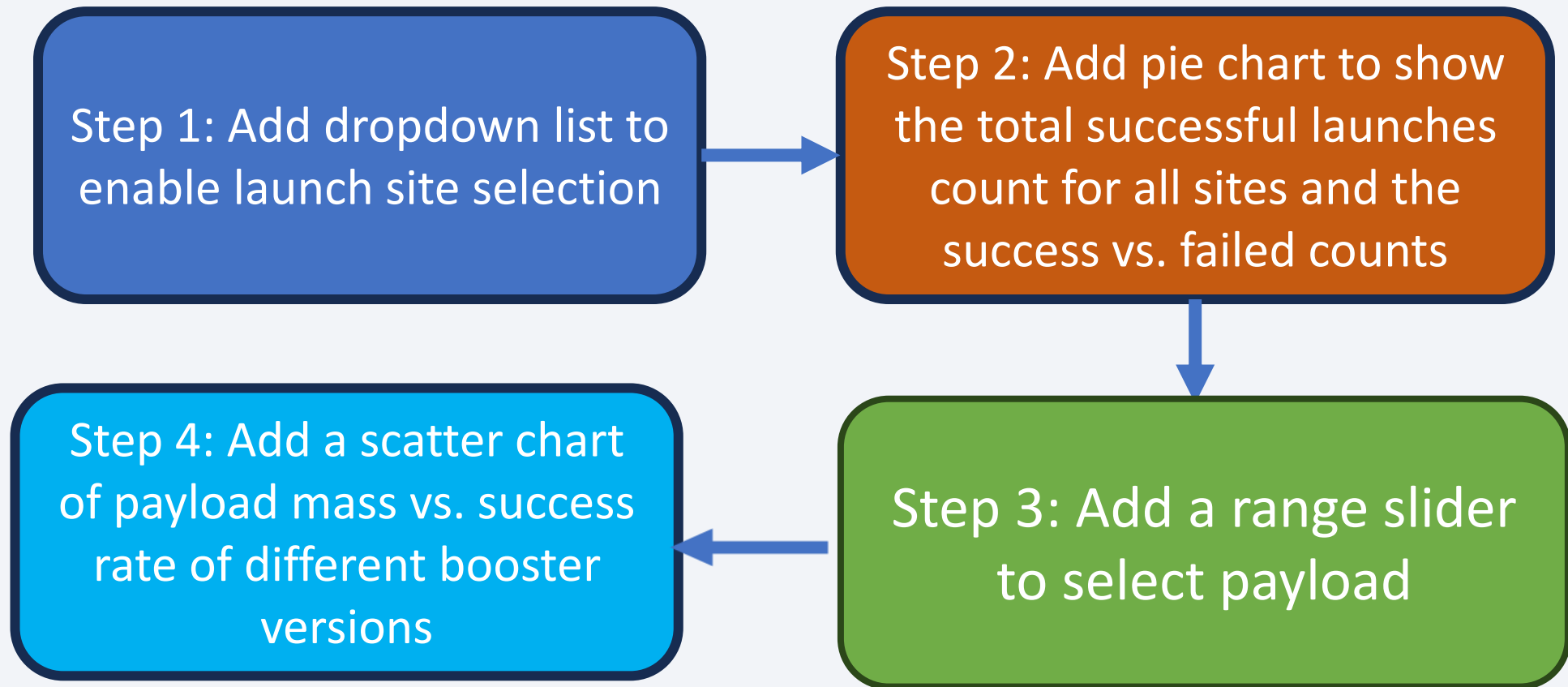
From the visual analysis of the launch site KSC LC-39A we can clearly see that it is:

- relative close to railway (15.23 km)
- relative close to highway (20.28 km)
- relative close to coastline (14.99 km)
- Also the launch site KSC LC-39A is relative close to its closest city Titusville (16.32 km).
- Failed rocket with its high speed can cover distances like 15-20 km in few seconds. It could be potentially dangerous to populated areas.

GitHubURL:

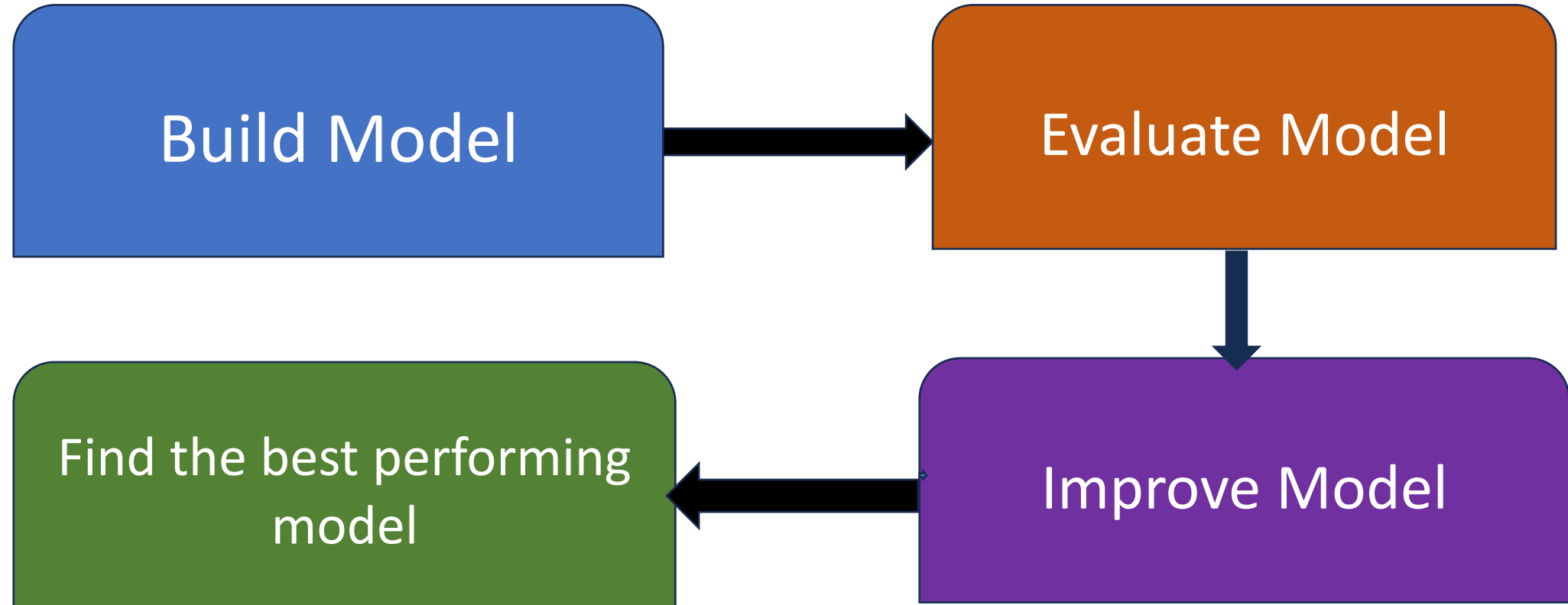
https://github.com/ocean2024/Ocean2024/blob/main/IBM_DA_Capstone_Project_Visual_Analytics_Folium.ipynb

