



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
    - With API
    - With Web Scrapping
  - Data Wrangling
  - Exploratory Data Analysis (EDA)
    - With SQL
    - With Visualization
    - Folium
  - Machine Learning (ML) Models
- Summary of all results
  - EDA Results
  - Interactive Analyses
  - ML Predictions

# Introduction

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- SpaceX is a company that offers new and improved rocket launches at a fraction of the price.
- A typical rocket launch used to cost at least \$165mn per launch depending upon the payload and other factors.
- SpaceX offers the same services at a base price of \$62mn. This is achieved because SpaceX reuses first stage by making it land after a launch.
- Some launches are successful and some are not.
- This project aims to find
  - Factors influencing landing outcome
  - The relationship among variables affecting the outcome
  - How to increase the successful-landing rate



Section 1

# Methodology

# Methodology

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## Executive Summary

- Data collection methodology:
  - The data was collected using SpaceX API and Web scrapping from Wikipedia
- Perform data wrangling
  - The missing payload values were replaced with mean payload.
  - A categorical variable named 'class' was added to show successful (1) and unsuccessful landing
  - All categorical variables were 'one-hot-encoded'
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models

# Data Collection

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- The data was collected from two different sources
  - SpaceX REST API
    - API EndPoint used was <https://api.spacexdata.com/v4/launches/past>
    - Requests Object was transformed into JSON string using `.json()` method
    - The JSON string was converted into Pandas DataFrame using `json_normalize()` method
  - Wikipedia via Web scrapping
- You need to present your data collection process use key phrases and flowcharts

# Data Collection – SpaceX API

```
In [6]: spacex_url="https://api.spacexdata.com/v4/launches/past"
```

```
In [7]: response = requests.get(spacex_url)
```

Now we decode the response content as a json using `.json()` and turn it into a Pandas dataframe using `.json_normalize()`

```
In [16]: # Use json_normalize method to convert the json result into a dataframe
temp = response.json()
#temp
data = pd.json_normalize(temp)
data.head()
```

```
Out[16]:
```

	static_fire_date_utc	static_fire_date_unix	net	window	rocket	success	failures	details	crew	ships	capsules
0	2006-03-17T00:00:00.000Z	1.142554e+09	False	0.0	5e9d0d95eda69955f709d1eb	False	[{'time': 33, 'altitude': None, 'reason': 'merlin engine failure'}]	Engine failure at 33 seconds and loss of vehicle			

[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-spacex-data-collection-api\\_SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-spacex-data-collection-api_SOL.ipynb)

Specify API Endpoint URL



Use Request Object to request API and assign it to response object



Convert Response to JSON string using `.json()` method



Convert JSON string to Pandas DataFrame using `.json_normalize()` method



# Data Collection - Scraping

```
In [4]: static_url = "https://en.wikipedia.org/w/index.php?title=List_of_Falcon_9_and_Falcon_Heavy_launches&oldid=1027686922"
```

```
In [10]: # use requests.get() method with the provided static_url
# assign the response to a object

response = requests.get(static_url).text
```

Create a BeautifulSoup object from the HTML response

```
In [11]: # Use BeautifulSoup() to create a BeautifulSoup object from a response text content
soup = BeautifulSoup(response, 'html.parser')
```

```
In [17]: # Use the find_all function in the BeautifulSoup object, with element type `table`
# Assign the result to a list called `html_tables`

html_tables = soup.find_all("table")
```

Specify Wikipedia URL



Use Request Object to get mentioned URL's text and assign it to response object



Use BeautifulSoup to parse the text



Find HTML Table and extract values

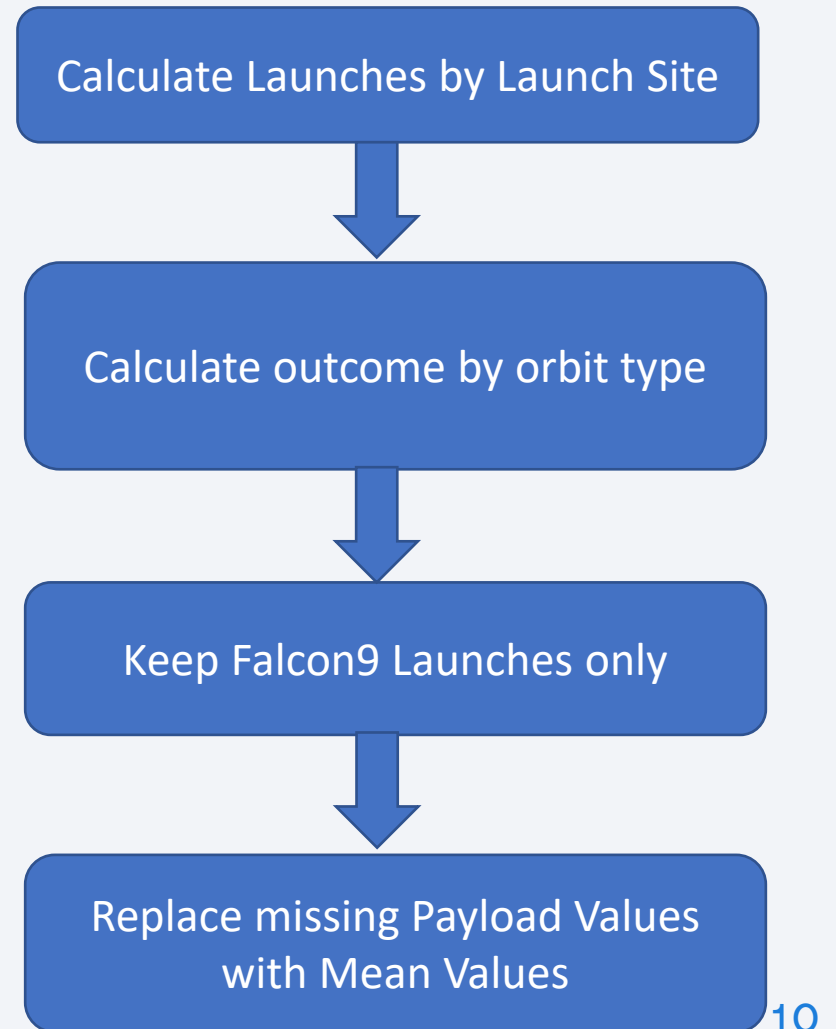
[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-webscraping\\_SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-webscraping_SOL.ipynb)

# Data Wrangling

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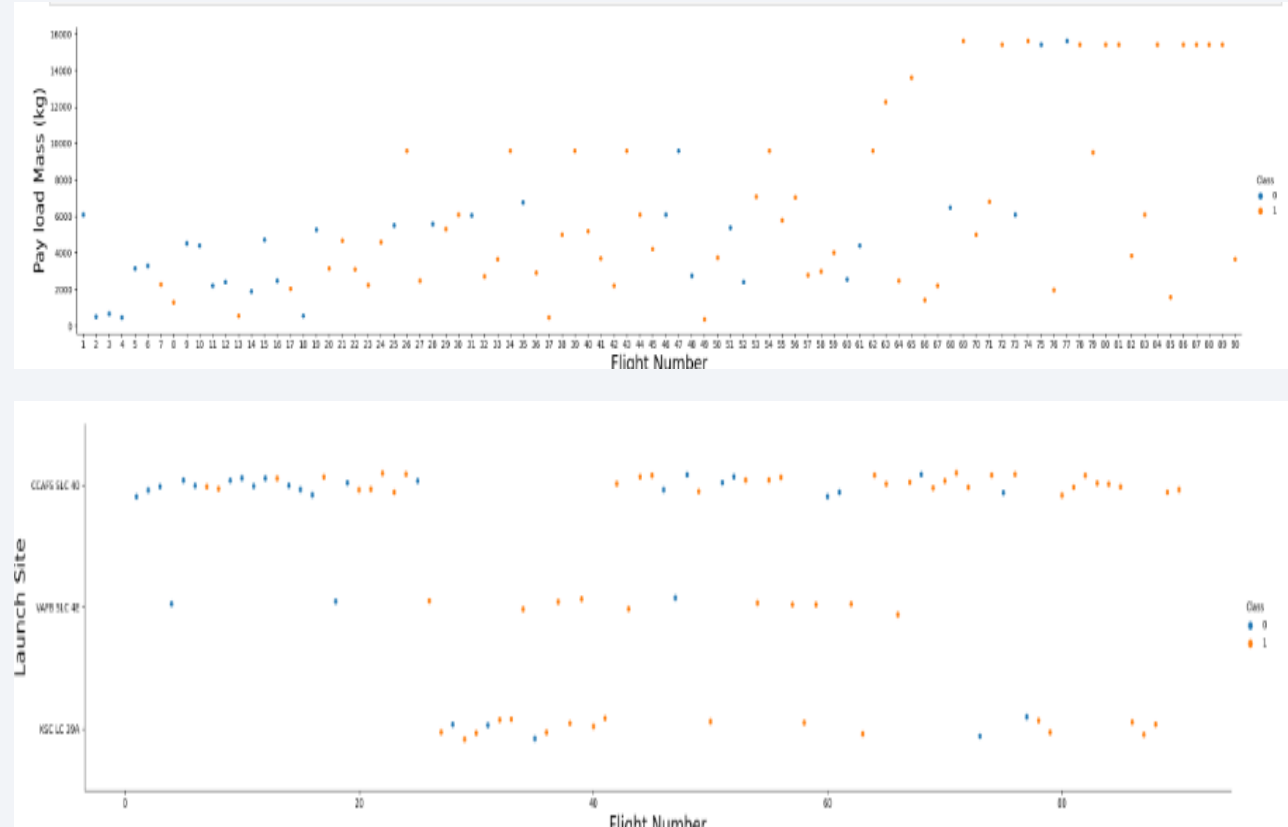
- Launches by launch site were calculated at first
- Then outcome by orbit type was calculated
- Falcon9 data is kept and the rest was discarded
- Missing PayloadMass values were replaced with mean/average values

