



IBM Developer  
SKILLS NETWORK

# Winning Space Race with Data Science

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# Outline

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

# Executive Summary

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- Summary of methodologies
  - **Data Collection** is done from sources such as SpaceX API and Webscraping from Wikipedia Webpage
  - **Data Wrangling** is done to transform the data, so that we can summarize the outputs for easy classification.
  - **EDA** is done using **Visualization and SQL** queries to gain first insights.
  - **Using interactive map** explore launch sites with regards to their outcomes, proximities to landmarks
  - **Classification** is done using Logistic Regression, SVM, Decision Tree, KNN and analyzed.
- Summary of all results
  - Though the **success rate** was very low start of experiment, towards the end it is very high
  - All launch sites are in good **proximity** to coastlines and equator and have good connectivity to roads, highway and railways
  - For the **orbits**, ES-L1, SSO and HEO, there is low payload and 100% success rate
  - KSC LC-39A is the **launch site** with highest success rate
  - Decision Tree Model showed the highest **classification** accuracy

# Introduction

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- Problem Statement - Predict if the Falcon 9 first stage will land successfully
  - SpaceX advertises their Falcon 9 rocket launches with a cost of 62 million dollars. Much of the savings is because SpaceX can reuse the first stage
  - Other providers cost upward of 165 million dollars each
  - If we can predict, that the first stage will land or not, we can determine the cost of a launch. And an alternate company can bid against accordingly
- Goals
  - Collect Falcon 9 launch data from SpaceX API and using Webscraping
  - Prepare data, from various outputs using Data Wrangling
  - Explore relation between variables like Flight Number, Payload, Orbit, Success etc.
  - Explore various Launch Sites and their Success Rates and Location Significance
  - Predict the launch success by various machine learning algorithms and compare



Section 1

# Methodology

# Methodology

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## Executive Summary

- Data collection methodology:
  - Falcon 9 data is obtained from SpaceX API and using Webscraping.
- Perform data wrangling
  - Transform different data outcomes to binary outcomes, 'Success' and 'Failure'
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Build a Machine Learning Pipeline for the prediction.
  - Its initial stages include preprocessing and train test dataset split.
  - Use different algorithms and determined their best hyperparameters using GridSearch
  - Determine model with best accuracy for training model and use confusion matrix to assess output

# Data Collection

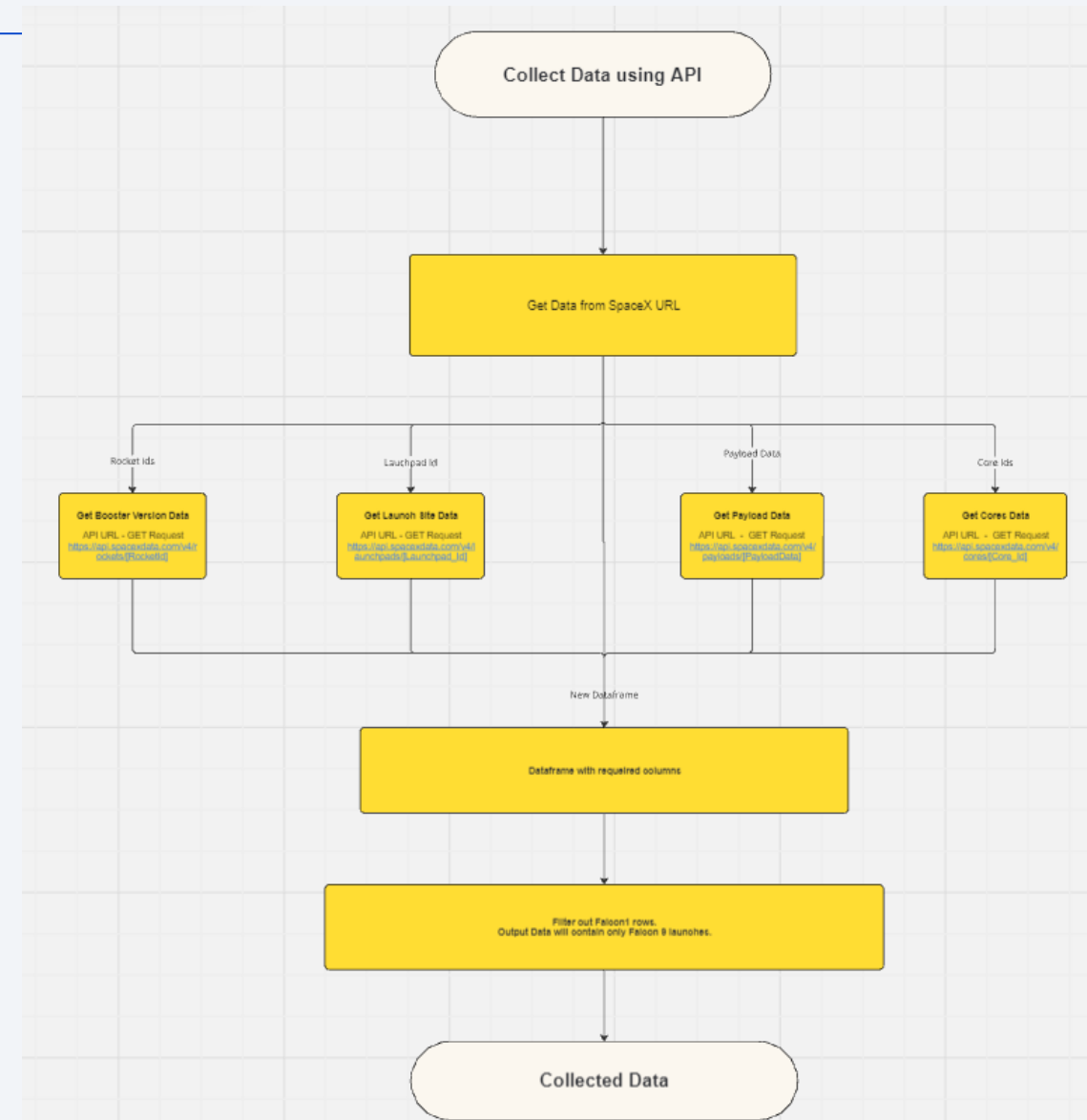
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- Data Collection is done from two sources
  1. SpaceX REST API
  2. Webscraping data from Wikipedia website
- Data collected will have data of all launches and we use filtering to get Falcon 9 data
- Data Wrangling is done to convert data outcomes to either Success or Failure

# Data Collection – SpaceX API

- Basic data of past launches is collected first by calling the main API url which has past launch site data.
- Detailed information on **BoosterVersion**, **Cores**, **PayloadMass**, **LaunchSite** is collected later by calling the required APIs using ids of them received in first API call and it is stored in the dataframe.
- GitHub URL –

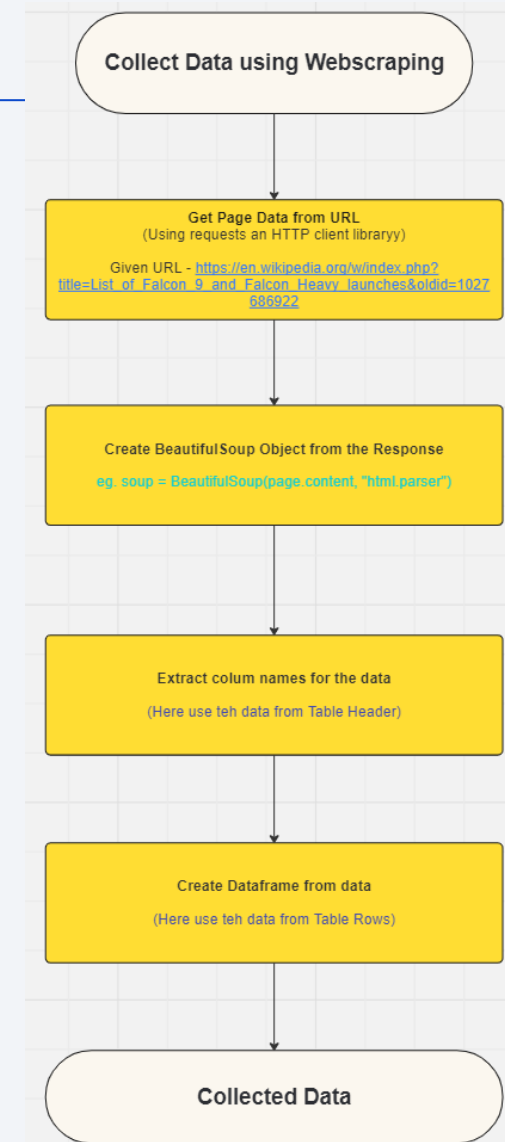
[https://github.com/devmanac/MOOC\\_Assignment/blob/main/jupyter-labs-spacex-data-collection-api.ipynb](https://github.com/devmanac/MOOC_Assignment/blob/main/jupyter-labs-spacex-data-collection-api.ipynb)





