



IBM Developer  
SKILLS NETWORK

# Winning Space Race with Data Science

Murat ÇELEBI  
25.03.2023



# Outline

---

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

# Executive Summary

---

- Summary of methodologies
  - Data Collection,
  - Web scraping,
  - EDA,
  - Site Location
  - Prediction with Machine Learning
- Summary of all results
  - Accuracy Score: 0.83

# Introduction

---

- Project background and context
  - Project is an end-to-end data science project where we are investigating whether the first stage landing of the SpaceX Falcon 9 rocket will be successful.
- Problems you want to find answers
  - Launched rockets are a huge cost element. Therefore, companies have placed the cost item on the top of their success.
  - The first question we need to answer in this project will be:
  - Will the first stage of a launched rocket land successfully again?



Section 1

# Methodology

# Methodology

---

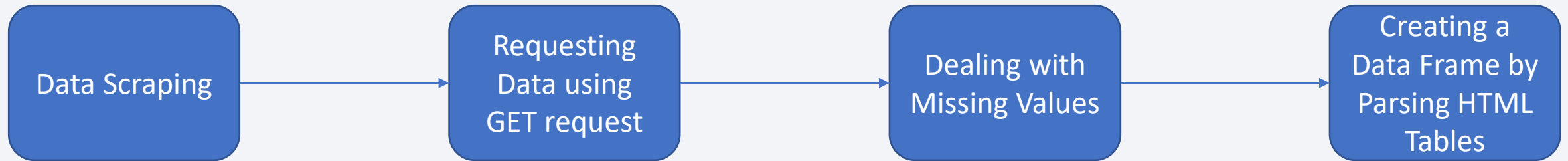
## Executive Summary

- Data collection methodology:
  - Describe how data was collected
- Perform data wrangling
  - Describe how data was processed
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - How to build, tune, evaluate classification models

# Data Collection

---

- Describe how data sets were collected.
  - I gathered the data by performing a fetch request from Wikipedia to the SpaceX API.
- You need to present your data collection process use key phrases and flowcharts



# Data Collection – SpaceX API

---

Github URL:

<https://l24.im/HnD>

<https://l24.im/3qO8L>

```
response = requests.get(spacex_url)
```

```
data = response.json()
data = pd.json_normalize(data)
```

```
data = pd.DataFrame({key: pd.Series(value) for key, value in launch_dict.items()})
```



# Data Collection - Scraping

---

- Requesting the Falcon 9 Launch Wiki page from its URL
- Extracting all column/variable names from HTML table header
- Creating a dataframe by parsing initialization HTML tables
- Github URL:
- <https://l24.im/2SPp>

Web Scraping Result Response object

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

Normalize

```
pd.json_normalize(requests.get(static_url))
```

A BeautifulSoup object from the HTML response

```
soup = BeautifulSoup(response)
```

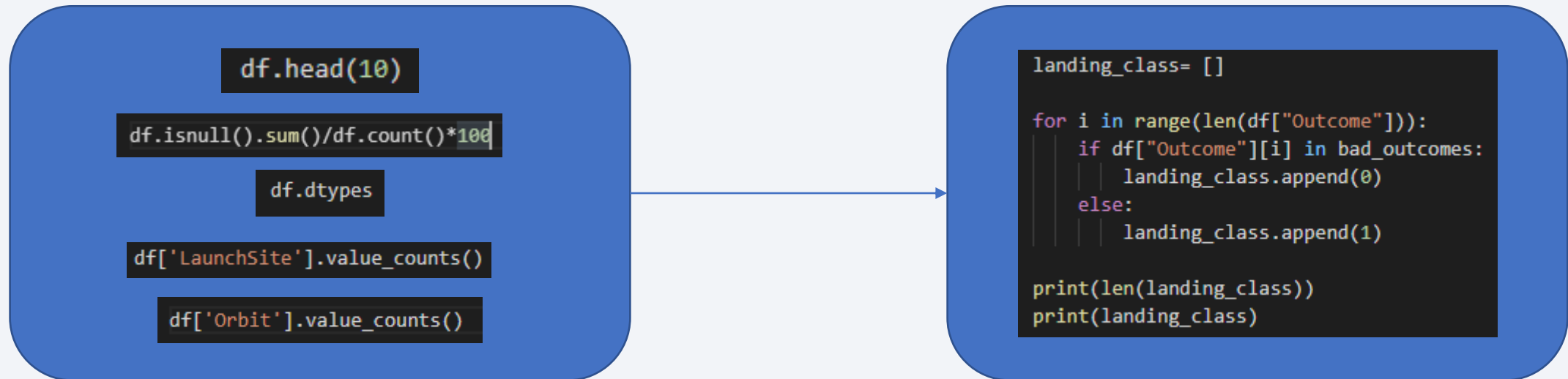
Extracting column names from HTML table header

```
html_tables = soup.find_all(name = 'table')
```