[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/labs-jupyter-spacex-Data%20wrangling SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/labs-jupyter-spacex-Data%20wrangling%20SOL.ipynb)



# EDA with Data Visualization

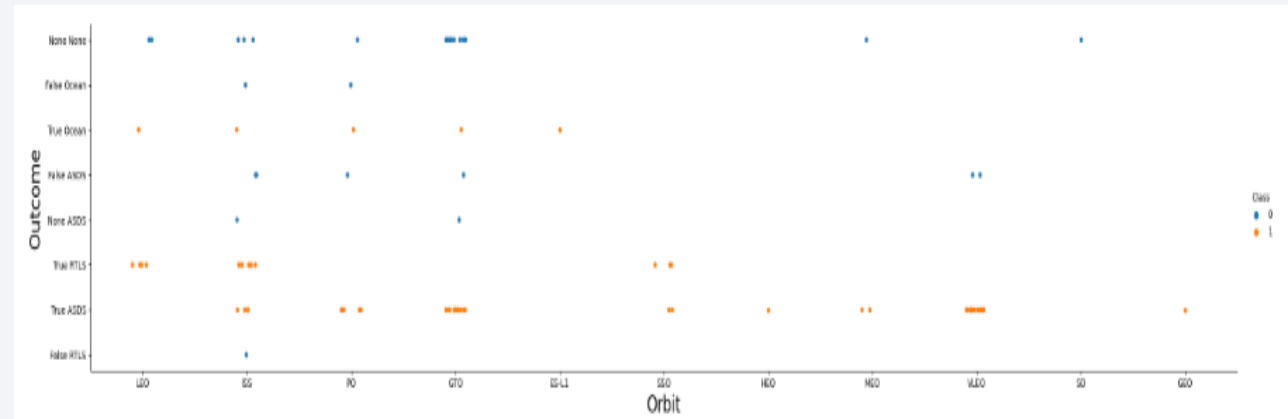
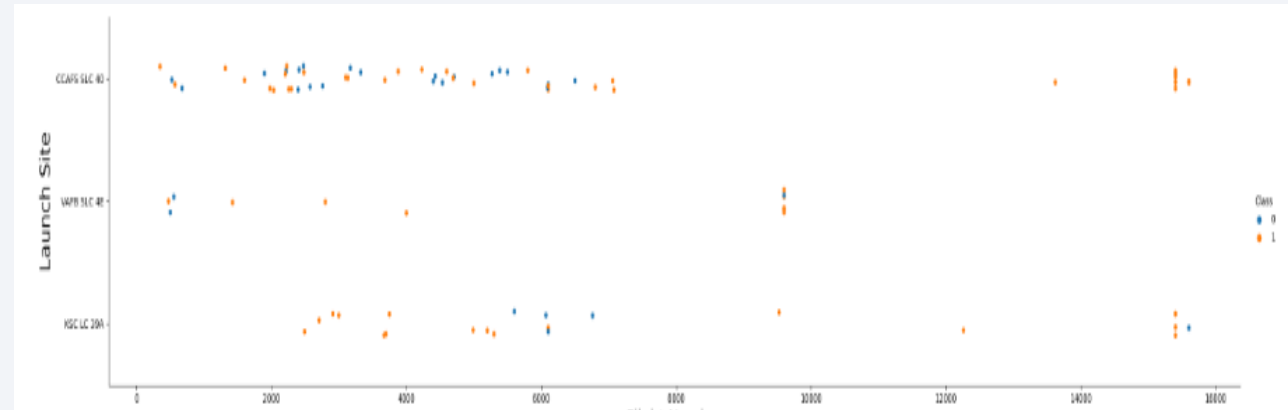
- FlightNumber vs PayloadMass
- FlightNumber vs LaunchSite
- PayloadMass vs LaunchSite
- Orbit vs Outcome
- Orbit vs FlightNumber
- Orbit vs PayloadMass



[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-dataviz\\_SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-dataviz_SOL.ipynb)

# EDA with Data Visualization

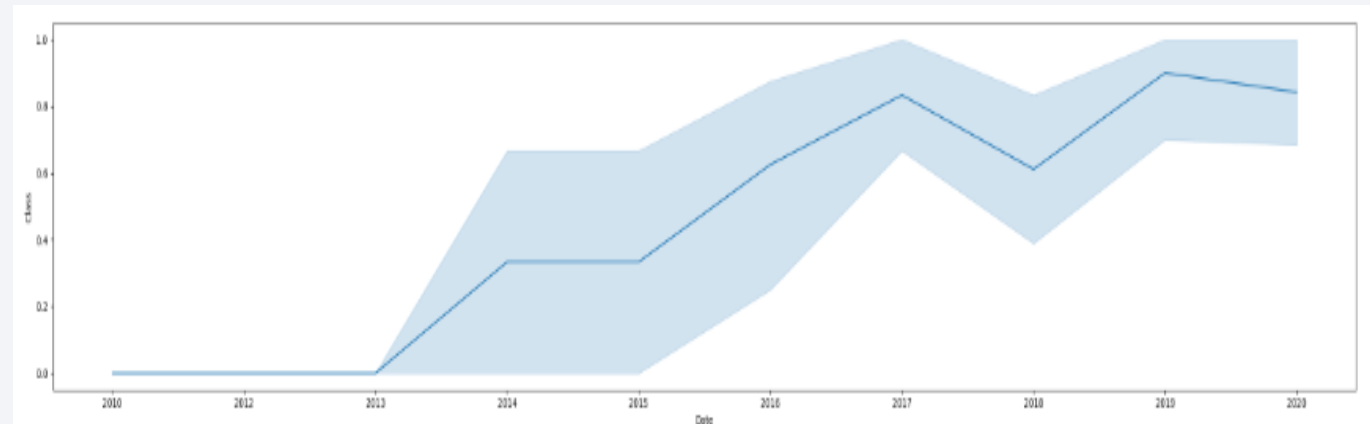
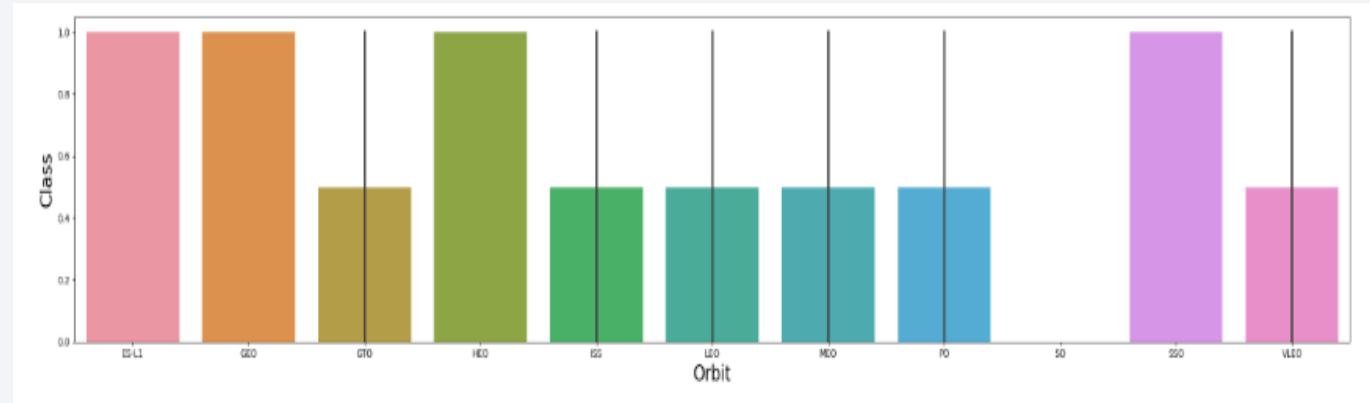
- FlightNumber vs PayloadMass
- FlightNumber vs LaunchSite
- PayloadMass vs LaunchSite
- Orbit vs Outcome
- Orbit vs FlightNumber
- Orbit vs PayloadMass