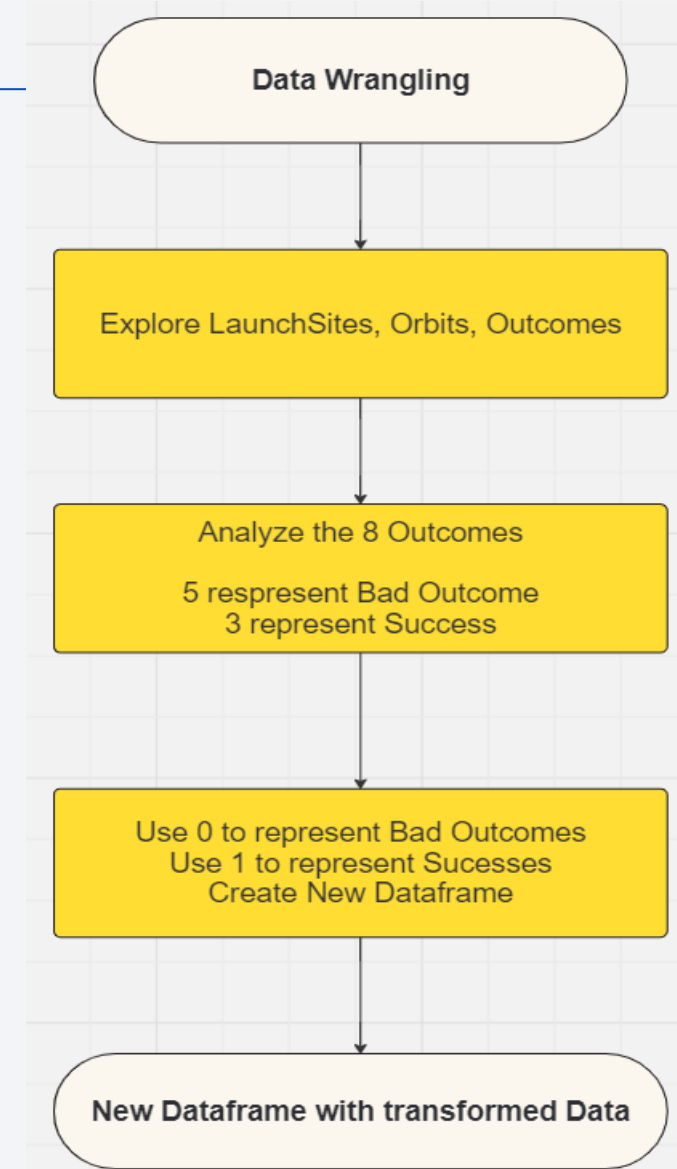
# Data Collection - Scraping

- For Webscraping first the **HTML page** is accessed and stored **into an object**
- From this objects the **headers and rows** to be filled in dataframe is **extracted**.
- GitHub URL - [https://github.com/devmanac/MOOC\\_Asignment/blob/main/jupyter-labs-webscraping.ipynb](https://github.com/devmanac/MOOC_Asignment/blob/main/jupyter-labs-webscraping.ipynb)



# Data Wrangling

- Explore the data to find the details and use Data Wrangling to create the Dataframe we need
- In the analysis we will see the types of Launchsites, Orbits and Outcomes.
- There are **8 outcomes** from which we need to find the outcome is 'Success' or 'Failure'. We reduce the outcomes to these binary outcomes for further analysis we need.
- GitHub URL - [https://github.com/devmanac/MOOC\\_Assignment/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb](https://github.com/devmanac/MOOC_Assignment/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb)



# EDA with Data Visualization

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Github URL -

[https://github.com/devmanac/MOOC\\_Assignment/blob/main/edadataviz.ipynb](https://github.com/devmanac/MOOC_Assignment/blob/main/edadataviz.ipynb)

- List of charts
  1. Seaborn Categorical Plot **Flight Number vs Payloadmass**. Outcome is presented as an overlay. We could see the progress in Payloadmass in time and also the Success Rate with time.
  2. Seaborn Categorical Plot **LaunchSite vs. Flight Number**. Outcome is presented as an overlay. Analyse the outcome in various LaunchSites in relation to FlightNumber.
  3. Seaborn **Scatter Plot LaunchSite vs. Flight Number**. Outcome is presented as an overlay
  4. Using **scatter plot** analyze the same relation, LaunchSites in relation to FlightNumber

# EDA with Data Visualization

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5. **Seaborn Barplot** to visualize relation between **Success Rate** and **Orbit Type**
6. Visualize the relationship between **FlightNumber** and **Orbit** with Outcome as an overlay in a Scatter Plot.
7. Visualize the relationship between **Payloadmass** and **Orbit** with Outcome as an overlay in a Scatter Plot.
8. Use a **line plot** to visualize **Success Rate** over the Years

# EDA with SQL

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- We **analyze the Launch Sites** in the first couple of SQL queries
- **Payload mass tendencies, Booster Version, Launches** related to particular customer etc. are analyzed in the next couple of SQL queries
- In the following few queries we analyze the **Landing Outcome** with regards to launch pad, booster version, payload mass, number, max value etc.
- In the last queries we check the **Success and Failure in missions** in certain years, time frames, their counts, booster version

GitHub URL -

[https://github.com/devmanac/MOOC\\_Assignment/blob/main/jupyter-labs-eda-sql-coursera\\_sqlite.ipynb](https://github.com/devmanac/MOOC_Assignment/blob/main/jupyter-labs-eda-sql-coursera_sqlite.ipynb)



# Build an Interactive Map with Folium

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- Initially the **number of launches** are shown in a **circle** over launch sites
- Clicking further we could see the Success and Failures denoted by **green and red markers**
- **Popups and labels** are provided to see the name of each launch site
- **Lines showing distance** is drawn to important geographic locations like highway, roads, coastline etc. with distance shown
- Launch sites are all **in close proximity** to equator, coastline, railway and roads
- Github URL - [https://github.com/devmanac/MOOC\\_Assignment/blob/main/lab\\_jupyter\\_launch\\_site\\_location.ipynb](https://github.com/devmanac/MOOC_Assignment/blob/main/lab_jupyter_launch_site_location.ipynb)