Build a Dashboard with Plotly Dash



The dashboard is built using Dash web

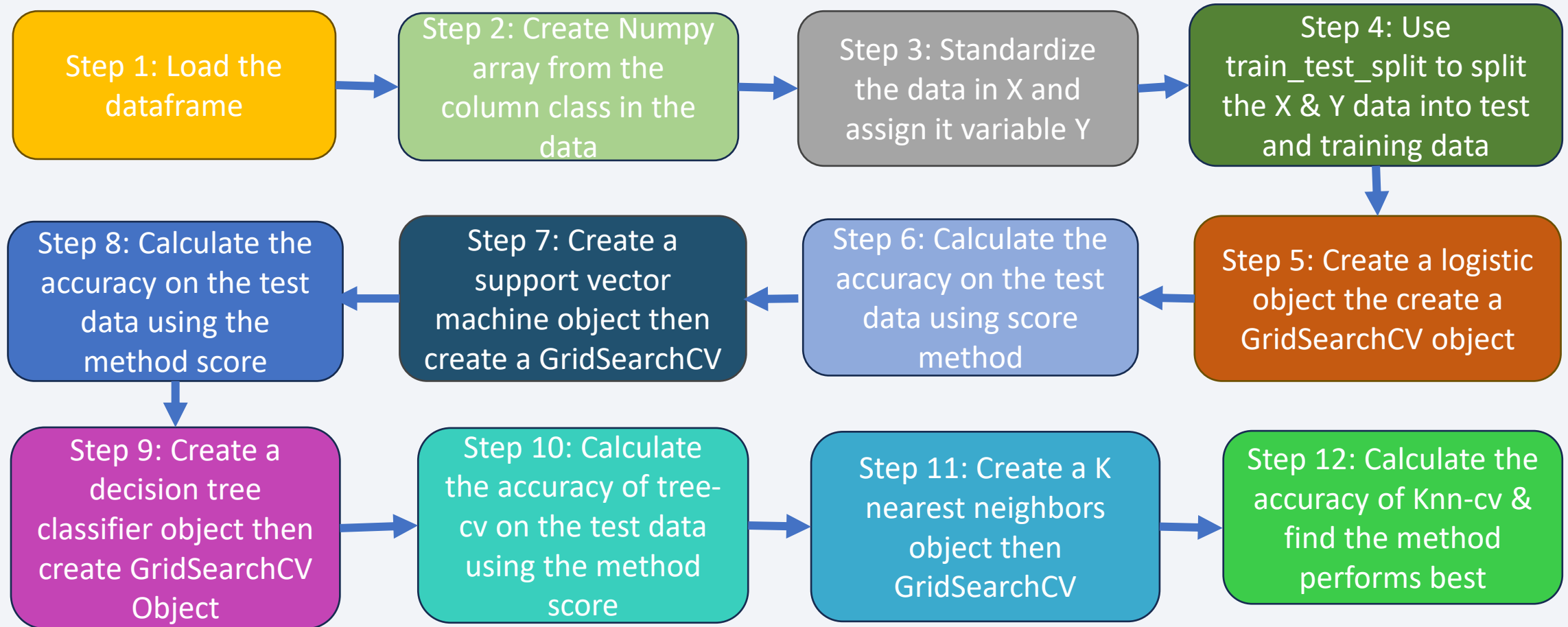
Predictive Analysis (Classification): Overview



GitHub URL:

https://github.com/ocean2024/Ocean2024/blob/main/IBM%20DS%20Capstone%20Project_ML%20Prediction.ipynb

Predictive Analysis (Classification) Steps



GitHub URL:

https://github.com/ocean2024/Ocean2024/blob/main/IBM%20DS%20Capstone%20Project_ML%20Prediction.ipynb

Results

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

Section 2

Insights drawn from EDA

FALCON 9

FIRST STAGE

[OVERVIEW](#) | [ENGINES](#) | [LANDING LEGS](#)

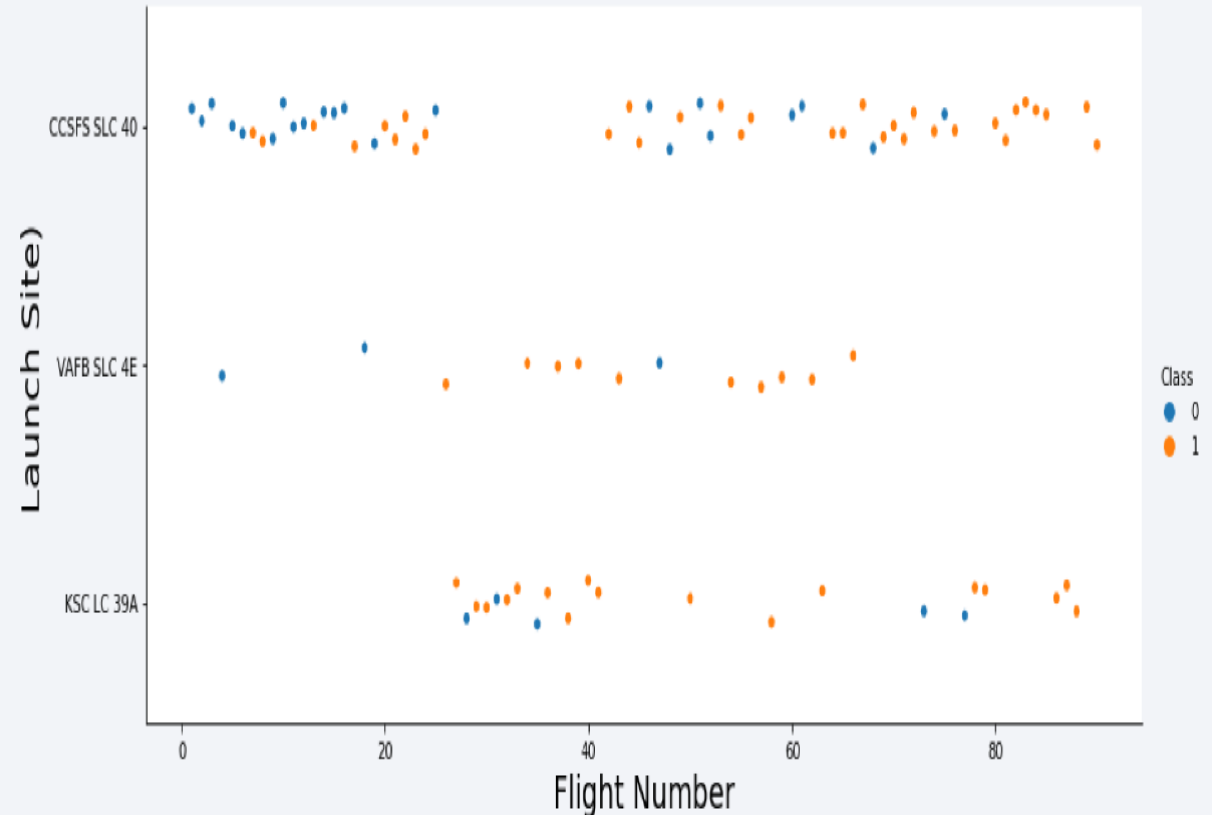
Falcon 9's first stage incorporates nine Merlin engines and aluminum-lithium alloy tanks containing liquid oxygen and rocket-grade kerosene (RP-1) propellant.

Falcon 9 generates more than 1.7 million pounds of thrust at sea level.



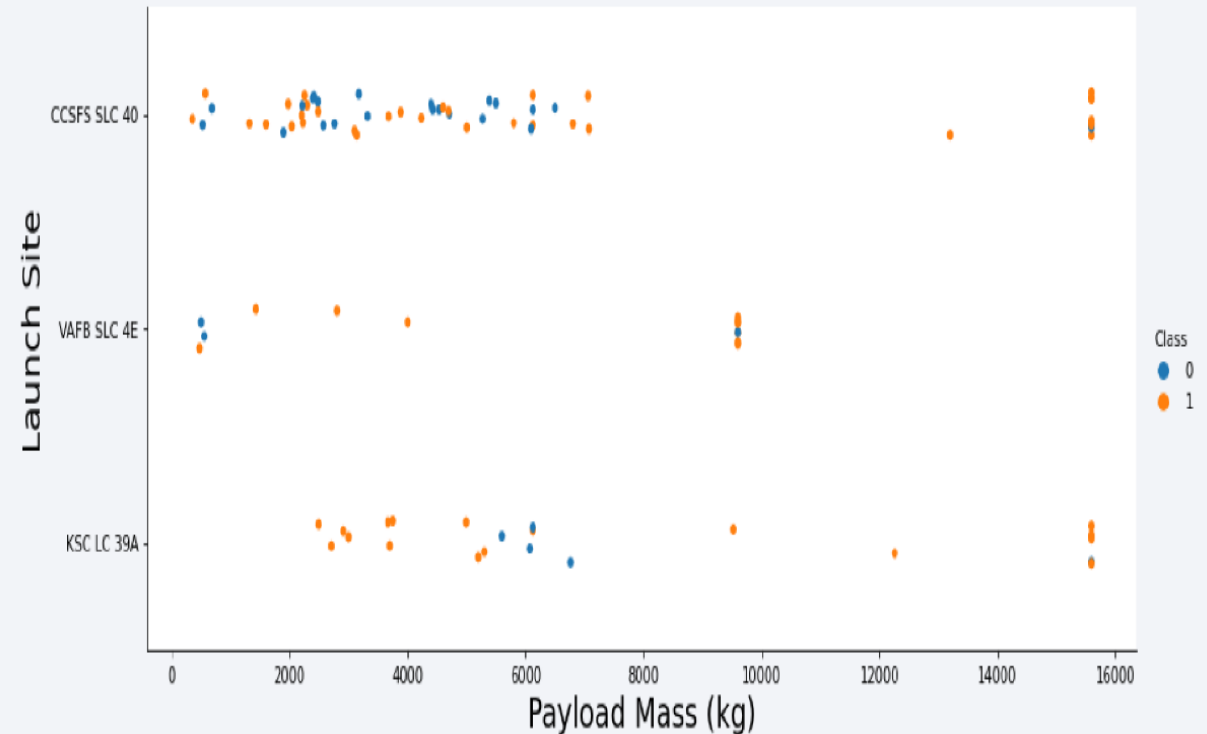
Flight Number vs. Launch Site

- The majority of the flights were launched from the CCAFS SLC 40 site.
- The VAFB SLC 4E and KSC LC 39A sites have higher success rates.
- Newer flights have higher success rate than older flights.