[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-dataviz\\_SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-dataviz_SOL.ipynb)

# EDA with Data Visualization

- FlightNumber vs PayloadMass
- FlightNumber vs LaunchSite
- PayloadMass vs LaunchSite
- Orbit vs Outcome
- Orbit vs FlightNumber
- Orbit vs PayloadMass



[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-dataviz\\_SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-dataviz_SOL.ipynb)



# EDA with SQL

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- Displayed

- the names of the unique launch sites in the space mission
- 5 records where launch sites begin with the string 'CCA'
- the total payload mass carried by boosters launched by NASA (CRS)
- average payload mass carried by booster version F9 v1.1
- the list and the date when the first successful landing outcome in ground pad was achieved.
- the list and the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000.
- the list and the total number of successful and failure mission outcomes
- the list and the names of the booster\_versions which have carried the maximum payload mass. Use a subquery
- the list and the records which will display the month names, failure landing\_outcomes in drone ship ,booster versions, launch\_site for the months in year 2015
- the 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

[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-sql-coursera\\_sqlite\\_SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/jupyter-labs-eda-sql-coursera_sqlite_SOL.ipynb)

# Build an Interactive Map with Folium

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- Latitude and longitudes of launch locations were marked with circular markers around each site.
- Green marker signifies successful outcome whereas the red marker signifies unsuccessful outcome
- The distance of railways, highways, and coastline from launch site was calculated

[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/lab\\_jupyter\\_launch\\_site\\_location\\_SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/lab_jupyter_launch_site_location_SOL.ipynb)

# Build a Dashboard with Plotly Dash

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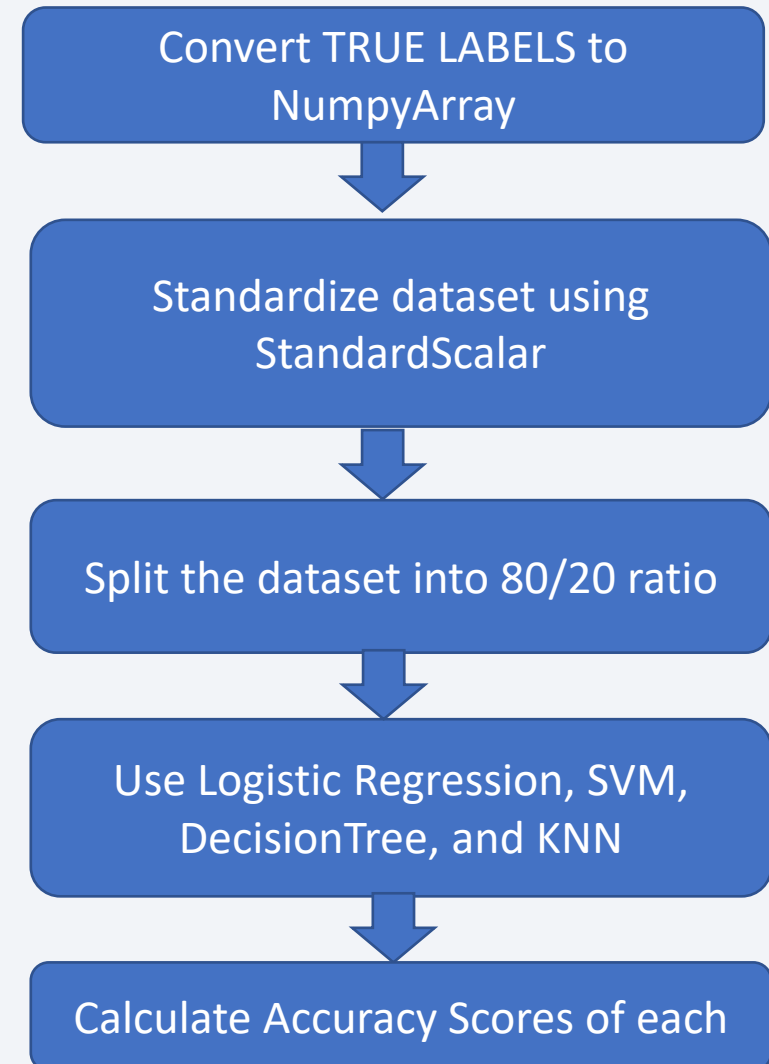
- An interactive dashboard was created
- User can extract charts based on
  - Launch site (Pie Charts)
  - Payload (Scatter Charts based on slider)

[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/spacex\\_dash\\_app\\_SOL.py](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/spacex_dash_app_SOL.py)

# Predictive Analysis (Classification)

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[https://github.com/mhnaqvi/DataScienceCapstone/blob/main/SpaceX Machine Learning Prediction Part 5.jupyterlite SOL.ipynb](https://github.com/mhnaqvi/DataScienceCapstone/blob/main/SpaceX%20Machine%20Learning%20Prediction%20Part%205.jupyterlite%20SOL.ipynb)



# Results

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- The results are categorized in the following categories
  - Exploratory data analysis results
  - Interactive analytics demo in screenshots
  - Predictive analysis results



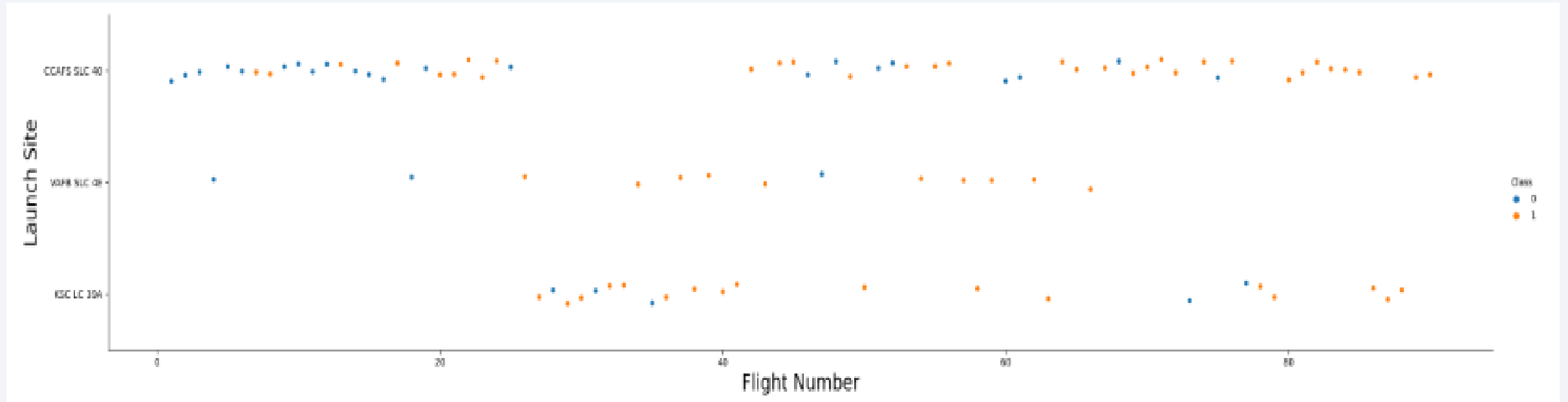
The background of the slide is an abstract composition. It features a dark blue field on the left side, which transitions into a complex pattern of diagonal streaks in shades of blue, red, and cyan on the right. These streaks have a textured, almost woven appearance. Overlaid on this pattern is a faint, light-colored grid that creates a sense of depth and structure.