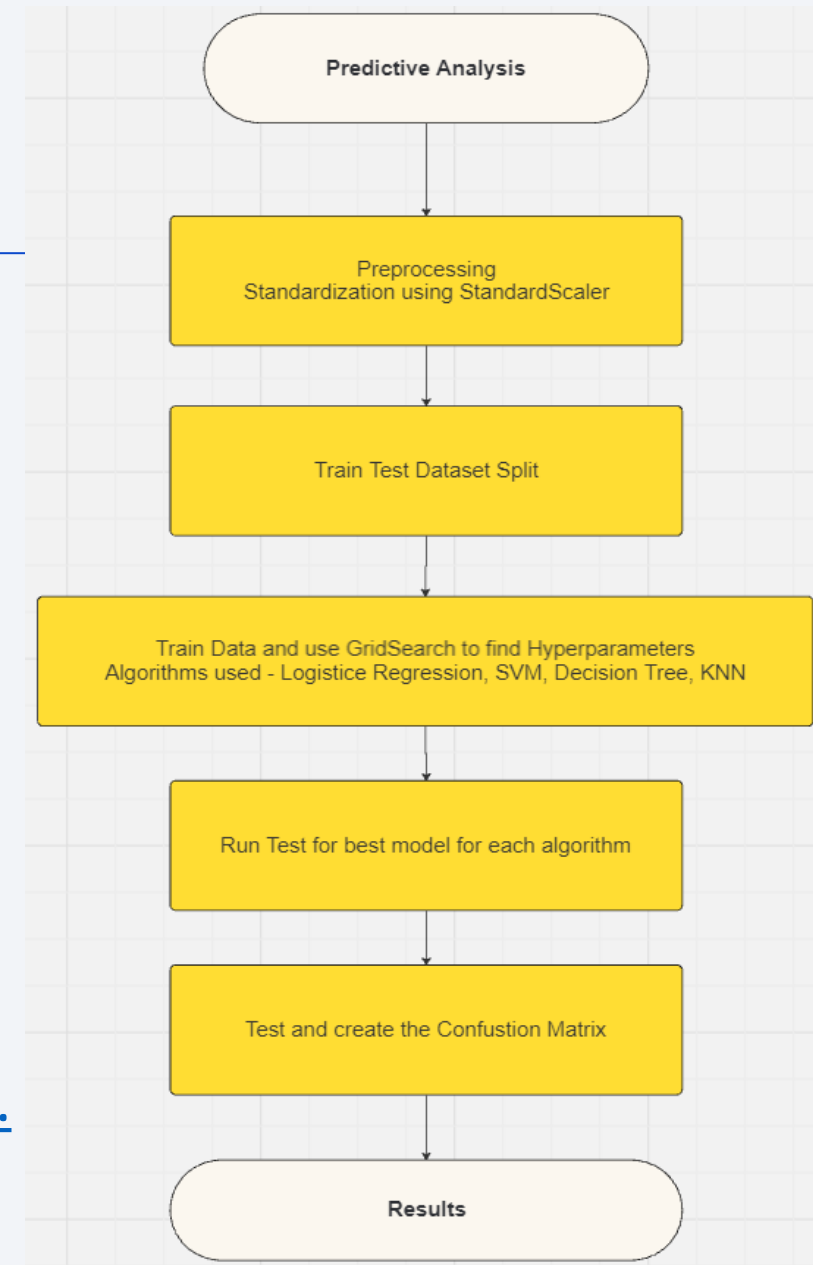
# Build a Dashboard with Plotly Dash

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- Dynamic loading of data is done in chart outputs with inputs from Dropdown and Slider values
- Dropdown contains the launch sites where all or any launch site could be selected
- Slider corresponds to Payload Mass
- Pie chart shows the success rate of the launch site(s) selected in Dropdown
- Scatter plot takes the launch site input from dropdown and payload range from slider as inputs and shows the Success or Failure with Booster Version information
- Github URL – [https://github.com/devmanac/MOOC\\_Assignment/blob/main/spacex\\_dash\\_app.py](https://github.com/devmanac/MOOC_Assignment/blob/main/spacex_dash_app.py)

# Predictive Analysis (Classification)

- We create a machine learning pipeline for prediction
  - Standardize Data
  - Train Test Split
  - Find best Hyperparameters during Training for different algorithms
  - Test the models and create Confusion Matrix to see the best model
- GitHub URL - [https://github.com/devmanac/MOOC\\_Assignment/blob/main/SpaceX\\_Machine%20Learning%20Prediction\\_Part\\_5.ipynb](https://github.com/devmanac/MOOC_Assignment/blob/main/SpaceX_Machine%20Learning%20Prediction_Part_5.ipynb)



# Results

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



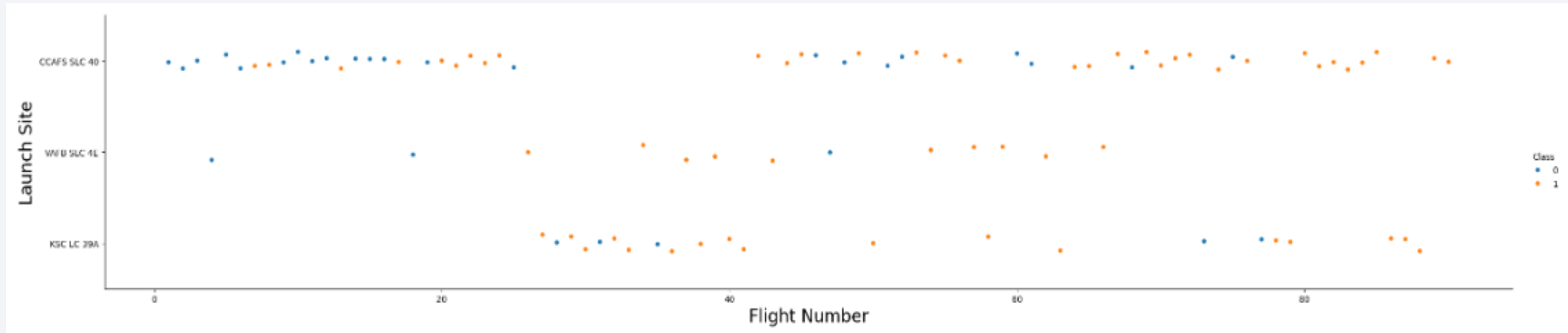


Section 2

# Insights drawn from EDA



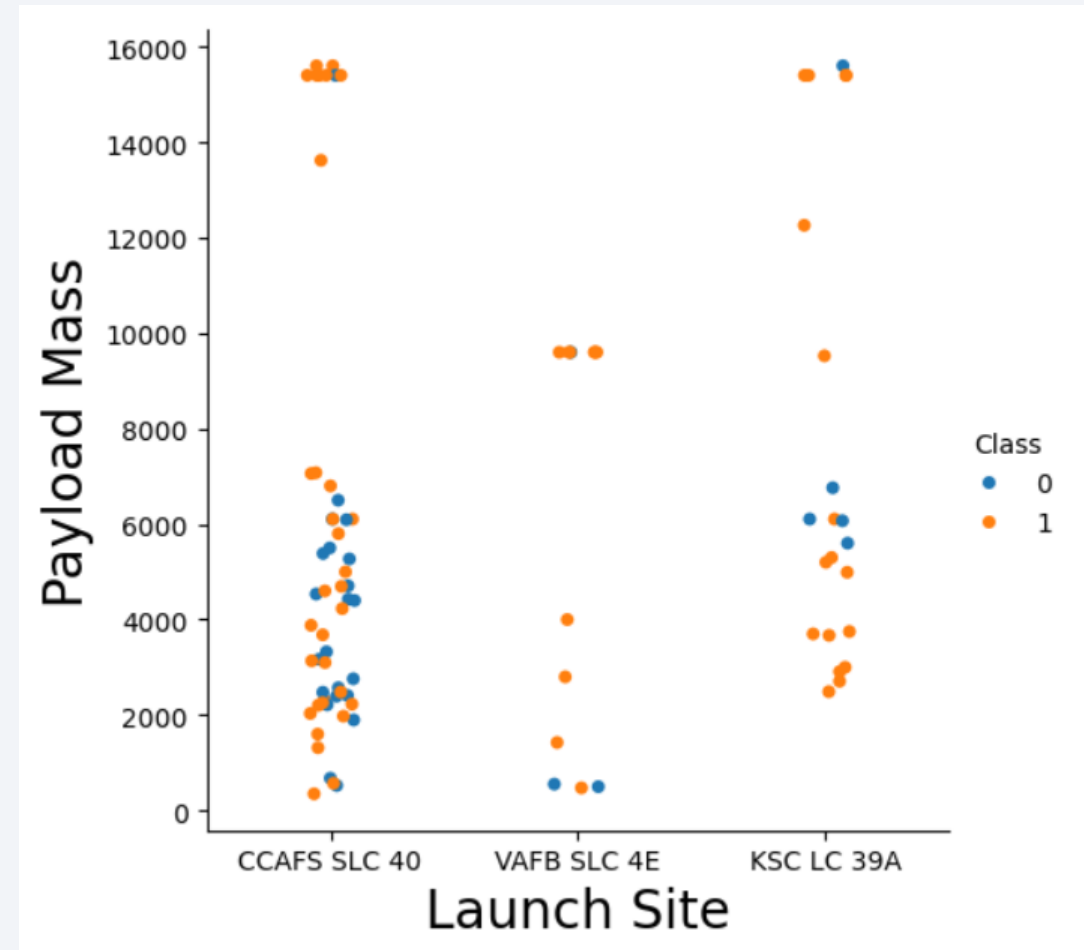
# Flight Number vs. Launch Site



- As the flight number increases, we could see the Success launches denoted by **orange dots increase** and towards the end we could see it **increases steeply** towards success for all the launch sites.
  - NB - The flight number increases with each launch. So, we could measure progress in time using the flight number.