# Data Wrangling

---

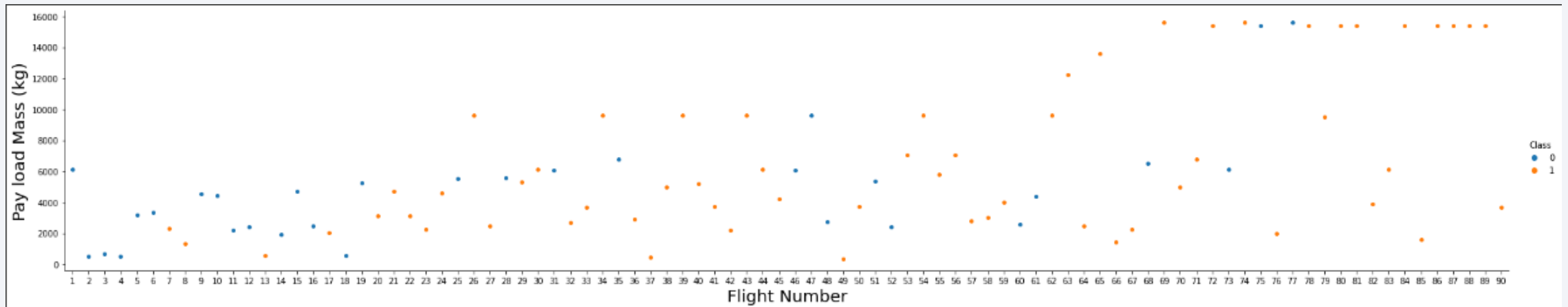
- We start by uploading our dataset. Then we perform a missing value analysis on the data set. Then we label the target variable as 0 and 1.



# EDA with Data Visualization

---

- He designed the catplot chart to visualize the weight of the weight in each flight unit and whether, depending on the series of flights, the rocket at that weight did not successfully land on its first step.