Section 2

# Insights drawn from EDA

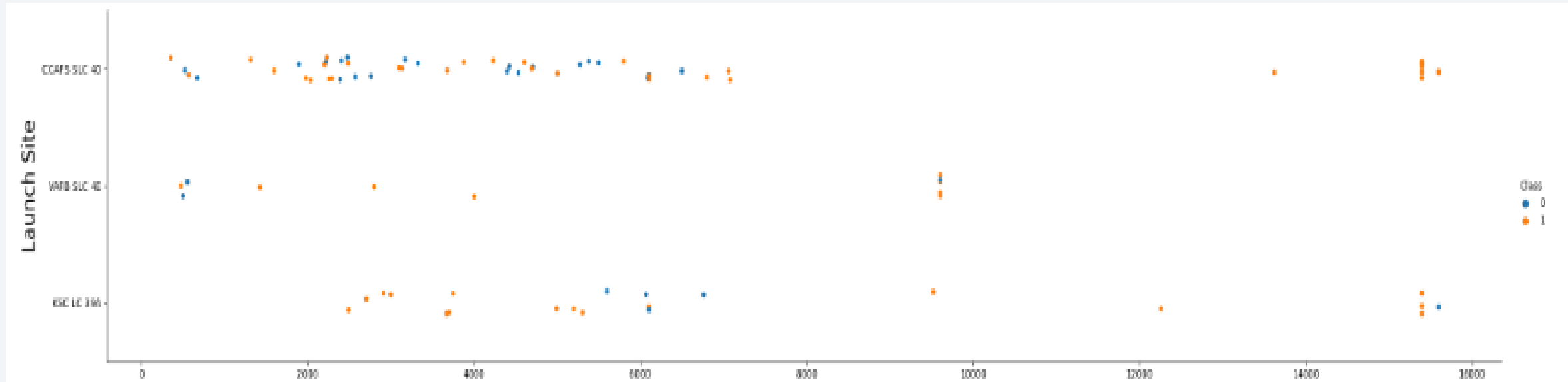


# Flight Number vs. Launch Site



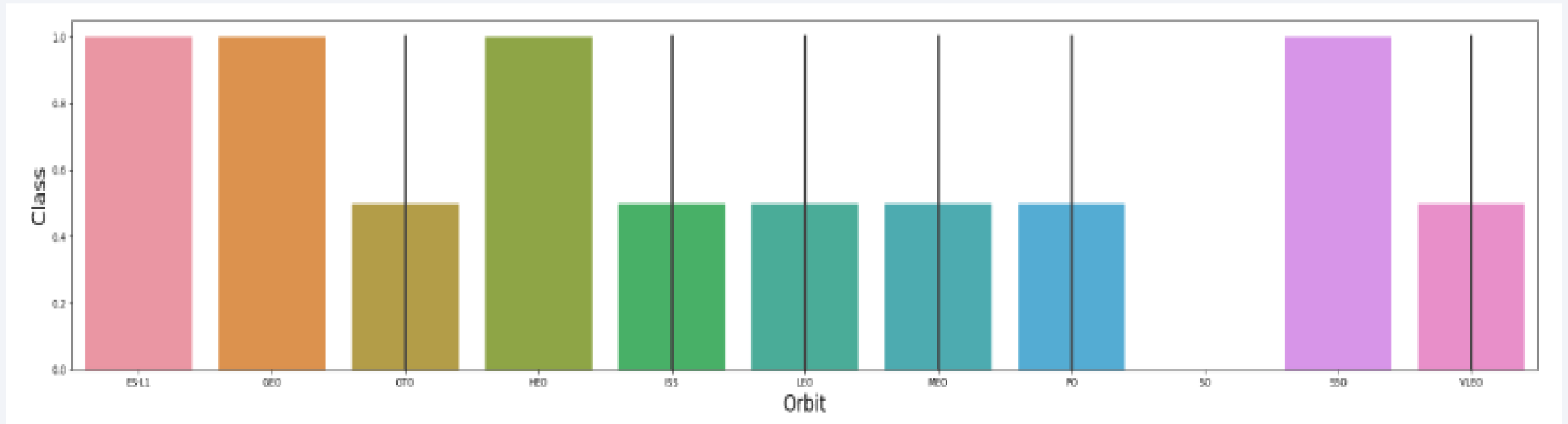
- This plot shows that the number of flights is directly proportional to launch sites.
- But, the CCAFS SLC 40 has a better record in terms of successful outcomes

# Payload vs. Launch Site



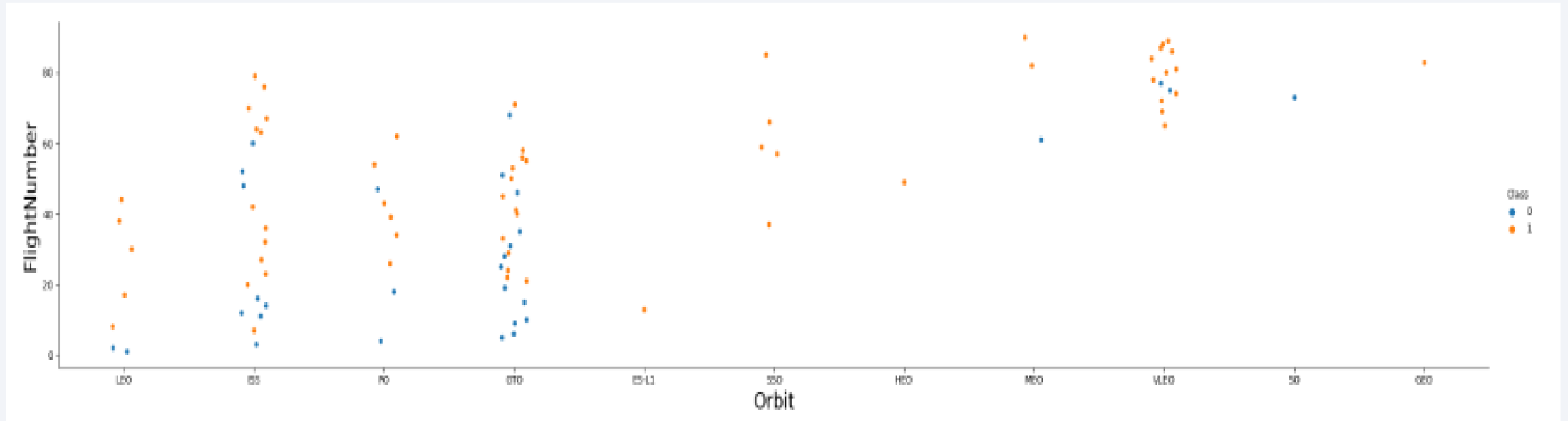
- This plot shows that as the payload mass increases the successful outcome increases.
- Launch site has no effect on successful outcome.

# Success Rate vs. Orbit Type



- No effective conclusion is drawn

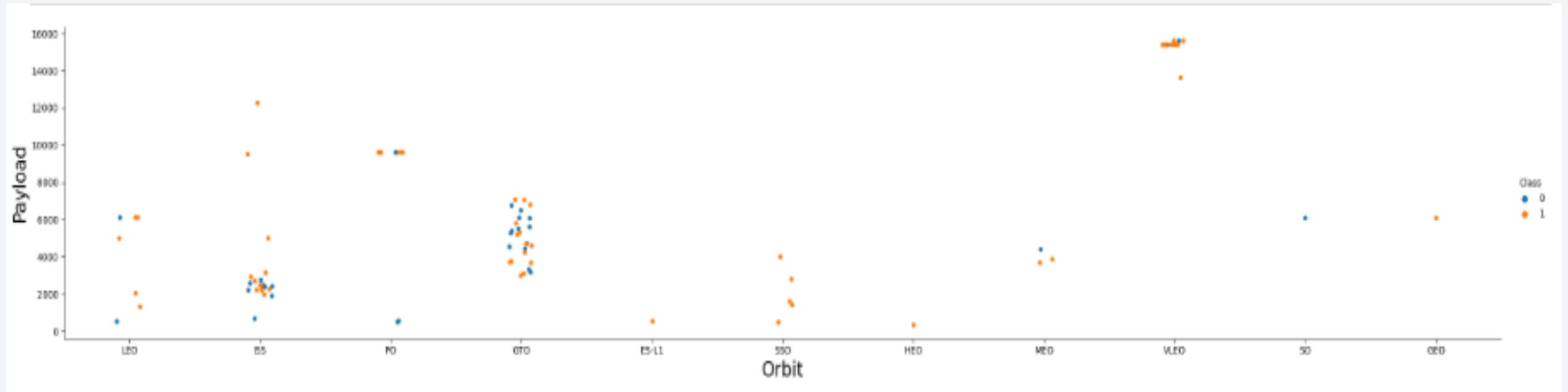
# Flight Number vs. Orbit Type



- Higher number of launches for each orbit leads to a higher number of successful outcome



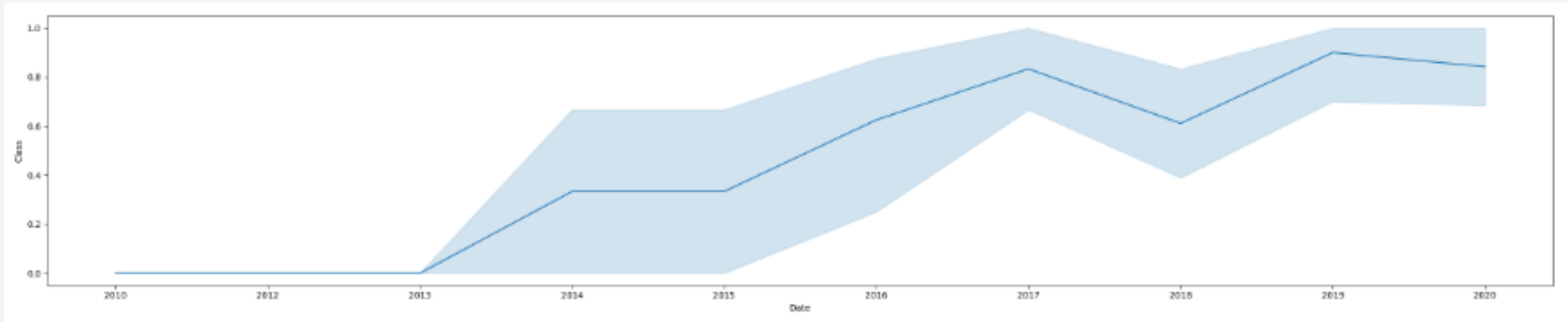
# Payload vs. Orbit Type



- Higher payload leads to successful outcome for LEO, ISS, and PO orbits. Whereas it has a negative impact on MEO and VLEO orbits
- GTO has no effect
- SO, GEO, and HEO are inconclusive