# Payload vs. Launch Site

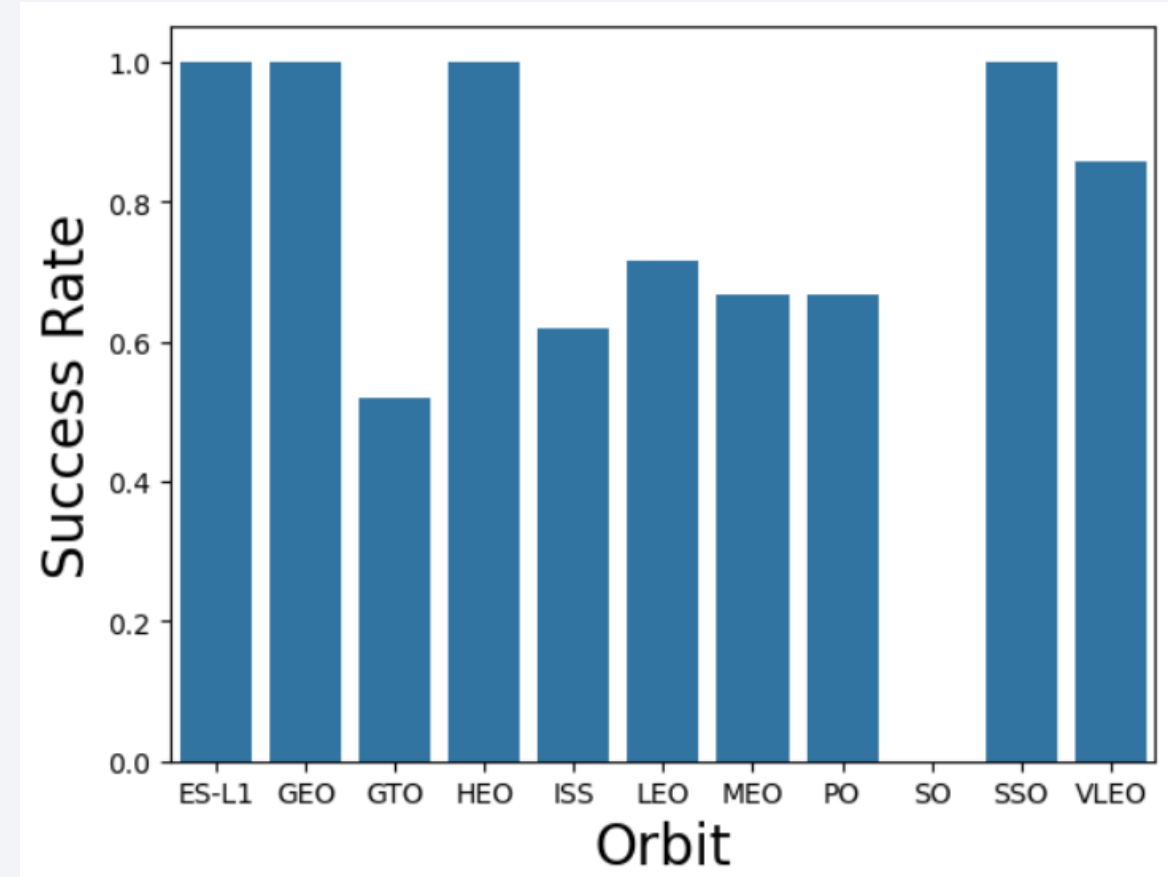
- For high payloads the success rate is quite high
- For the launch site, KSC LC-39A, even for very low payload success rate is very high



# Success Rate vs. Orbit Type

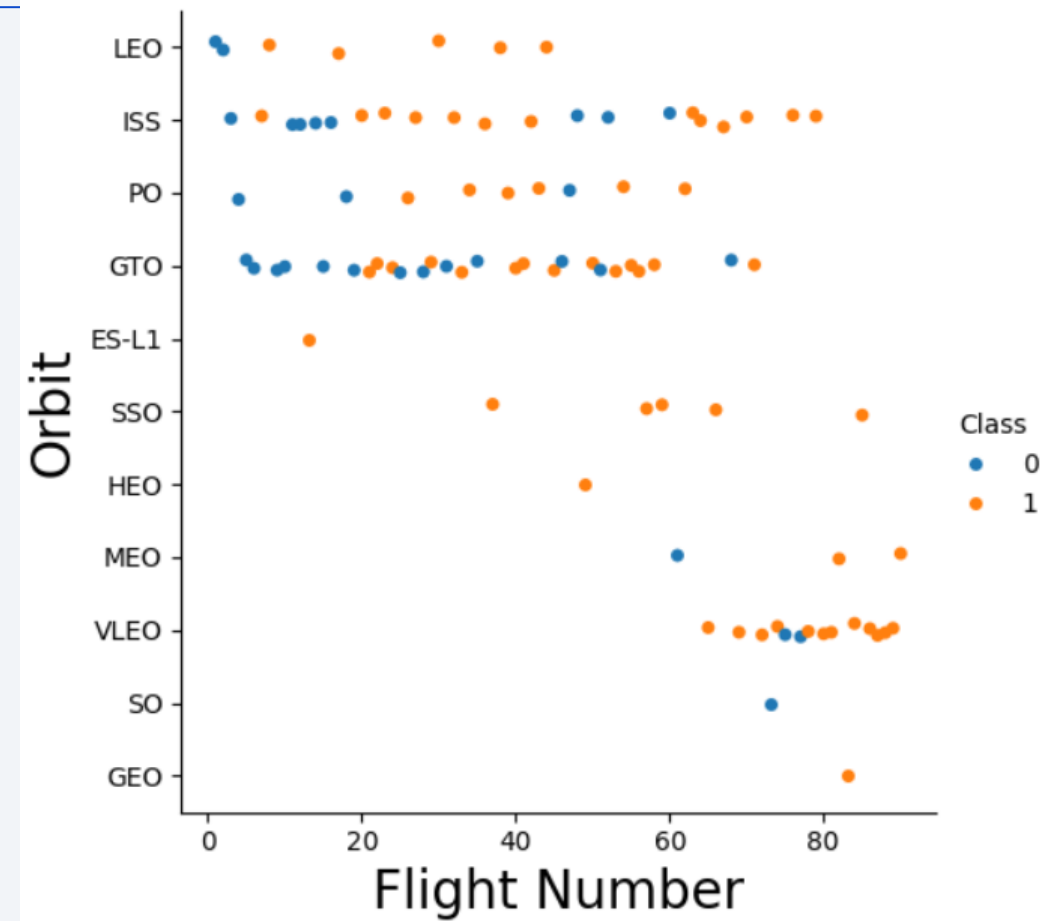
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- For the orbits ES-L1, GEO, HEO and SSO the success rate is 100%
- VLEO and LEO follows with approximately 80% and 70%
- For GTO orbit the success rate is below 50%



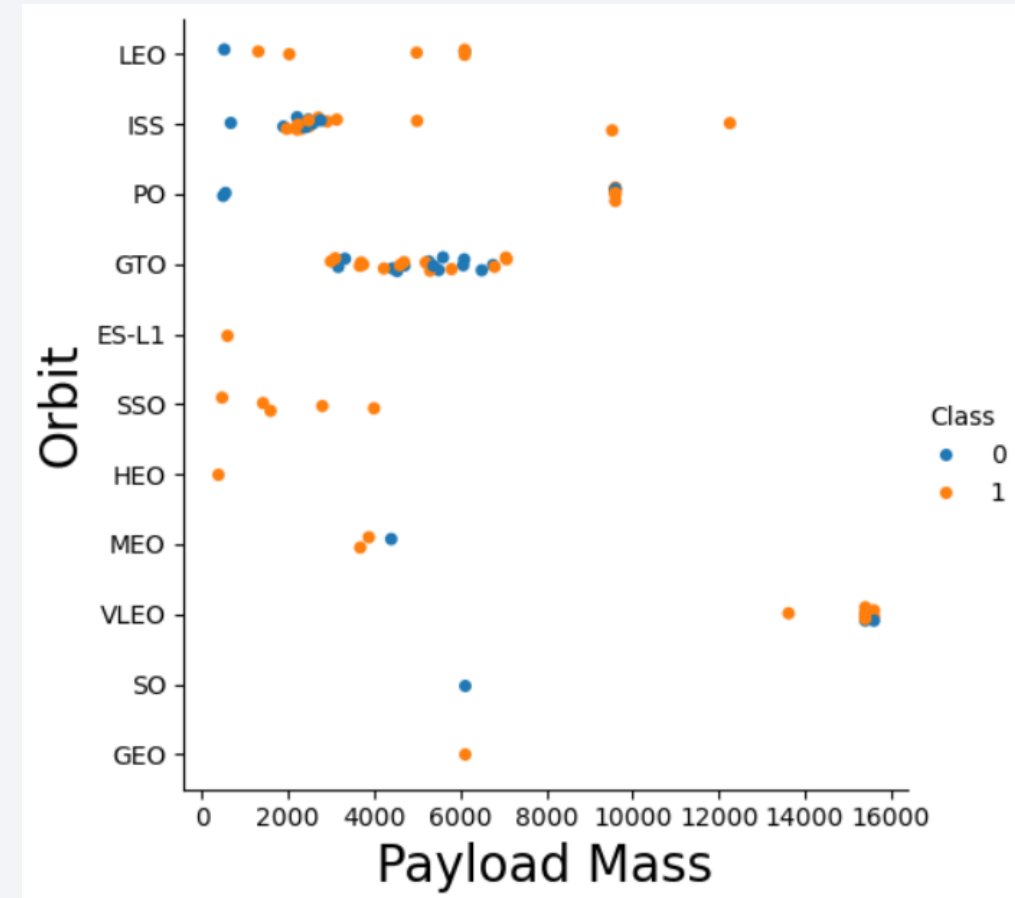
# Flight Number vs. Orbit Type

- Towards the end of flight number range, the success rate is 100% for the all Orbits
- In the starting of launches, generally many failures represented by blue dots can be seen
- But steadily the orange dot density increases for all orbits.