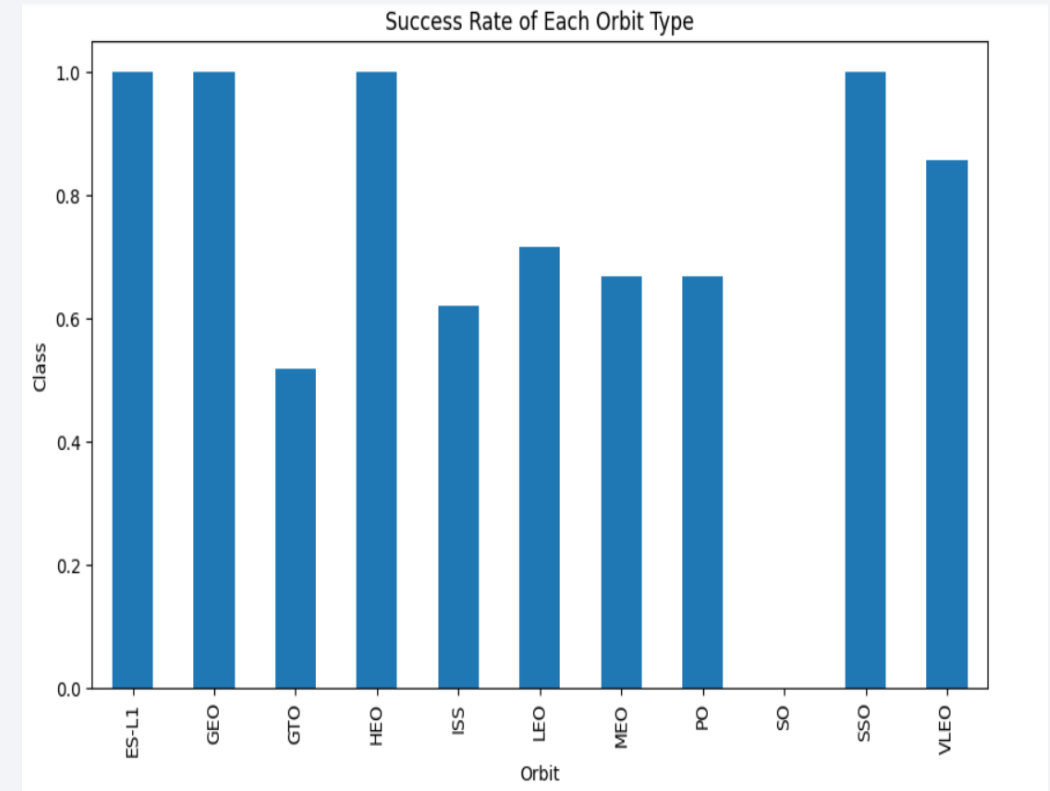
Payload Mass vs. Launch Site

- The majority of the flights with payload mass above 7000 Kg were successful.
- KSC LC 39A success rate for payload mass under 5500 kg is 100%.
- For all launch site the success rate is proportional to the payload mass.



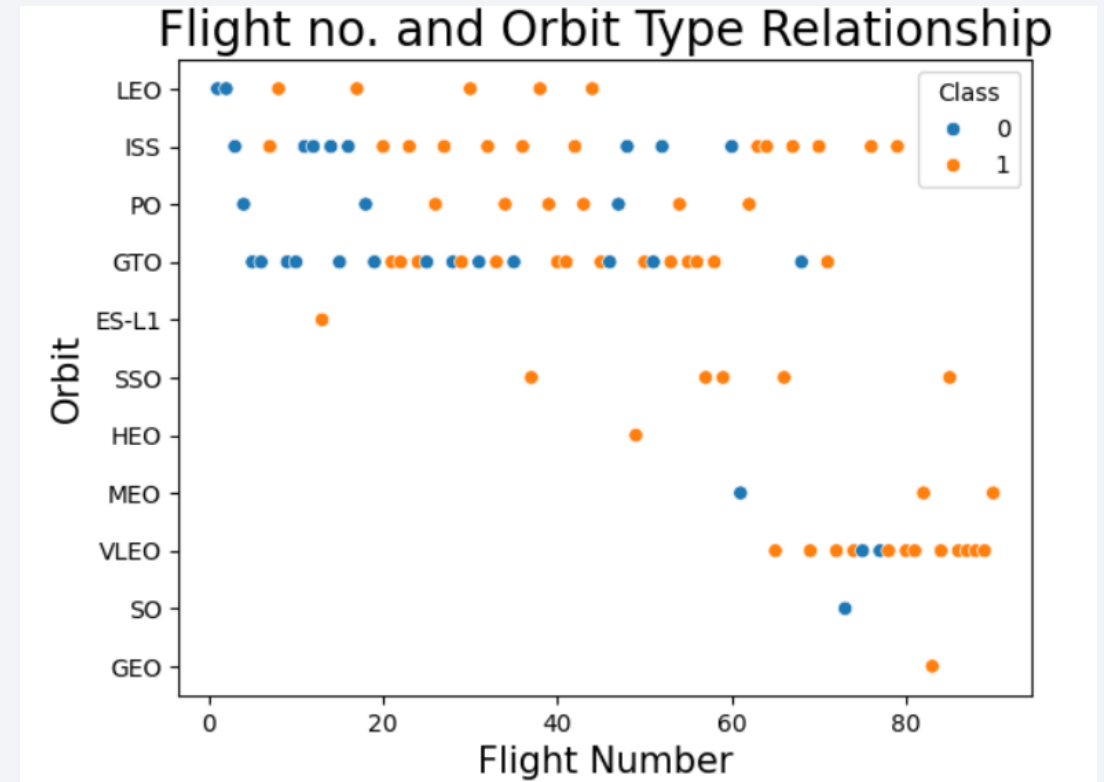
Success Rate vs. Orbit Type

- The OS orbit has 0% success rate.
- The ELS-1, GEO, HEO and SSO orbits have 100% success rate.
- Orbits GTO, ISS, LEO, MEO and PO success rate is higher than 50% and lower than 75% success rate.