# Launch Success Yearly Trend

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- Since 2013, the success is increasing until the available date i.e. 2020

# All Launch Site Names

---

Display the names of the unique launch sites in the space mission

```
In [9]: %sql select distinct(Launch_Site) from SPACEXTBL
```

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

```
Done.
```

```
Out[9]: Launch_Site
```

```
CCAFS LC-40
```

```
VAFB SLC-4E
```

```
KSC LC-39A
```

```
CCAFS SLC-40
```

```
None
```

- Distinct is used to display the site name exactly once

# Launch Site Names Begin with 'CCA'

Display 5 records where launch sites begin with the string 'CCA'

```
In [11]: %sql select * from SPACEXTBL where Launch_site like 'CCA%' limit 5
```

\* sqlite:///my\_data1.db

Done.

```
Out[11]:
```

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG_	Orbit	Customer	Mission_Outcome	Landing_Outc
------	------------	-----------------	-------------	---------	------------------	-------	----------	-----------------	--------------

06/04/2010	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0.0	LEO	SpaceX	Success	Failure (parachute)
------------	----------	---------------	-------------	---	-----	-----	--------	---------	---------------------

12/08/2010	15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0.0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
------------	----------	---------------	-------------	---	-----	--------------	-----------------------	---------	---------------------

22/05/2012	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525.0	LEO (ISS)	NASA (COTS)	Success	No attachment
------------	---------	---------------	-------------	-----------------------------	-------	--------------	----------------	---------	---------------

10/08/2012	0:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500.0	LEO (ISS)	NASA (CRS)	Success	No attachment
------------	---------	---------------	-------------	-----------------	-------	--------------	---------------	---------	---------------

03/01/2013	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677.0	LEO (ISS)	NASA (CRS)	Success	No attachment
------------	----------	---------------	-------------	-----------------	-------	--------------	---------------	---------	---------------

- Five records starting where the launch site name starts from are displayed

# Total Payload Mass

---

## Task 3

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

```
In [16]: %sql select sum(PAYLOAD_MASS_KG_) from SPACEXTBL where Customer='NASA (CRS)'
```

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

```
Done.
```

```
Out[16]: sum(PAYLOAD_MASS_KG_)  
         45596.0
```

- Total payload mass where the client is NASA (CRS) is calculated



# Average Payload Mass by F9 v1.1

---

## Task 4

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

```
In [17]: %sql select avg(PAYLOAD_MASS_KG_) from SPACEXTBL where Booster_version='F9 v1.1'
* sqlite:///my_data1.db
Done.
Out[17]: avg(PAYLOAD_MASS_KG_)
          2928.4
```

- The average payload mass carried by booster version F9 v1.1 is calculated

# 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*

```
In [28]: %sql select min(DATE) from SPACEXTBL where Landing_Outcome='Success (ground pad)'
```

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

```
Out[28]: min(DATE)  
01/08/2018
```

- The date of the first successful landing outcome on ground pad is calculated

# Successful Drone Ship Landing with Payload between 4000 and 6000

## 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

```
In [30]: %sql select Booster_version from SPACEXTBL where Landing_Outcome = 'Success (drone ship)' and PAYLOAD_MASS__KG_ > 4000 and P
* sqlite:///my_data1.db
Done.
```

Out[30]: **Booster\_Version**

F9 FT B1022

F9 FT B1026

F9 FT B1021.2

F9 FT B1031.2

- Lists the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

# Total Number of Successful and Failure Mission Outcomes

## Task 7

List the total number of successful and failure mission outcomes

```
In [34]: %sql select * from SPACEXTBL limit 5
%sql select count(mission_outcome) from SPACEXTBL where mission_outcome = 'Success' or mission_outcome='Failure'
```

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

```
Out[34]: count(mission_outcome)
          98
```

- This query displays the total number of successful and total number of failures

# Boosters Carried Maximum Payload

## Task 8

List the names of the booster\_versions which have carried the maximum payload mass. Use a subquery

```
In [35]: %sql select * from SPACEXTBL limit 5
%sql select distinct(booster_version) from SPACEXTBL where PAYLOAD_MASS_KG_ = (select max(PAYLOAD_MASS_KG_) from SPACEXTBL)
```

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

```
Out[35]: 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

- The names of the booster which have carried the maximum payload mass are listed

# 2015 Launch Records

## 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, 4, 2) as month to get the months and substr(Date,7,4)='2015' for year.**

```
In [50]: %sql select * from SPACEXTBL
%sql select substr(Date, 4,2) as Month, substr(Date, 7,4) as Year, Landing_outcome, booster_version, launch_site from SPACEXTBL

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

Out[50]:
```

Month	Year	Landing_Outcome	Booster_Version	Launch_Site
10	2015	Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
04	2015	Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40
01	2016	Failure (drone ship)	F9 v1.1 B1017	VAFB SLC-4E
04	2016	Failure (drone ship)	F9 FT B1020	CCAFS LC-40
06	2016	Failure (drone ship)	F9 FT B1024	CCAFS LC-40

- The query displays the list of the failed landing\_outcomes in drone ship, their booster versions, and launch site names for since 2015

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

## 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.

```
In [58]: %sql select count(landing_outcome) from SPACEXTBL where Date between '04/06/2010' and '20/03/2017' order by date desc
```

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