- Github URL: <https://l24.im/NnQPD6>

# EDA with SQL

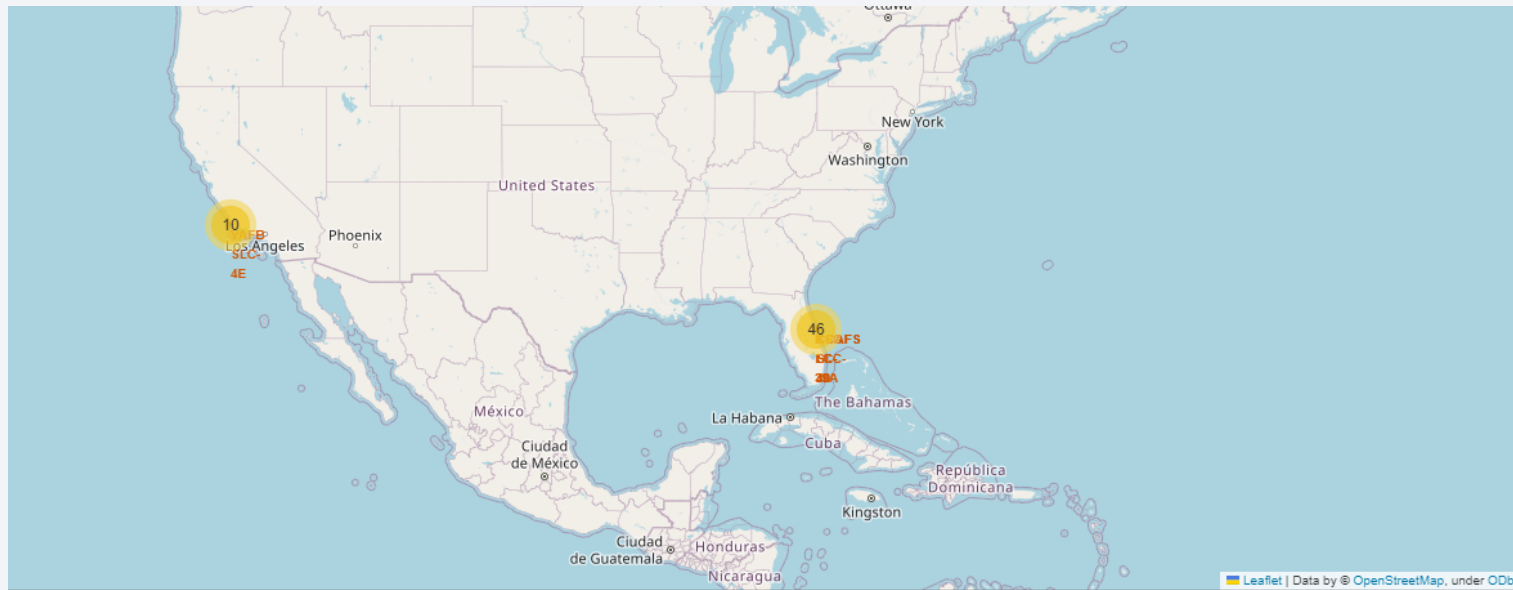
---

- The names of the unique launch sites in the space mission have been questioned.
- Questioning the total mass of payload carried by boosters launched by NASA (CRS).
- The date on which the first successful landing result was obtained on the ground pad was questioned.
- The total number of successful and unsuccessful mission results was questioned.
- Github URL: <https://l24.im/12j>

# Build an Interactive Map with Folium

---

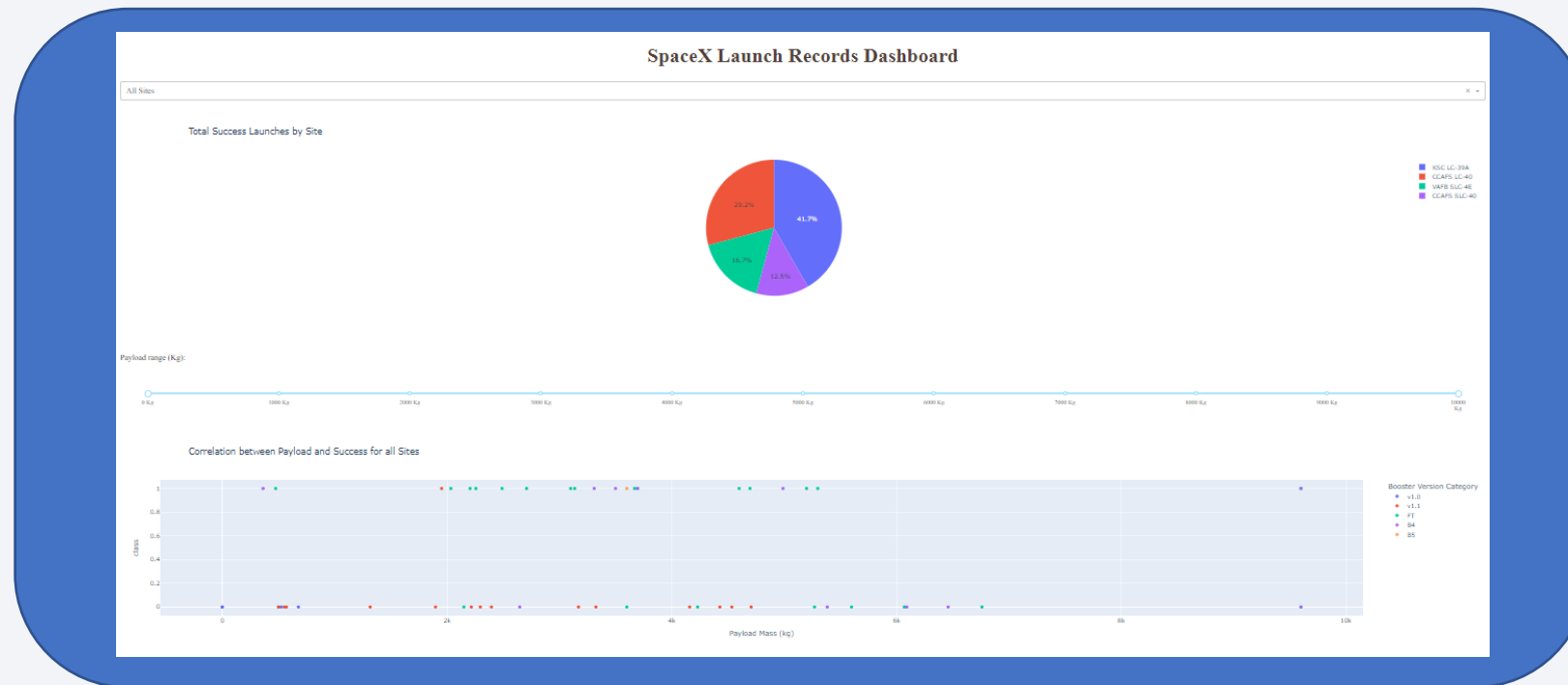
- I added markers to the leaf map, and all other objects to find launch pads on the world map and perform analysis.