# Payload vs. Orbit Type

- For the orbits, ES-L1, SSO and HEO, there is low payload and 100% success rate
- For orbits, LEO, ISS and PO, lower the payload, higher the failures
- For very high payload success rate for orbits, LEO, ISS and PO success rate is 100%

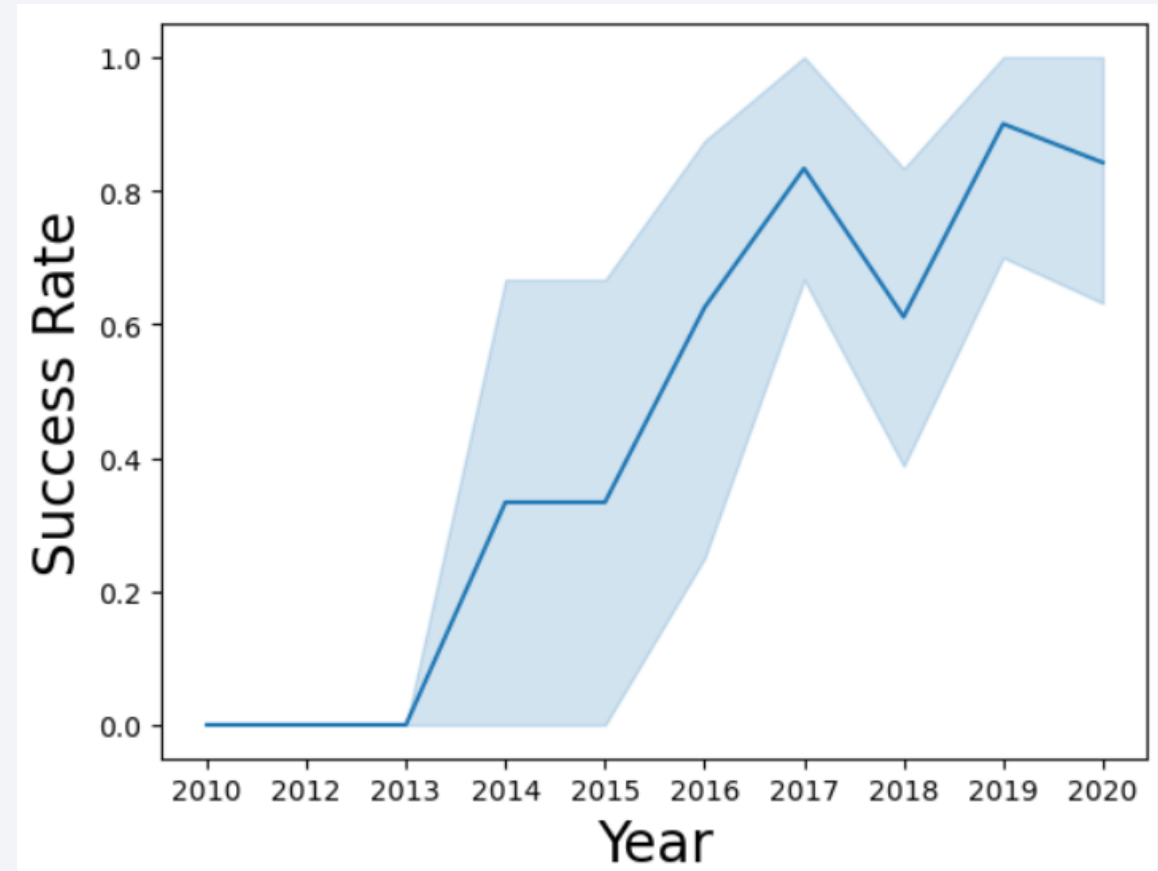




# Launch Success Yearly Trend

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- We could see there are **no successes** till the year **2013**
- After **2013** till **2017** there is very steep increase in success rate from **0** to **85%**
- In year **2019** the highest overall success rate of around **90%** can be seen



# Launch Sites

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```
%sql SELECT DISTINCT Launch_Site FROM SPACEXTABLE
* sqlite:///my_data1.db
Done.
```

Launch_Site
CCAFS LC-40
VAFB SLC-4E
KSC LC-39A
CCAFS SLC-40

- **DISTINCT** keyword is used to get the Launch Sites without duplicates from all the Launch Sites selected

# Launch Site Names starting with 'CCA'

```
%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

- **LIKE** keyword is used with the query string including % symbol to select all Launch Sites which start with CCA
- **LIMIT** keyword is used to limit the row number to 5

# Total Payload Mass

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```
%sql SELECT SUM(PAYLOAD_MASS__KG_) TOTAL_PAYLOAD_MASS FROM SPACEXTABLE where Customer='NASA (CRS)'
```

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

```
Done.
```

TOTAL_PAYLOAD_MASS
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45596
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- To select rows corresponding to customer NASA we use the string for NASA in the WHERE clause
- SUM() keyword is used to get the total of the Payload Mass launched for the customer

# Average Payload Mass by F9 v1.1

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```
%sql SELECT AVG(PAYLOAD_MASS__KG_) AVG_PAYLOAD_MASS FROM SPACEXTABLE where Booster_Version='F9 v1.1'
```

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

AVG_PAYLOAD_MASS
2928.4

- To select rows for **Booster Version F9 v1.1** the comparison string is used in the WHERE clause
- **AVG()** keyword is used to calculate the average of the Payload Mass



# First Successful Ground Landing Date

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```
%sql SELECT MIN(Date) EARLIEST_GROUNDPAD_LANDING FROM SPACEXTABLE where Landing_Outcome = 'Success (ground pad)'
```

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

```
Done.
```

```
EARLIEST_GROUNDPAD_LANDING
```

```
2015-12-22
```

- We use the **MIN()** keyword for the date column to obtain the earliest ground pad landing date.
- We filter the **Landing Outcomes** to **Success (ground pad)** in WHERE clause
- Result was **22<sup>nd</sup> Dec 2022** as earliest successful landing on ground pad

# Successful Drone Ship Landing Payload 4000 - 6000

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

```
%sql SELECT DISTINCT(Booster_Version) BOOSTERS FROM SPACEXTABLE where Landing_Outcome = 'Success (drone ship)'  
AND PAYLOAD_MASS__KG_ BETWEEN 4000 AND 6000
```

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

## BOOSTERS

F9 FT B1022

F9 FT B1026

F9 FT B1021.2

F9 FT B1031.2

- In the **WHERE** clause, we use **BETWEEN** keyword to obtain rows with the specified **payload mass** and Success (drone ship) as **Landing Outcome**
- We use **DISTINCT** keyword to avoid duplicates

# Total Number of Successful and Failure Mission Outcomes

```
%%sql SELECT 'SUCSESSES' Mission_Outcome, COUNT(Mission_Outcome) AS COUNT FROM SPACEXTABLE
WHERE Mission_Outcome LIKE 'Success%'
UNION ALL
SELECT 'FAILURES' Mission_Outcome, COUNT(Mission_Outcome) FROM SPACEXTABLE
WHERE Mission_Outcome LIKE 'Failure%'
```

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

Mission_Outcome	COUNT
SUCSESSES	100
FAILURES	1

- We analyze the **Mission Outcomes** using **LIKE** keyword and comparing the respective keywords
- We find the count using the **COUNT()** keyword
- Finally, the whole result is obtained using **UNION ALL** which shows only **1 failure** and **100 successes**

# Boosters Carried Maximum Payload