```
Out[58]: count(landing_outcome)  
          _____  
          55
```

- 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

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.

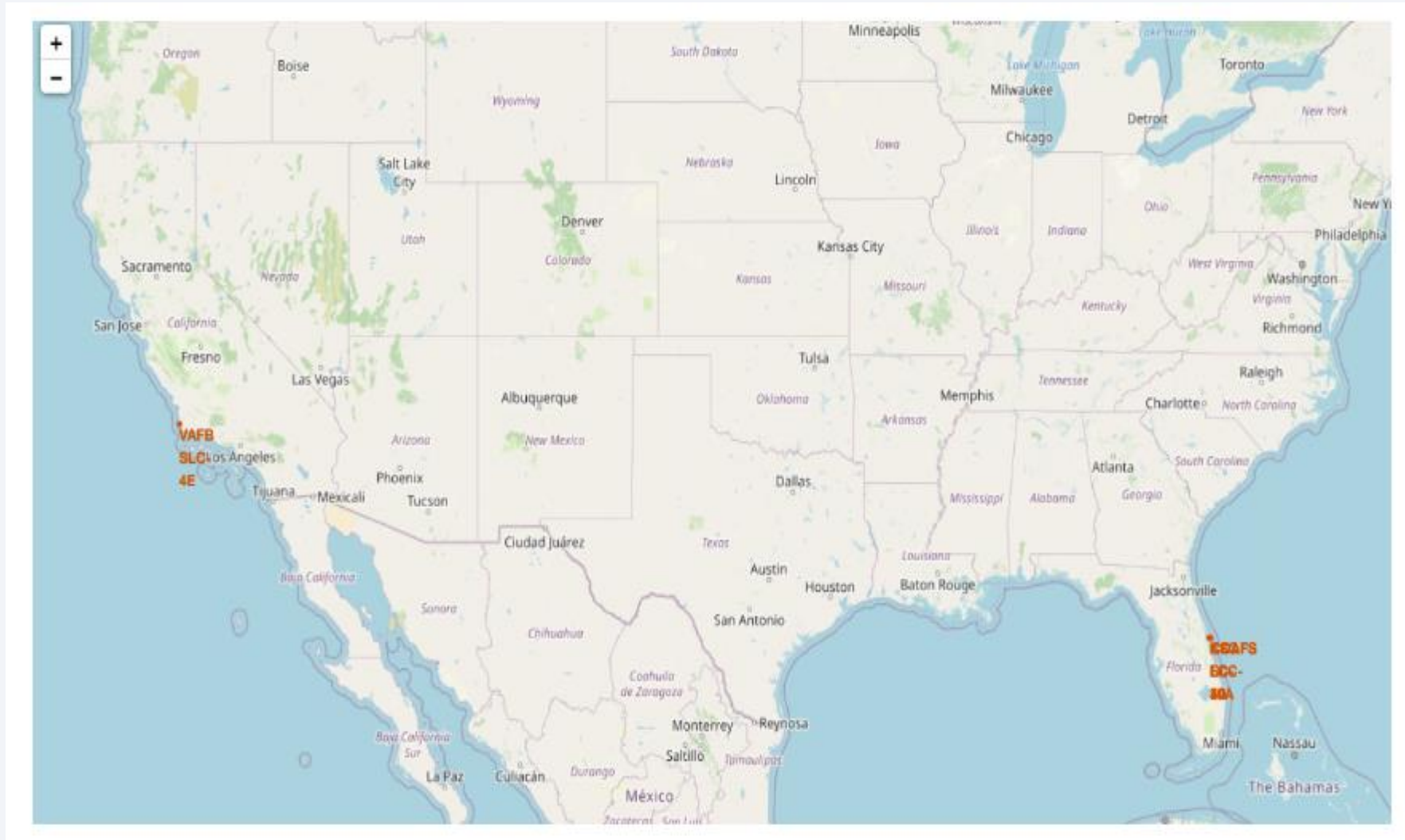
Section 3

# Launch Sites Proximities Analysis



# Launch Sites Locations

- This folium chart shows the SpaceX's launch sites in United States



# Marked Folium Map



- This map shows the marked sites. Green for successful landings and red for unsuccessful landings

# Distance Folium Map



- Distance to coastline



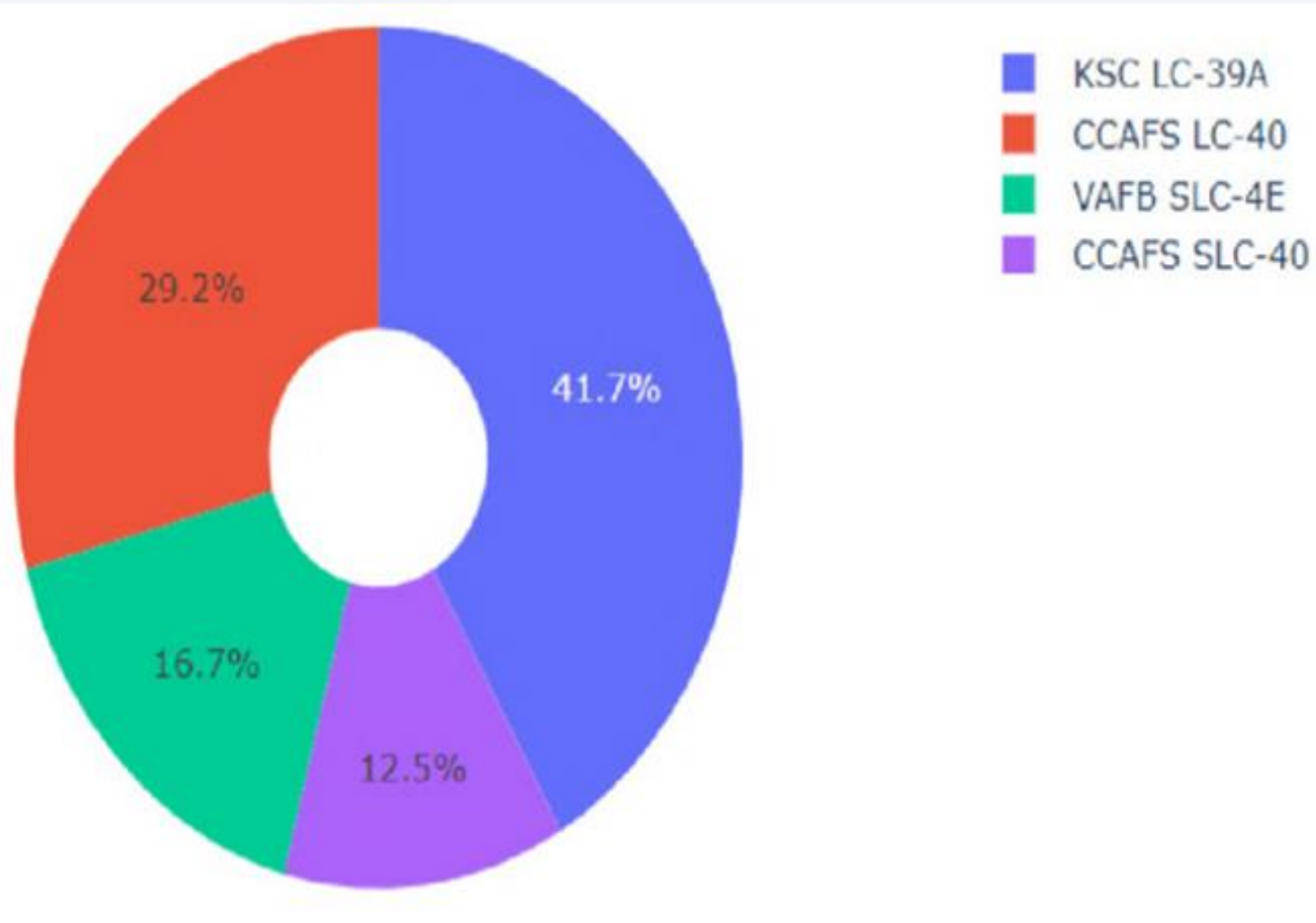


Section 4

# Build a Dashboard with Plotly Dash

# Site-wise Success Rate

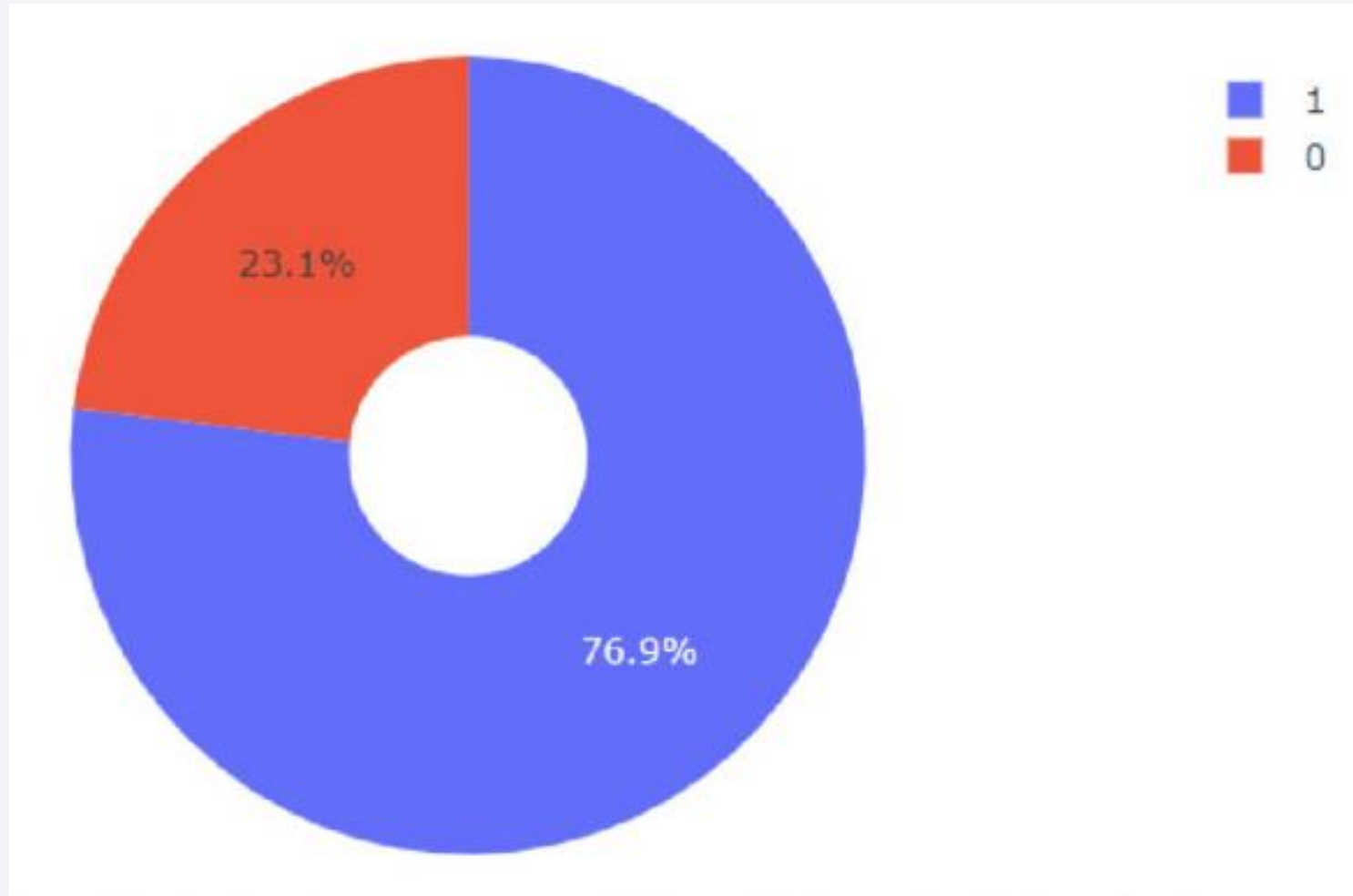
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- 41.70% of successful landings are achieved by KSC LC-39A