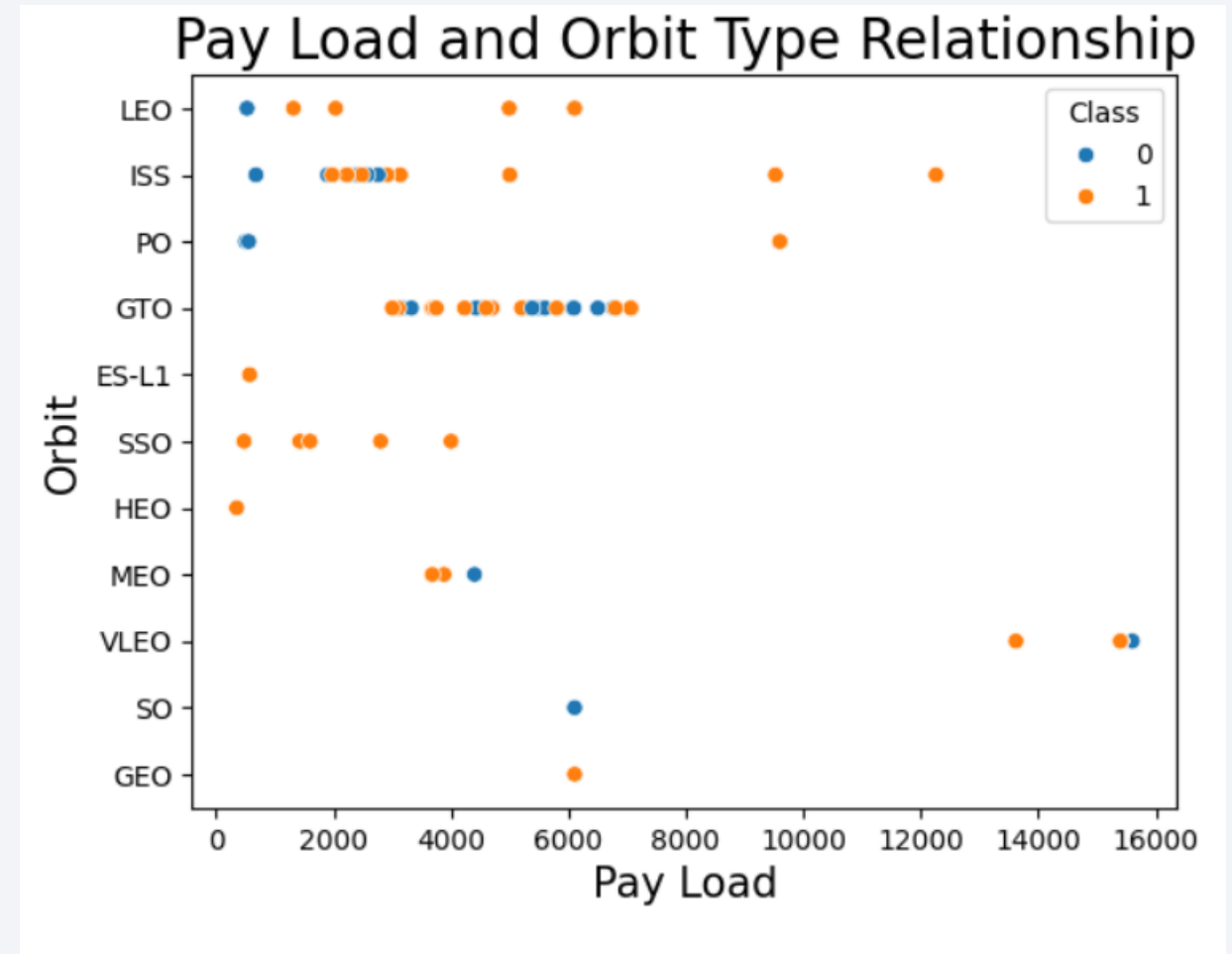
Flight Number vs. Orbit Type

- Flights to the ISS and GTO orbits have the majority of launches.
- No relationship between the flight number and the orbit type.



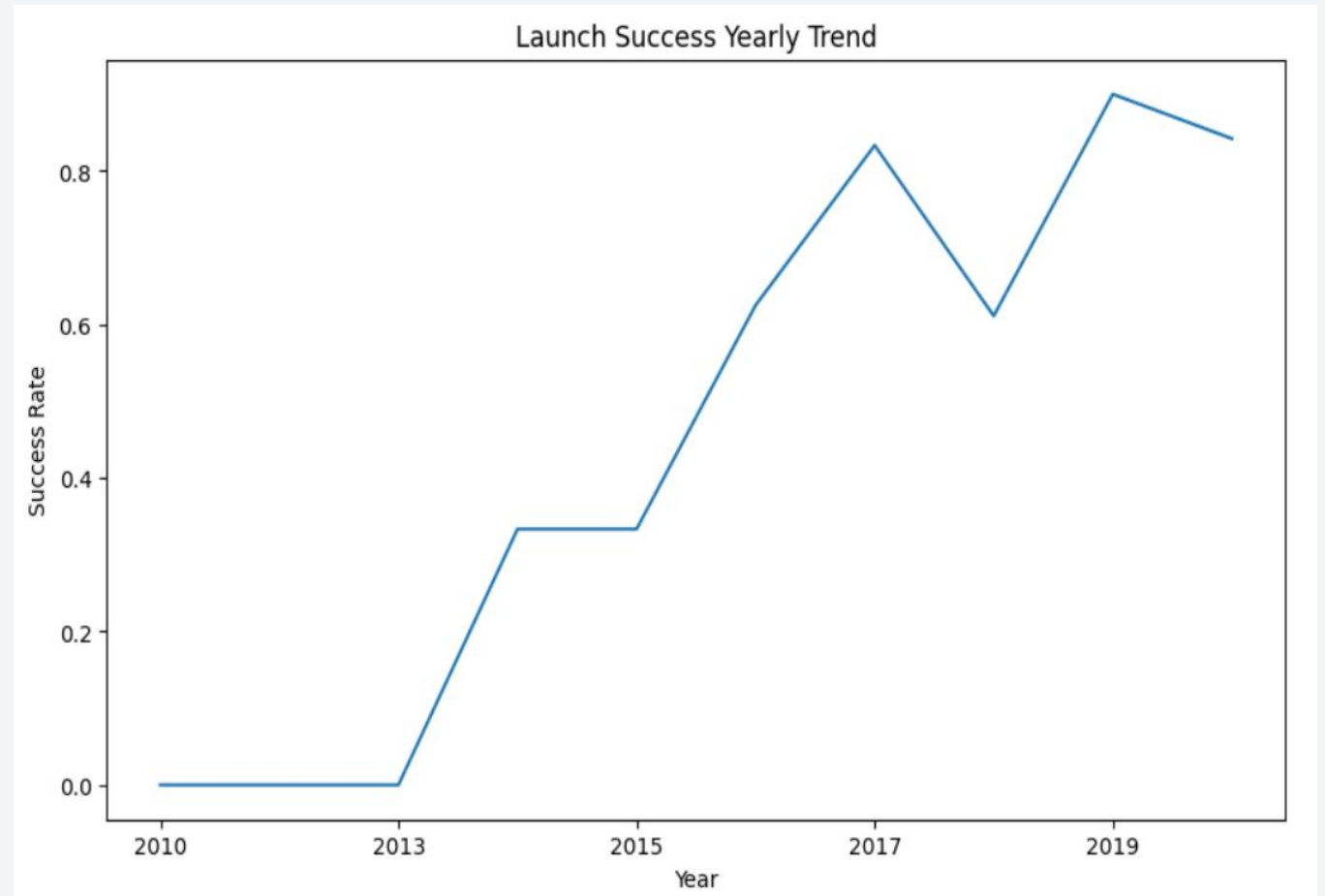
Payload Mass vs. Orbit Type

- Payload masses above 10000 Kg were placed in PO, ISS and LEO orbits.
- Payload masses above 4000 and less than 8000 Kg were placed in the GTO orbit.



Launch Success Yearly Trend

- The launches success rate increased since 2013.
- The increase in the success rate between 2013 and 2017 was linear.
- During 2018 there was a drop in the success rate.



All Launch Site Names

The names of the unique launch sites and the query structure for obtaining these sites us shown below.

```
1 features= df[['LaunchSite']]
2 features.head()
```

	LaunchSite
0	CCAFS SLC 40
1	CCAFS SLC 40
2	CCAFS SLC 40
3	VAFB SLC 4E
4	CCAFS SLC 40

Launch Site Names Begin with 'CCA'

5 records for launch sites begin with the string 'CCA' and the query used for obtaining the information is show below.

```
[ ] 1 %sql SELECT * FROM SPACEXTABLE WHERE launch_site LIKE 'CCA%' LIMIT 5;
```

 * sqlite:///my_data1.db
Done.