```
%%sql SELECT DISTINCT(Booster_Version), PAYLOAD_MASS_KG_ FROM SPACEXTABLE
WHERE PAYLOAD_MASS_KG_ =
(SELECT MAX(PAYLOAD_MASS_KG_) as HIGHEST_MASS FROM SPACEXTABLE)
```

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

Booster_Version	PAYLOAD_MASS_KG_
F9 B5 B1048.4	15600
F9 B5 B1048.5	15600
F9 B5 B1049.4	15600
F9 B5 B1049.5	15600
F9 B5 B1049.7	15600
F9 B5 B1051.3	15600
F9 B5 B1051.4	15600
F9 B5 B1051.6	15600
F9 B5 B1056.4	15600
F9 B5 B1058.3	15600
F9 B5 B1060.2	15600
F9 B5 B1060.3	15600

- Using the **MAX()** function in subquery we determine the highest payload which came out to be **15600 kg**
- We will find the **Booster Versions** which carry this load without duplicates with **DISTINCT** keyword

# 2015 Launch Records

```
%sql SELECT substr(Date, 6,2) as month, Landing_Outcome, Booster_Version, Launch_Site  
FROM SPACEXTABLE  
WHERE Landing_Outcome='Failure (drone ship)' and substr(Date,0,5)='2015'
```

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

month	Landing_Outcome	Booster_Version	Launch_Site
01	Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
04	Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40

- To select rows for Failure using drone ship we use query string comparison with the 'Landing\_Outcome' column
- `substr()` function is used to extract 'year' from date and we compare it with string '2015' to get corresponding rows
- We could see there were 2 failures in year 2015, one each in months of January and April

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