- Followed by CCAFS LC-40 at 29.2%, VAFB SLC-4E at 16.7%, and CCAFS SLC-40 at the last

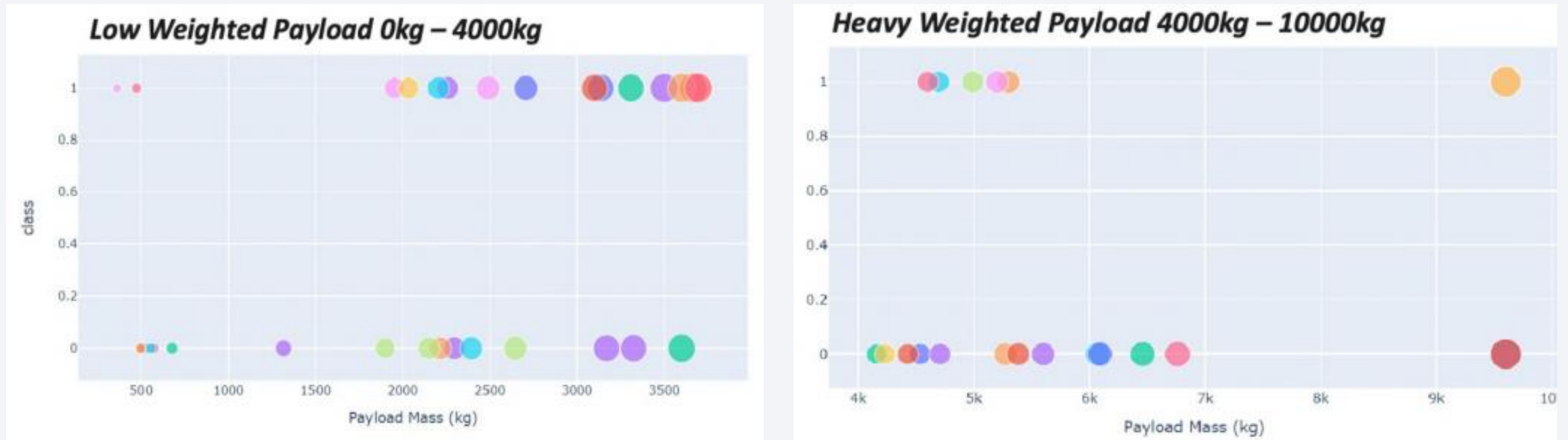
# KSC LC-39A Launch Ratio

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- Launch success ratio of KSC LC-39A
- It achieved 76.90% success

# PayloadMass vs Outcome Plot



- Low payload has a higher success rate





Section 5

# Predictive Analysis (Classification)

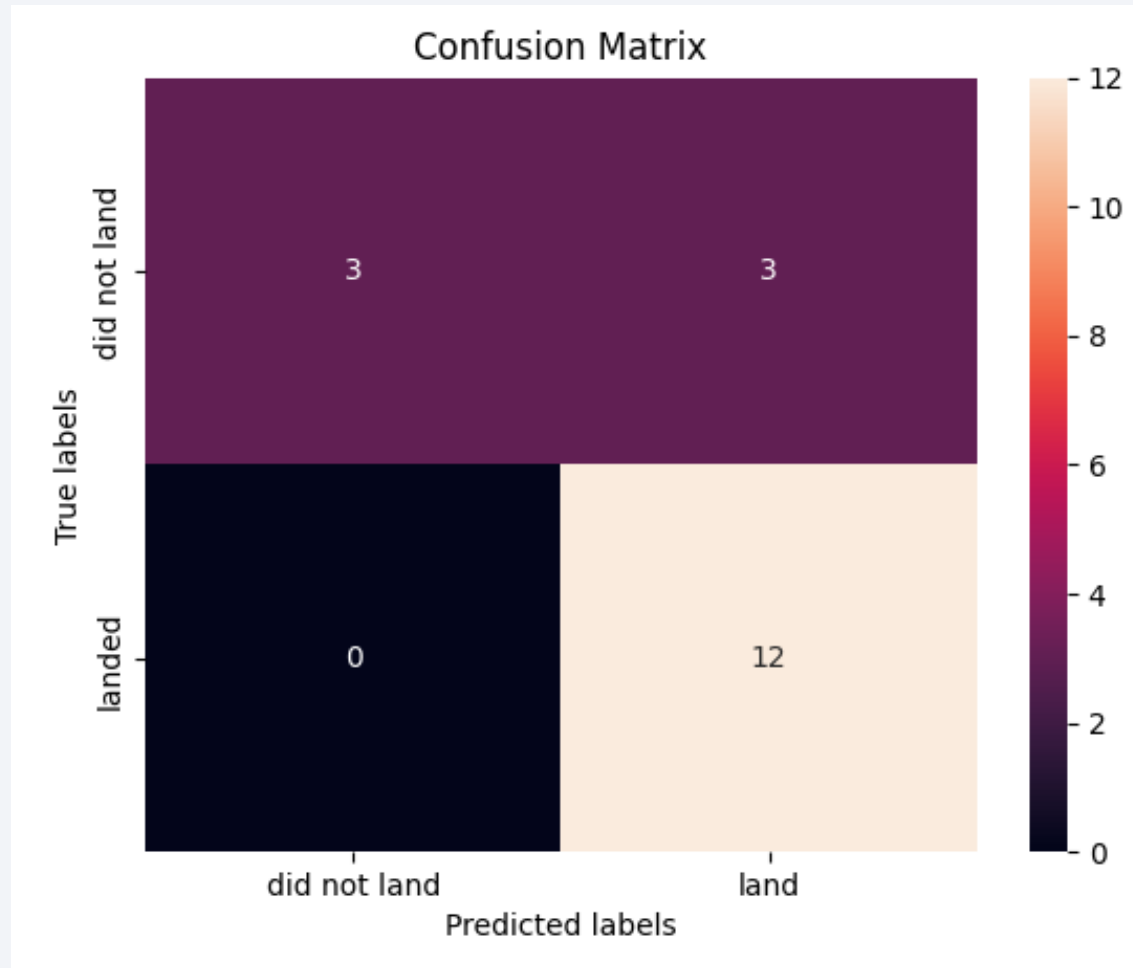


# Classification Accuracy

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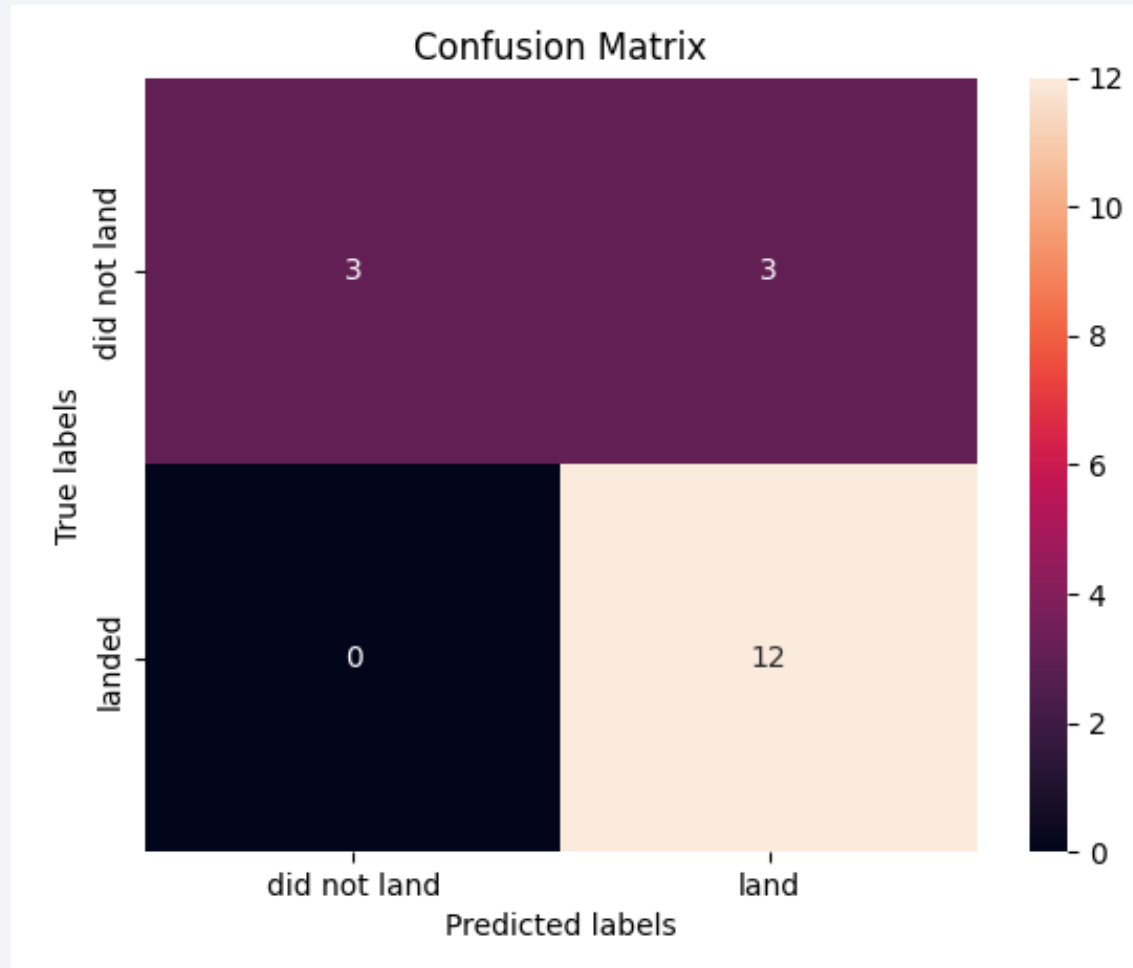
- Visualize the built model accuracy for all built classification models, in a bar chart
- Find which model has the highest classification accuracy