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

Total Payload Mass

- The calculated total payload mass carried by boosters from NASA=45596 Kg.
- The query for obtaining the total payload mass is shown below.

```
%sql select SUM(PAYLOAD_MASS_KG_) from SPACEXTABLE where "Customer" like 'NASA (CRS)%'
* sqlite:///my_data1.db
Done.
SUM(PAYLOAD_MASS_KG_)
48213
```

Average Payload Mass by F9 v1.1

- The average payload mass carried by booster version F9 v1.1=2534.7 Kg.
- The query used to calculate the average payload mass carried by booster F9 v1.1 is shown below.

```
%sql select AVG(PAYLOAD_MASS__KG_) from SPACEXTABLE where "Booster_Version" LIKE 'F9 v1.1%'
* sqlite:///my_data1.db
Done.
AVG(PAYLOAD_MASS__KG_)
2534.6666666666665
```

First Successful Ground Landing Date

- The first successful landing outcome on ground pad is 2015-12-22.
- The query result is shown below.

```
[ ] 1 %sql SELECT MIN(Date) AS first_successful_landing_date FROM SPACEXTABLE WHERE landing_outcome = 'Success (ground pad)';
```

```
➔ * sqlite:///my_data1.db  
Done.  
first_successful_landing_date  
2015-12-22
```


Successful Drone Ship Landing with Payload between 4000 and 6000

- List of names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000.
- The query used in obtaining this information is shown below.

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

```
[ ] 1 %sql SELECT Booster_Version FROM SPACEXTABLE WHERE landing_outcome = 'Success (drone ship)' AND PAYLOAD_MASS__KG_ > 4000 AND PAYLOAD_MASS__KG_ < 6000
    2
```

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

Done.

Booster_Version

F9 FT B1022

F9 FT B1026

F9 FT B1021.2

F9 FT B1031.2

Total Number of Successful and Failure Mission Outcomes

- The total number of successful and failure mission outcomes:
 - Failure (in flight)= 1
 - Successful number of flights= 98
- The query result is shown below.

```
[ ] 1 %sql SELECT mission_outcome, COUNT(*) AS total_count FROM SPACEXTABLE WHERE mission_outcome IN ('Success', 'Failure (in flight)') GROUP BY mis
2
```

```
↳ * sqlite:///my_data1.db
Done.
Mission_Outcome total_count
Failure (in flight)    1
Success                98
```

Boosters Carried Maximum Payload

- List of the booster which have carried the maximum payload mass are shown below.
- The query used in obtaining the names is shown below.

List the names of the booster_versions which have carried the maximum payload mass. Use a subquery

```
[ ] 1 %sql SELECT booster_version FROM SPACEXTABLE WHERE PAYLOAD_MASS__KG_ = (SELECT MAX (PAYLOAD_MASS__KG_) FROM SPACEXTABLE);  
2
```

```
↳ * sqlite:///my_data1.db
```

Done.

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

2015 Launch Records

- List of the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015 are listed below.
- The query used in obtaining the information is shown below.

```
[ ] 1 %sql SELECT CASE WHEN substr(Date, 6, 2) = '01' THEN 'January' WHEN substr(Date, 6, 2) = '02' THEN 'February' WHEN substr(Date, 6, 2) = '03' TH
2
```

```
↳ * sqlite:///my_data1.db
```

Done.

month_name	Booster_Version	Launch_Site	Landing_Outcome
January	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
April	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

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

- A rank of 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 is shown below.
- The query used to obtain the results is shown below.

```
[ ] 1 %sql SELECT landing_outcome, COUNT(*) AS outcome_count FROM SPACEXTABLE WHERE Date BETWEEN '2010-06-04' AND '2017-03-20' GROUP BY landing_outcc  
2
```

```
➔ * sqlite:///my_data1.db  
Done.  
Landing_Outcome outcome_count  
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
```

Section 3

Launch Sites Proximities Analysis

FALCON 9

INTERSTAGE

The interstage is a composite structure that connects the first and second stages, and houses the pneumatic pushers that allow the first and second stage to separate during flight.

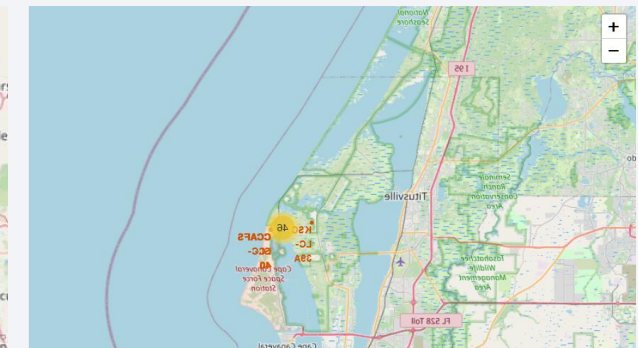
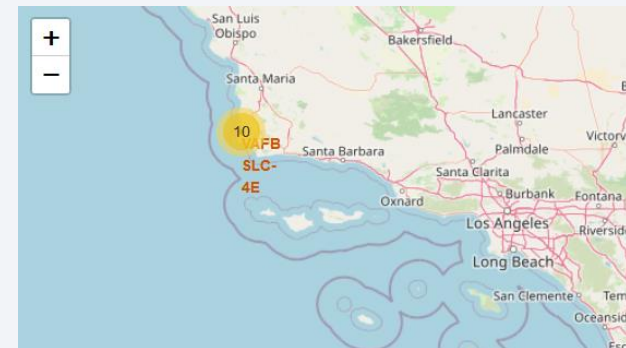
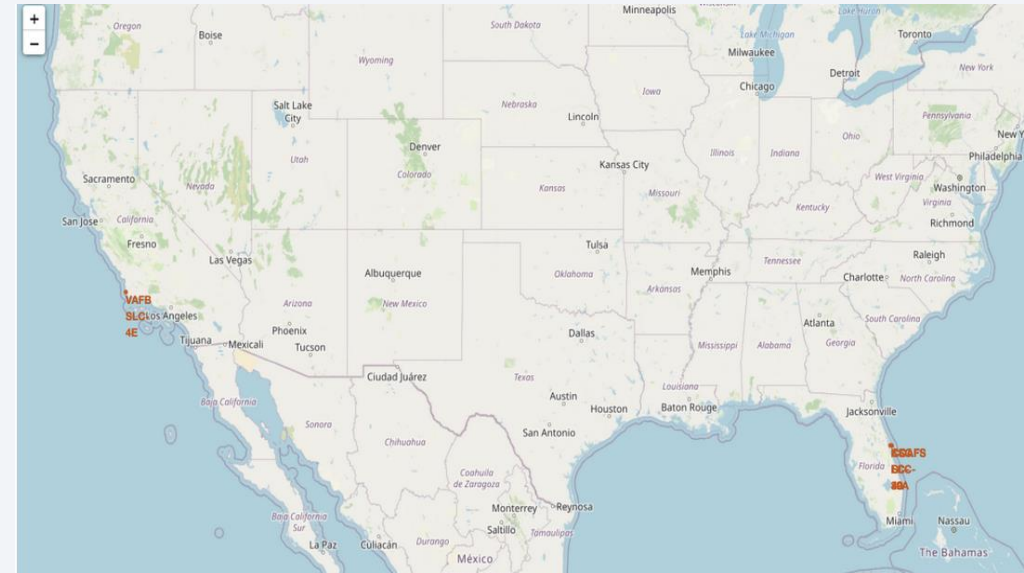
GRID FINS

Falcon 9 is equipped with four hypersonic grid fins positioned at the base of the interstage. They orient the rocket during reentry by moving the center of pressure.



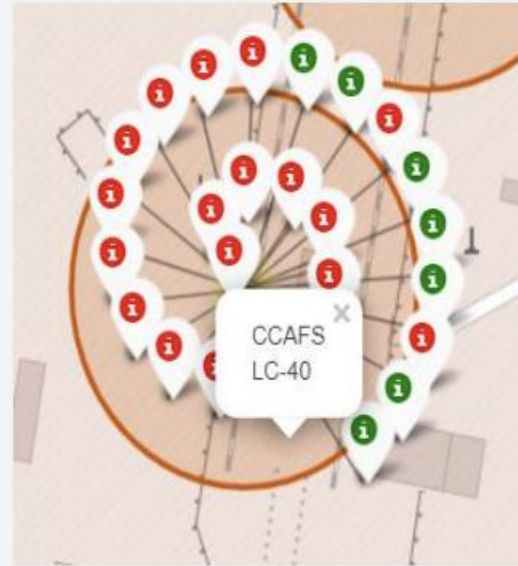
USA Launch Sites in California and Florida

- Most of Launch sites considered in this project are in proximity to the Equator line. Launch sites are made at the closest point possible to Equator line, because anything on the surface of the Earth at the equator is already moving at the maximum speed (1670 kilometers per hour). For example launching from the equator makes the spacecraft move almost 500 km/hour faster once it is launched compared half way to north pole.
- All launch sites considered in this project are in very close proximity to the coast While starting rockets towards the ocean we minimize the risk of having any debris dropping or exploding near people.