```
%%sql SELECT Landing_Outcome, COUNT(Landing_Outcome) AS Count FROM SPACEXTABLE
WHERE Date BETWEEN '2010-06-04' and '2017-03-20'
GROUP BY Landing_Outcome
ORDER BY Count DESC
```

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

Landing_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

- **BETWEEN** keyword is used in **WHERE** clause to get rows in the **date range**
- **GROUP BY** is used to group the rows on Landing Outcome and we calculate the corresponding count using **COUNT()** keyword
- To sort in descending order, we use **DESC** keyword with **ORDER BY**

A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The background is a deep blue gradient.

Section 3

# Launch Sites Proximities Analysis



# Launch Sites

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- There are 4 launch sites, one in the west coast and the other three in the east coast
- Location of launch sites is near to the equator and coastlines



# Launch Site Outcomes



- Launch site success is marked with **green** and failure with **red**.
- In a **popup** we could see the **name** of the launch site
- **Clicking** on the launch sites we could see the **markers** in green and red
- We could see, success Rate is relatively high for launch sites like **KSC LC-39A**

# Proximity to Geographic Locations



- From the launch site we have drawn lines with distance marked to nearest coastline, to highway and to road from the launch site, CCAFS SLC-40
- Distance to nearest coastline is 0.87 km
- Distance to highway is approximately 0.60 km
- Distance to main road is approximately 0.98km

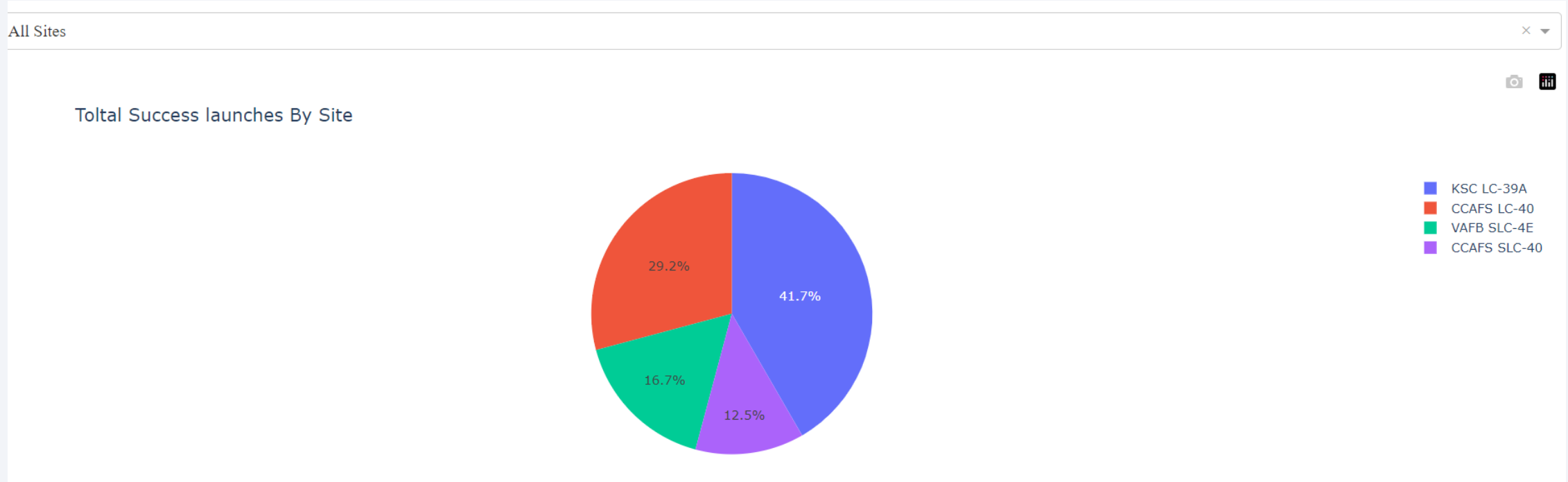




Section 4

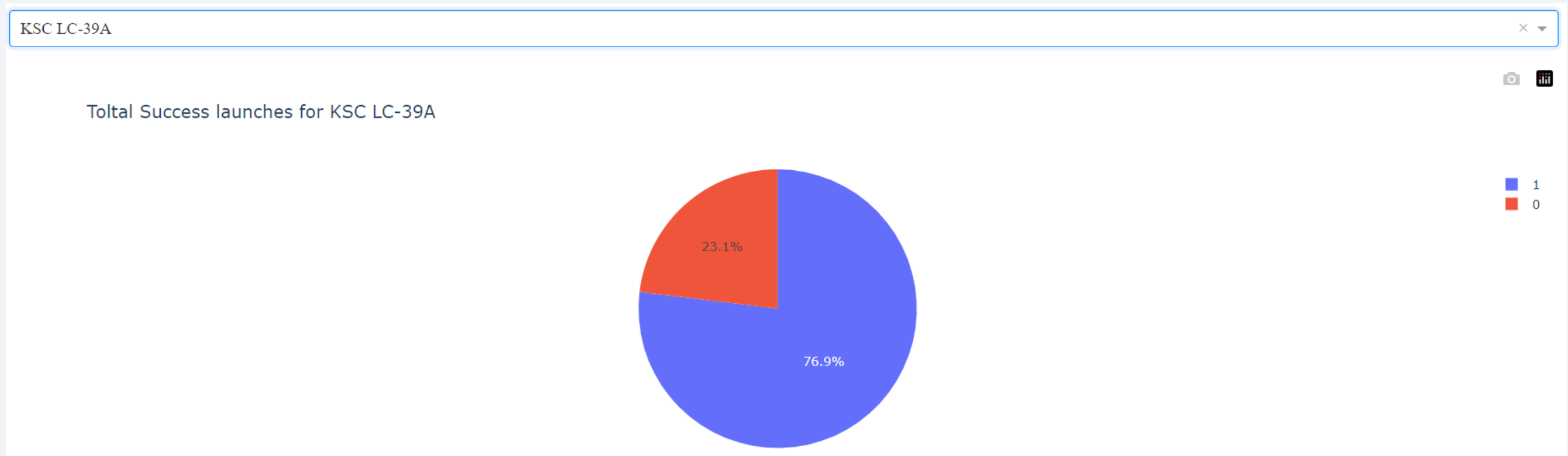
# Build a Dashboard with Plotly Dash

# Launch Success by Site



- There are 4 launch sites KSC LC-39A, CCAFS LC-40, VAFB SLC-4E and CCAFS SLC-40
- KSC LC-39A has the Highest Number of Successes with 41.7%

# Launch Site with Highest Success Rate

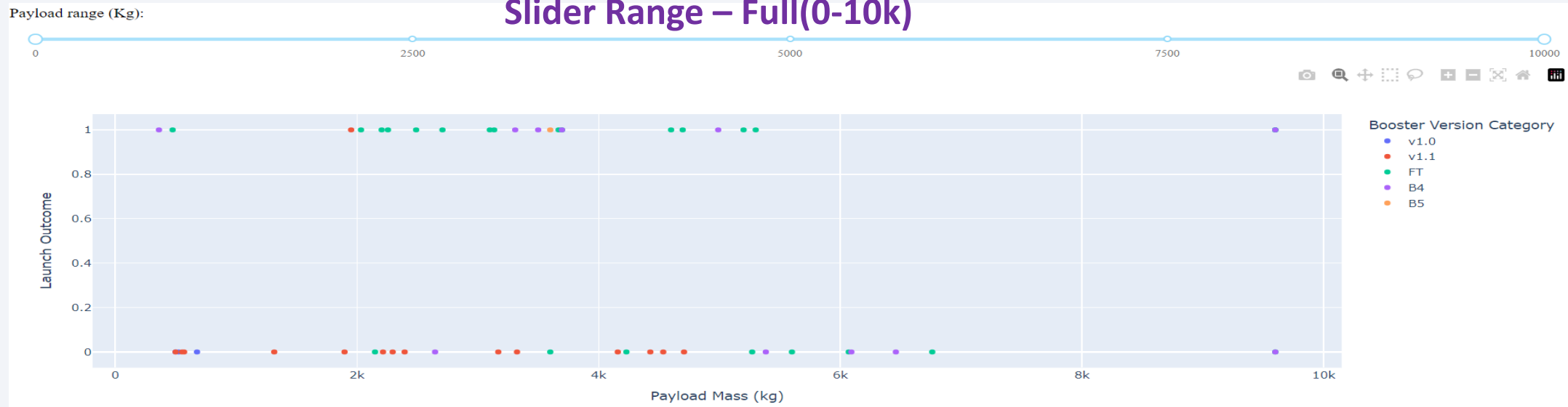


- KSC LC-39A is the Launch Site with Highest Success Rate which is 76.9%

# Payload Vs. Launch Outcome

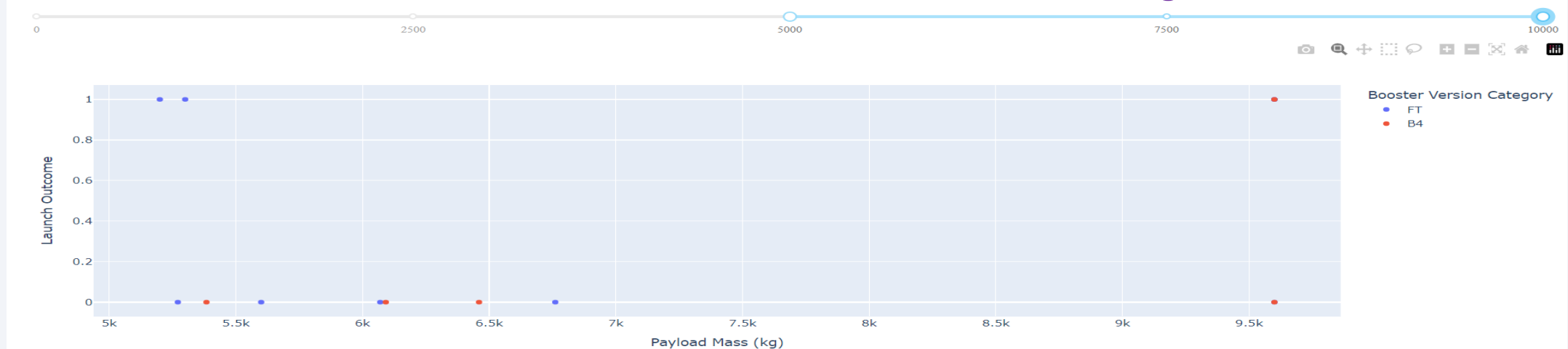
Payload range (Kg):