# Confusion Matrix – Logistic Regression



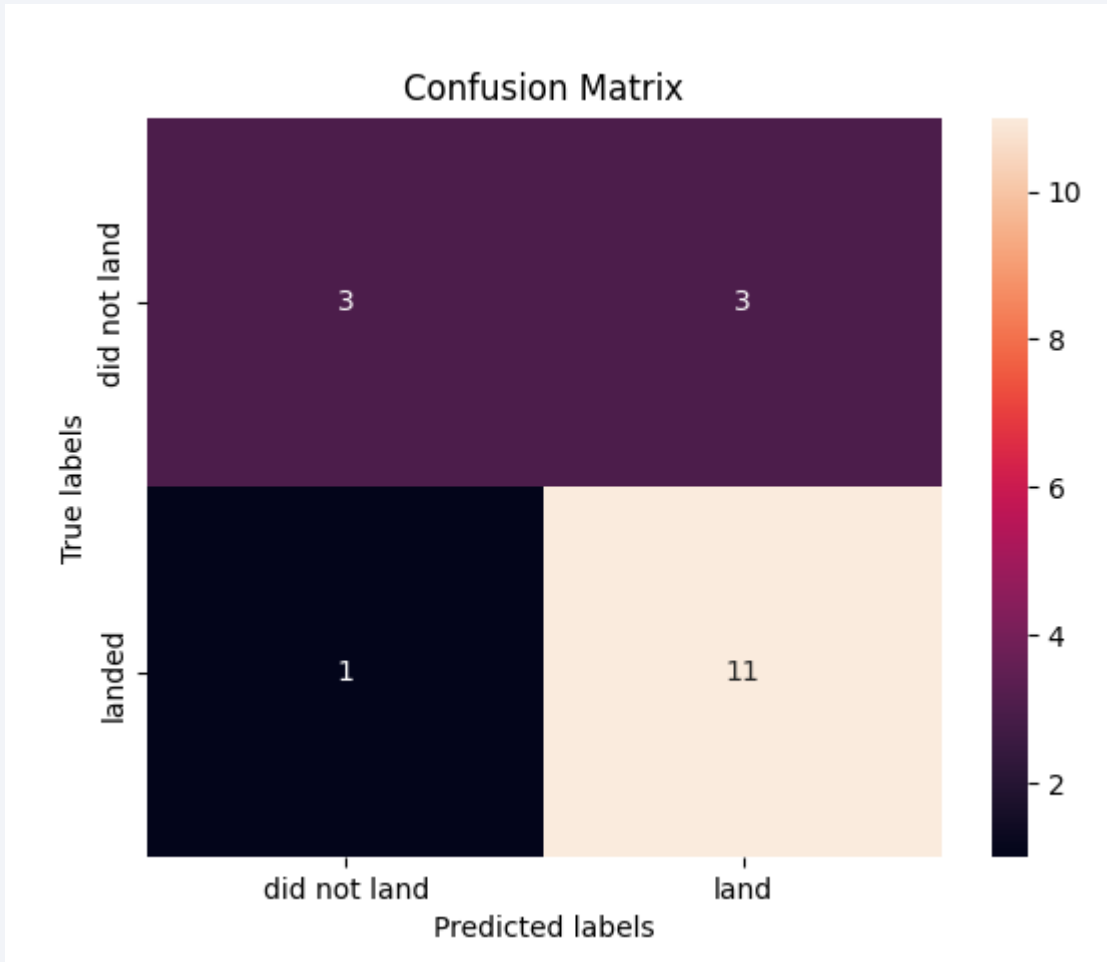
- Show the confusion matrix of the best performing model with an explanation

# Confusion Matrix – SVM



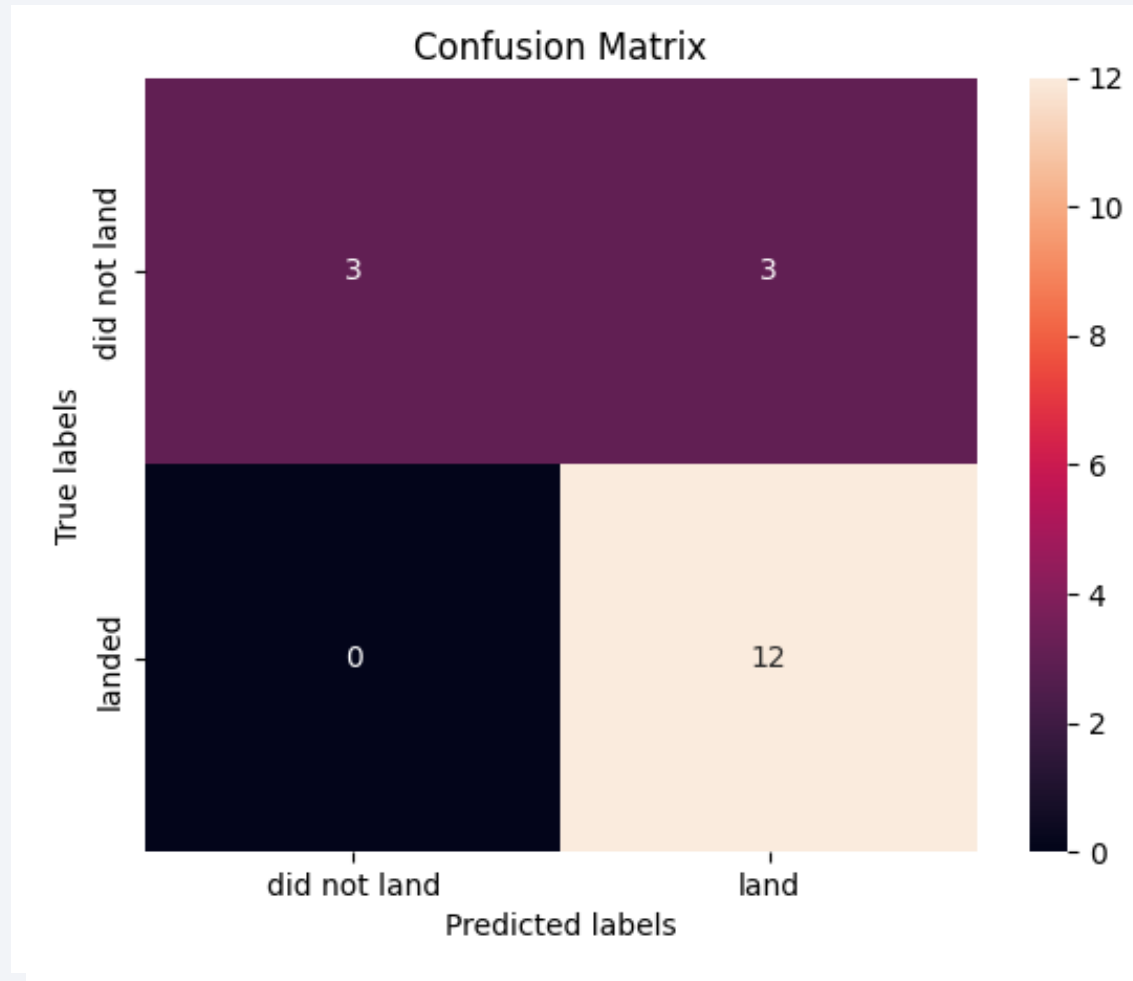
- Show the confusion matrix of the best performing model with an explanation

# Confusion Matrix – Decision Tree



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

# Confusion Matrix – KNN



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

# Conclusions

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- KSC LC39-A is the best performing site
- Light payloads performed better
- Since 2013 the launches have improved in terms of success
- Decision Tree was the worst performing algorithm

# 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

Thank you!