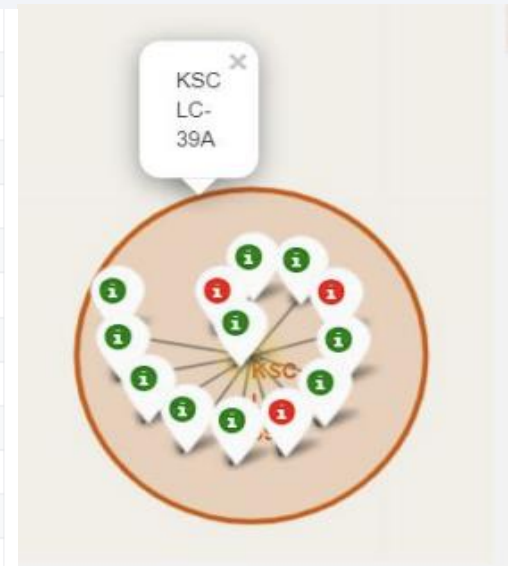


Color Labels Showing the Launch Sites on a Map

	Launch Site	Lat	Long	marker_color
0	CCAFS LC-40	28.562302	-80.577356	red
1	CCAFS LC-40	28.562302	-80.577356	red
2	CCAFS LC-40	28.562302	-80.577356	red
3	CCAFS LC-40	28.562302	-80.577356	red
4	CCAFS LC-40	28.562302	-80.577356	red
5	CCAFS LC-40	28.562302	-80.577356	red
6	CCAFS LC-40	28.562302	-80.577356	red
7	CCAFS LC-40	28.562302	-80.577356	red
8	CCAFS LC-40	28.562302	-80.577356	red
9	CCAFS LC-40	28.562302	-80.577356	red
10	CCAFS LC-40	28.562302	-80.577356	red
11	CCAFS LC-40	28.562302	-80.577356	red
12	CCAFS LC-40	28.562302	-80.577356	red
13	CCAFS LC-40	28.562302	-80.577356	red
14	CCAFS LC-40	28.562302	-80.577356	red
15	CCAFS LC-40	28.562302	-80.577356	red
16	CCAFS LC-40	28.562302	-80.577356	red
17	CCAFS LC-40	28.562302	-80.577356	green
18	CCAFS LC-40	28.562302	-80.577356	green
19	CCAFS LC-40	28.562302	-80.577356	red
20	CCAFS LC-40	28.562302	-80.577356	green



36	KSC LC-39A	28.573255	-80.646895	green
37	KSC LC-39A	28.573255	-80.646895	red
38	KSC LC-39A	28.573255	-80.646895	green
39	KSC LC-39A	28.573255	-80.646895	green
40	KSC LC-39A	28.573255	-80.646895	red
41	KSC LC-39A	28.573255	-80.646895	green
42	KSC LC-39A	28.573255	-80.646895	green
43	KSC LC-39A	28.573255	-80.646895	red
44	KSC LC-39A	28.573255	-80.646895	green
45	KSC LC-39A	28.573255	-80.646895	green
46	KSC LC-39A	28.573255	-80.646895	green
47	KSC LC-39A	28.573255	-80.646895	green
48	KSC LC-39A	28.573255	-80.646895	green

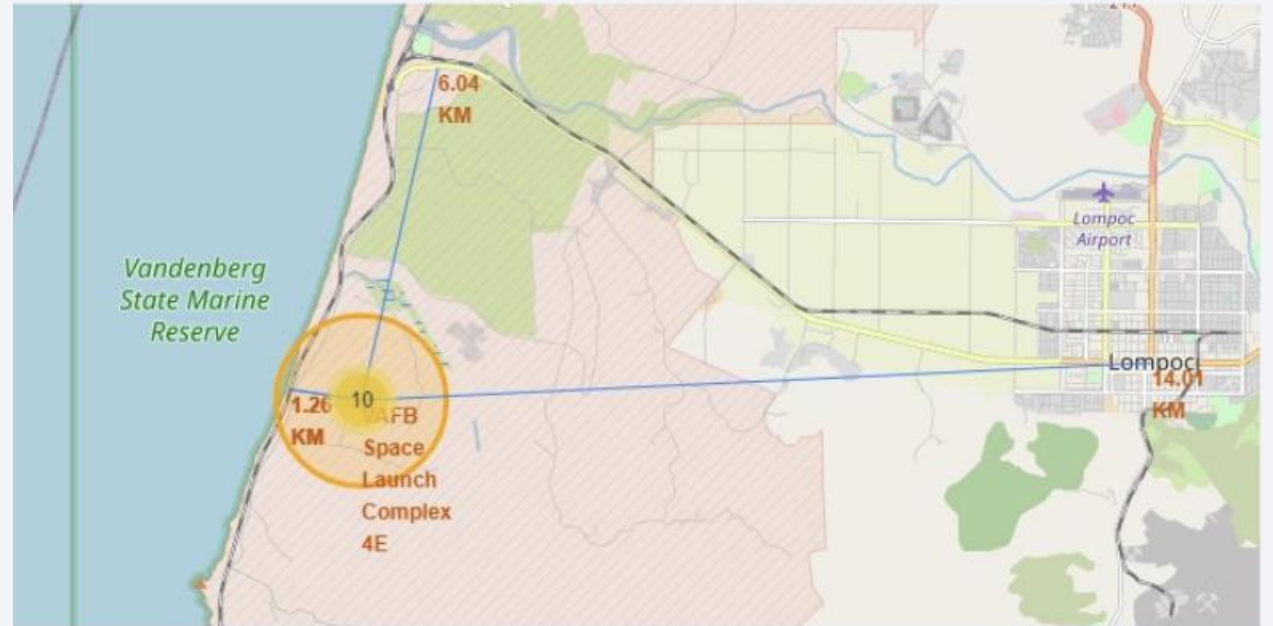


Green= Successful Launch

Red= Failed Launch

Safe Distance to Launch Site

Launch sites at. safe distance from railway and cities.



Section 4

Build a Dashboard with Plotly Dash

FALCON 9

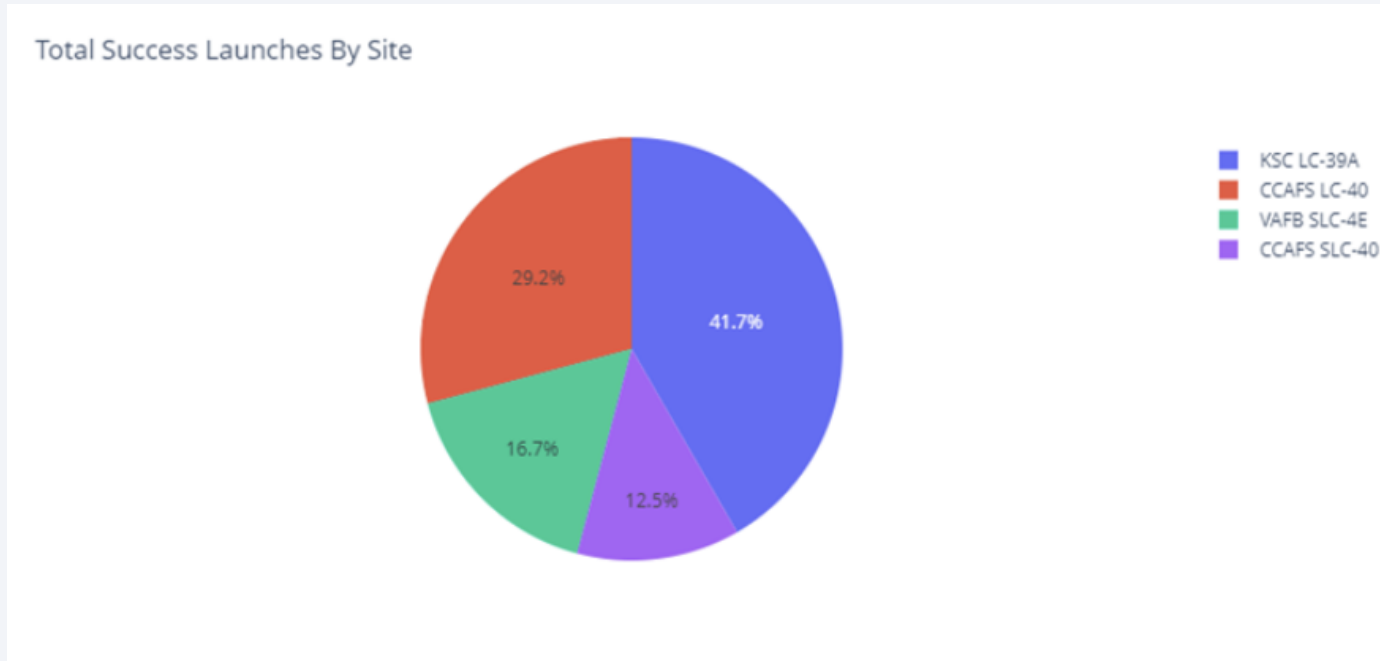
SECOND STAGE

The second stage, powered by a single Merlin Vacuum Engine, delivers Falcon 9's payload to the desired orbit. The second stage engine ignites a few seconds after stage separation, and can be restarted multiple times to place multiple payloads into different orbits.