Slider Range – Full(0-10k)

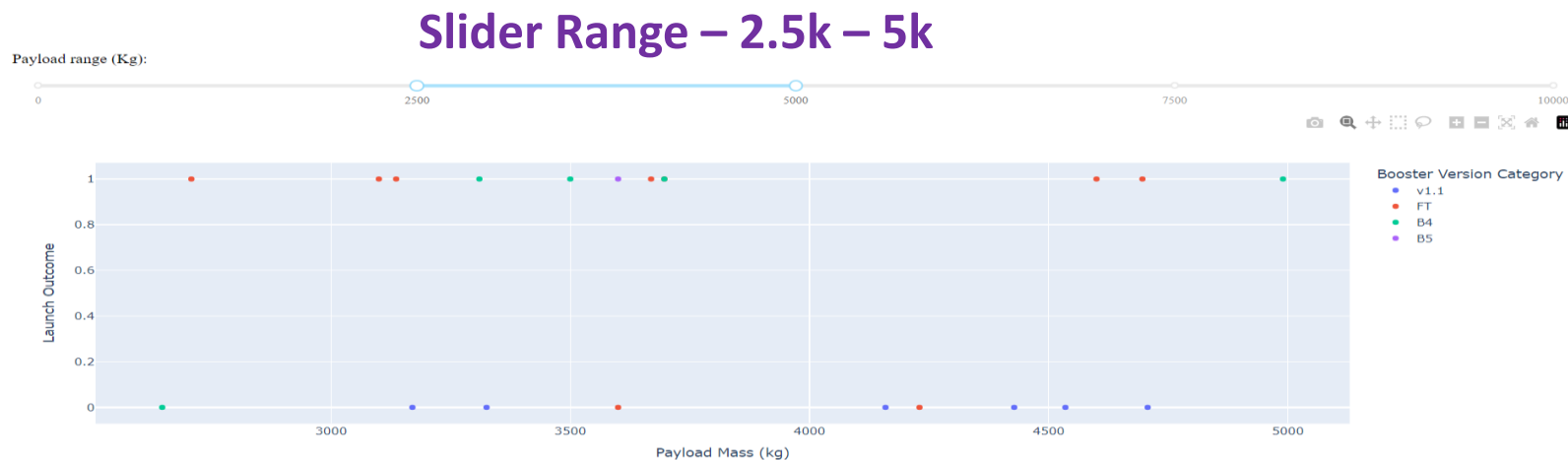
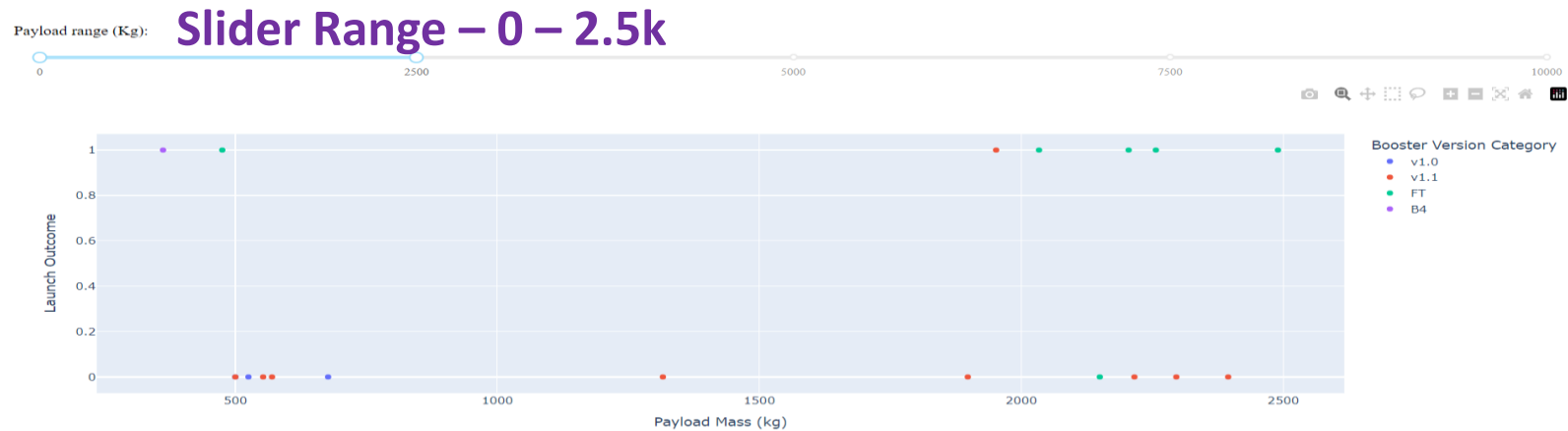


Slider Range – 5k -10k

Payload range (Kg):



# Payload Vs. Launch Outcome



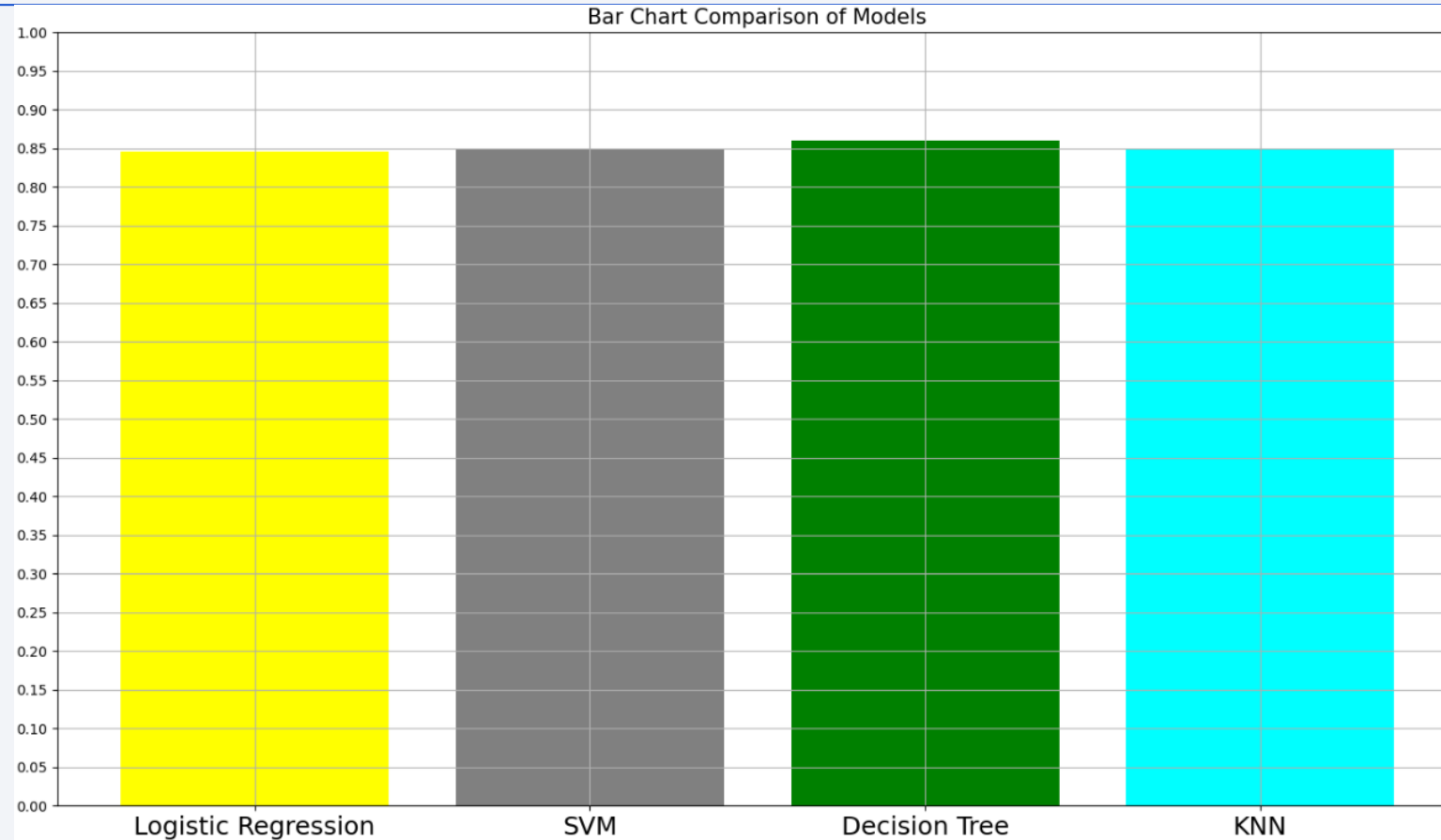
- Success rate is **very low** in the payload range **5k – 10k**
- Booster versions **BT** and **F4** are used in this payload range.
- In the range **0 – 2.5k** we could see many success for **FT** Booster Version

Section 5

# Predictive Analysis (Classification)



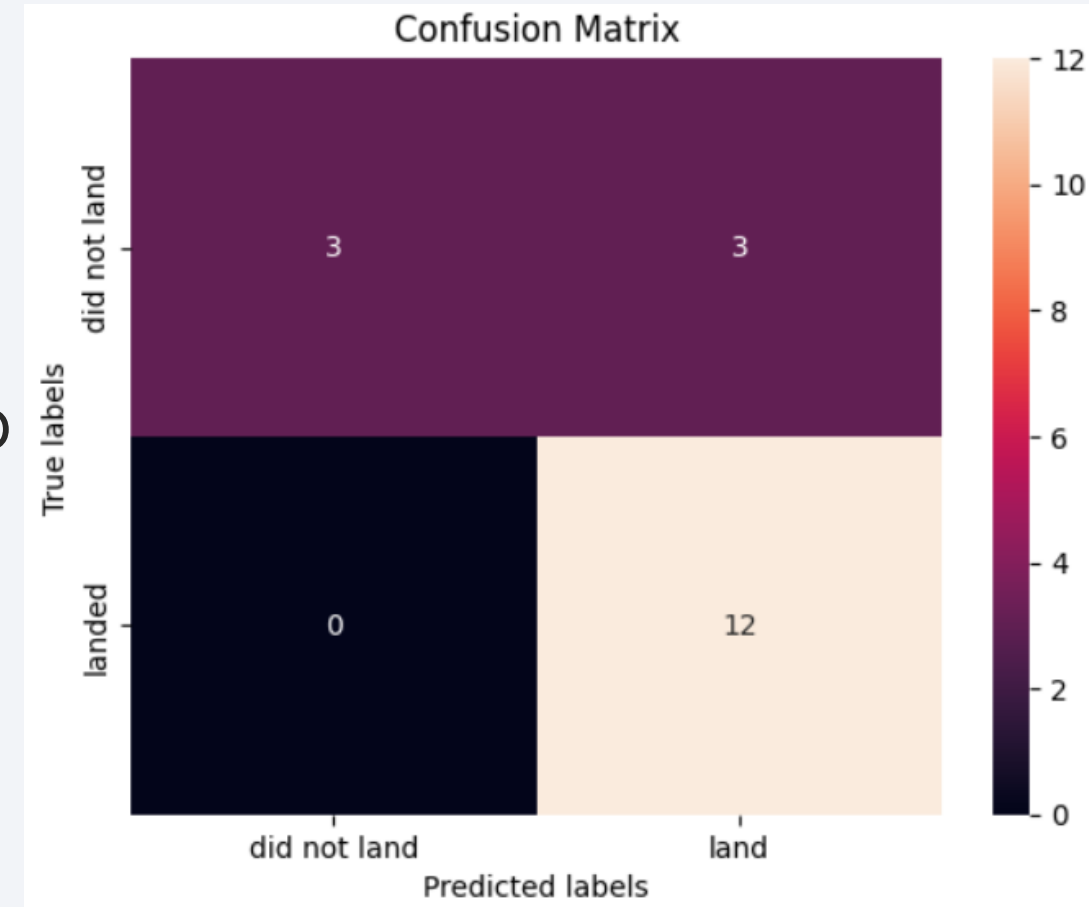
# Classification Accuracy



- **Decision Tree Model** has the highest classification accuracy

# Confusion Matrix

- Out of 18, 3 predictions were wrong
- These 3 were false positives which predicted a successful landing
- Precision (P) –  $TP / (TP + FP) = 12 / 12 + 3 = 0.80$
- Recall (R) –  $TP / (TP + FN) = 12 / (12 + 0) = 1$
- F1 score –  $2PR / (P + R)$ 
  - $2 \times (0.8 \times 1) / (0.8 + 1)$   
 $= 0.888$



# Conclusions

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- When the experiment started in with first stage return, we could see many failures. But after the few years the success rate in increased steeply to almost no failure.
- For the orbits ES-L1, GEO, HEO and SSO the success rate is 100%
- For GTO orbit the success rate is below 50%
- For orbits, LEO, ISS and PO, lower the payload, higher the failures
- KSC LC-39A is the Launch Site with Highest Success Rate
- All launch sites are in good proximity to coastlines and equator. There is good connectivity to roads, highway and railways
- Decision Tree Model had the highest classification accuracy
- As a concern, there are some false positives in prediction

# Improvements

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- Concerns
  - All the wrong predictions are Type I errors
- Improvements
  - We need to improve the model to avoid Type I errors
- Suggestions
  - Do more analysis with more data, different machine algorithms and techniques

# Appendix

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- TP – True Positive
- FP – False Positive
- FN – False Negative
- P – Precision
- R – Recall

Thank you!