NUMBER OF ENGINES	1 vacuum
BURN TIME	397 sec
THRUST	981 kN / 220,500 lbf



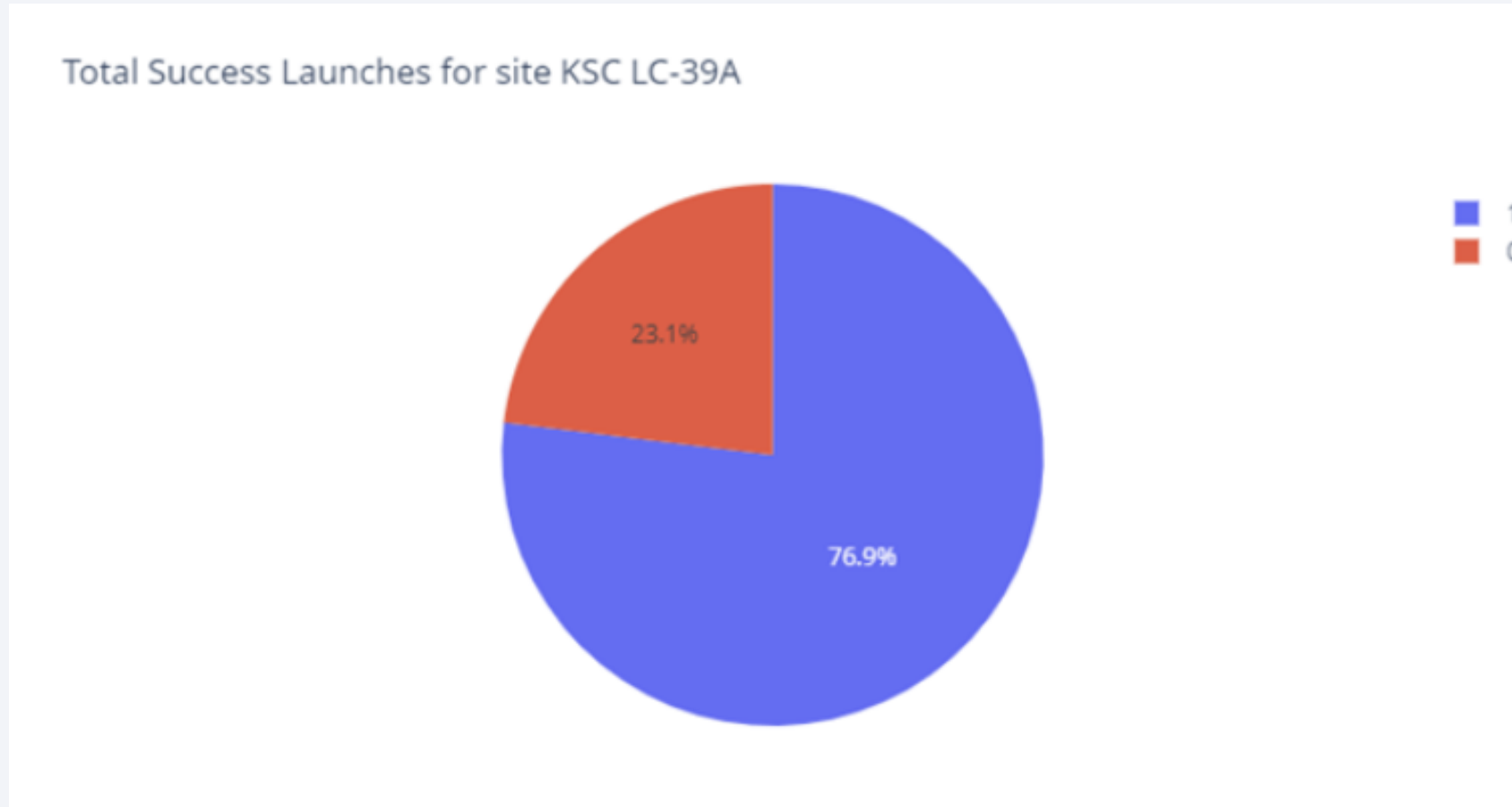
Total Launch Success for All Sites



The Highest Success Launch Sites are :

1. KSC LC-39A (41.7%)
2. CCAFS LC-40 (29.2%)

KSC LC-39 Launch Site Success Rate



Site KSC LC-39 success rate is 76.9%

Payload vs. Launch Outcome for All Sites

**Highest success rate for
Payloads Between 2000
and 5500 Kgs**



Section 5

Predictive Analysis (Classification)

FALCON 9

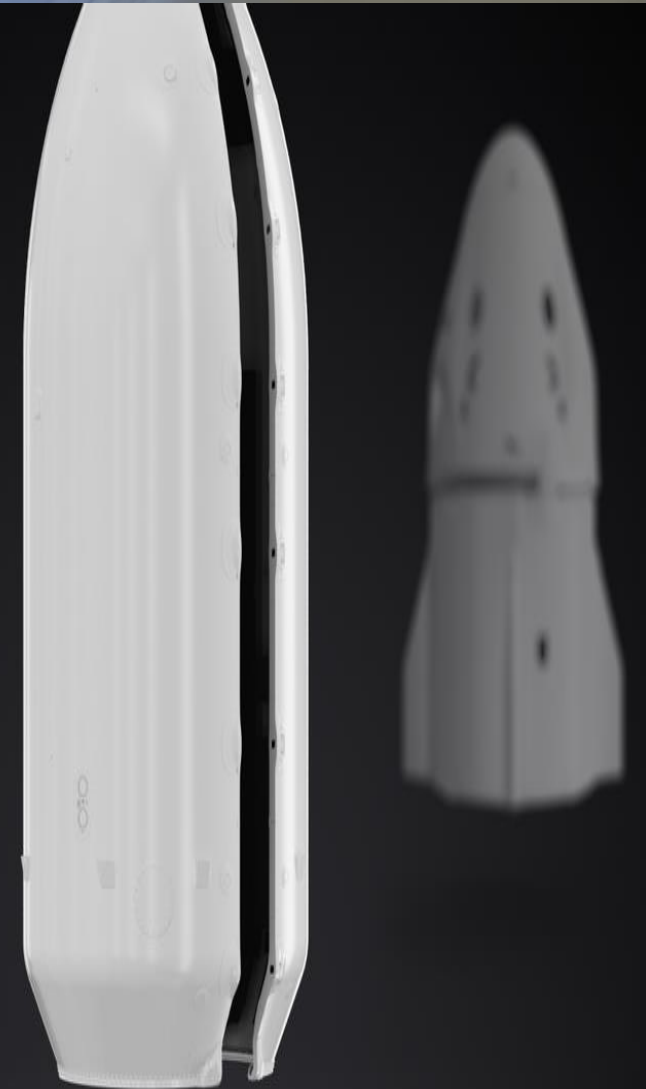
PAYLOAD

FAIRING | DRAGON

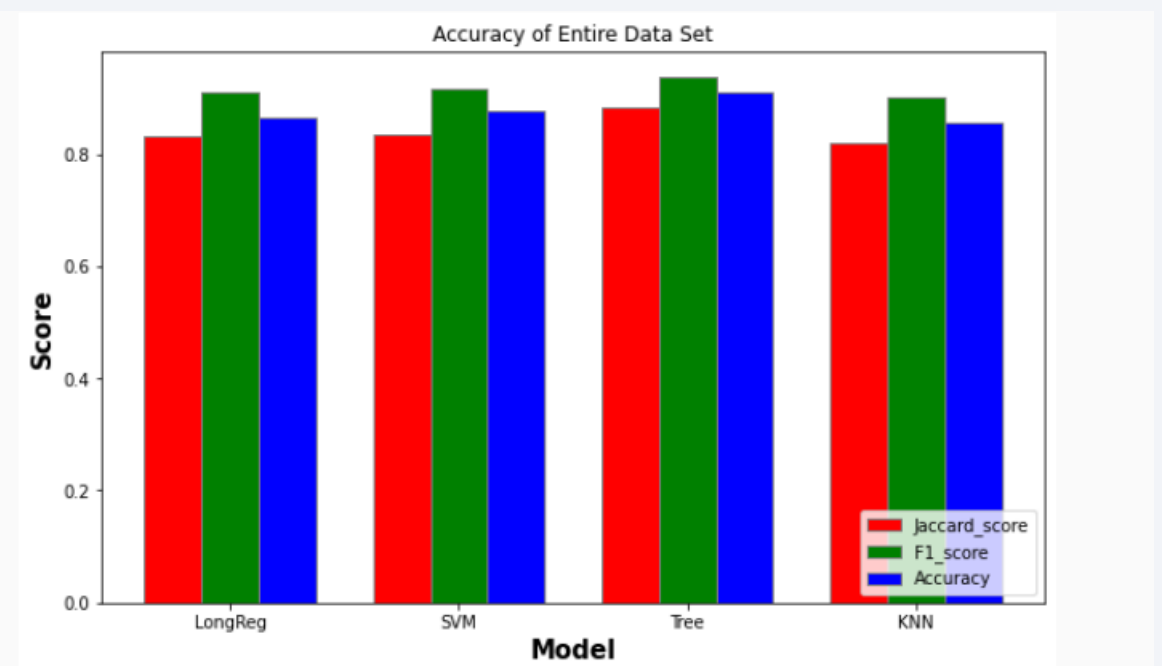
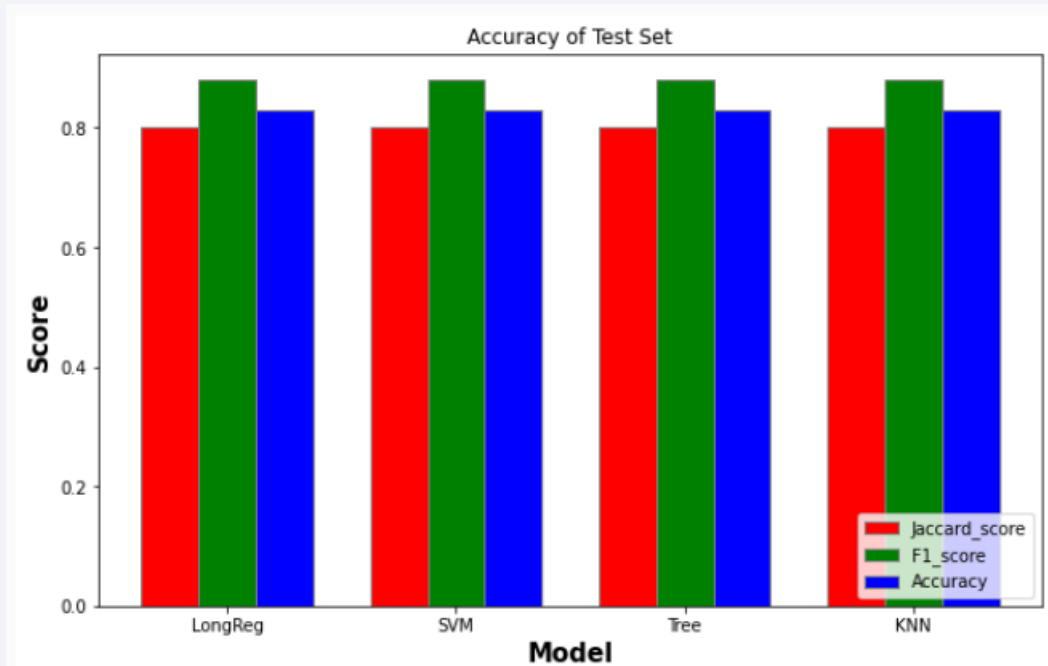
Made of a carbon composite material, the fairing protects satellites on their way to orbit. The fairing is jettisoned approximately 3 minutes into flight, and SpaceX continues to recover fairings for reuse on future missions.

HEIGHT 13.1 m / 43 ft

DIAMETER 5.2 m / 17.1 ft



Classification Accuracy

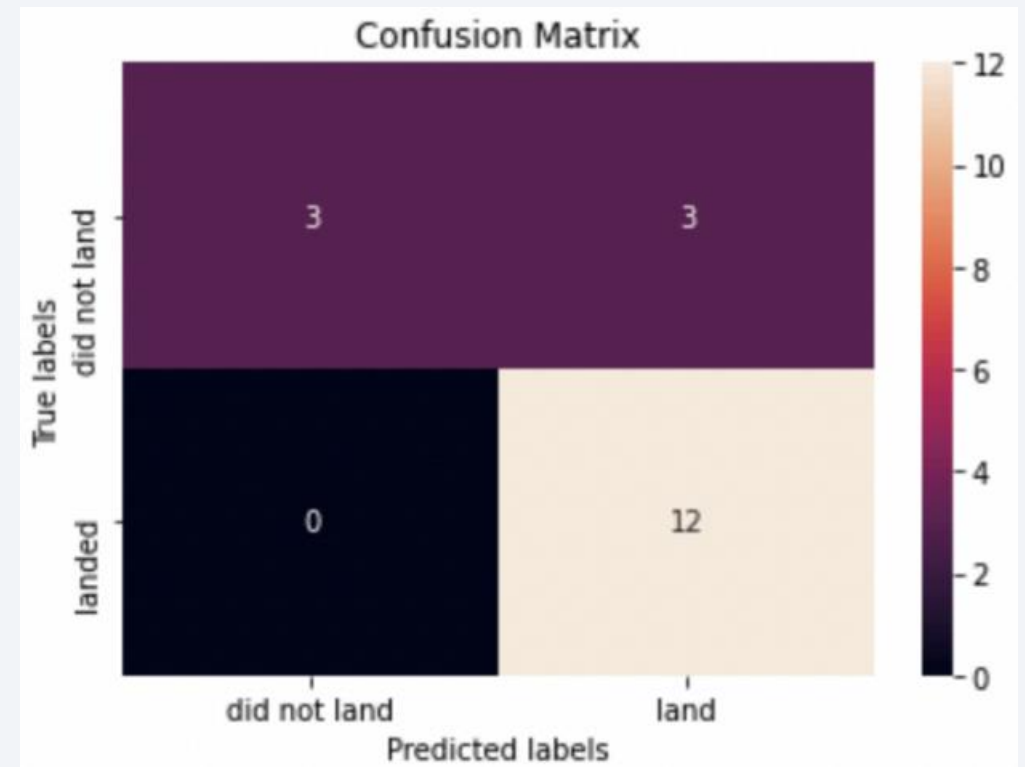


- Using the test set the same accuracy results were obtained from the four models.
- However, using the entire data set the Tree model provided the best results.

Confusion Matrix

- Show the confusion matrix of the best performing model which is the Logistic Regression model.
- The confusion matrix predicts 13 true positives, 3 false positives, 3 true negative, and 0 false negative.

		Predicted Values	
		Negative	Positive
Actual Values	Negative	TN	FP
	Positive	FN	TP



Conclusions

- The success rate of the rocket launches increased with the years since 2013.
- Orbits GEO, HEO, ES-L1 and SSO have 100% launch success rate.
- Launch site KSC LC-39A has the highest success rate
- Decision tree model is the best ML algorithm for analyzing the SpaceX data set providing the best accuracy.



Appendix

<https://github.com/ocean2024/Ocean2024/tree/main>

https://www.coursera.org/professional-certificates/ibm-data-science?campaignid=1876641588&adgroupid=70740725700&device=c&keyword=ibm%20data%20science%20professional%20certificate&matchtype=b&network=g&devicemodel=&adposition=&creativeid=347445112274&hide_mobile_promo=&gad_source=1



Thank you!