- Github URL: <https://l24.im/AE0e>

# Build a Dashboard with Plotly Dash

- The distribution of graphs and drawings in our data and the drawings on the Dashboard have been retained for live display in summary.



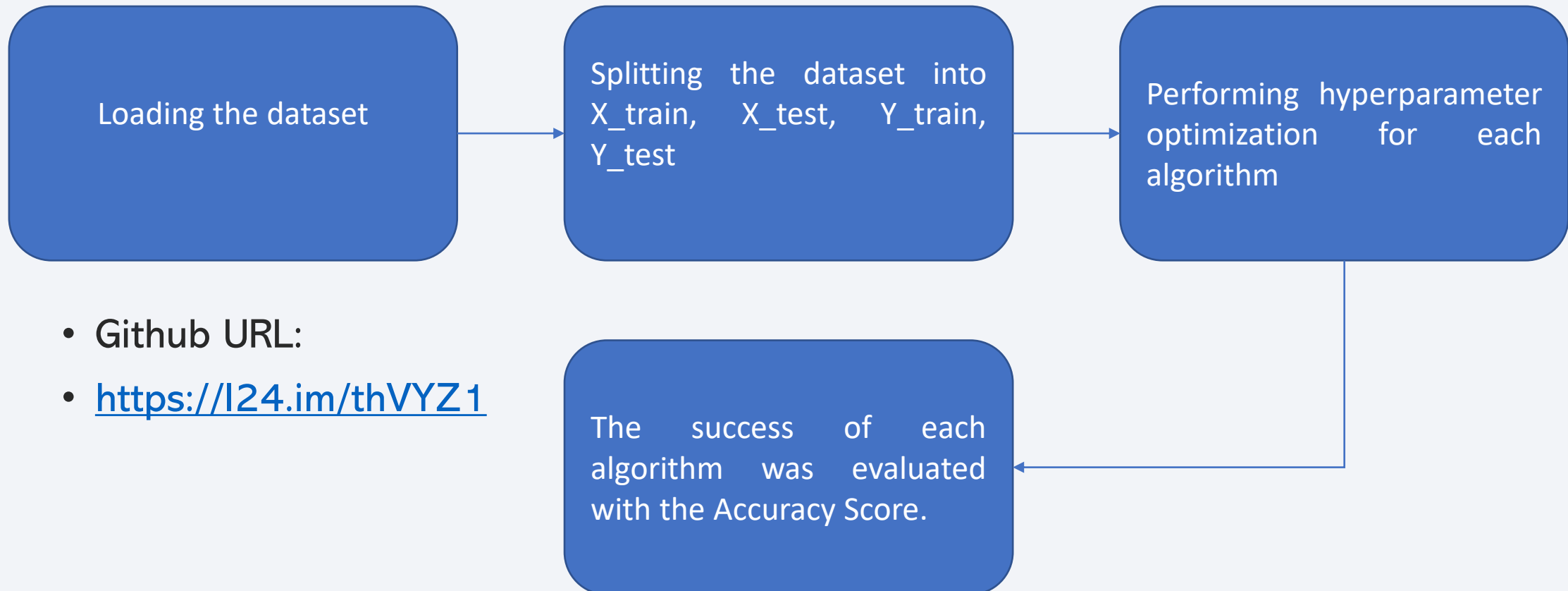
- Github URL: <https://l24.im/yad>



# Predictive Analysis (Classification)

---

- I used five or four classification algorithms in the project and evaluated the results with Accuracy Score.



- Github URL:
- <https://l24.im/thVYZ1>

# Results

---

- Exploratory data analysis results
- In Exploratory Data Analysis, it was observed that missing values, correlation between variables and successful first landings increased more over the years.
- Predictive analysis results

	Model	Accuracy Score (Hyperparameter Optimization)
0	Logistic Regression	0.833333
1	Support Vector Machine	0.833333
2	Decision Tree	0.888889
3	K Nearest Neighbor	0.833333



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 red and cyan. A faint, light blue grid pattern is also visible, particularly in the lower half of the image. The overall effect is dynamic and technological.

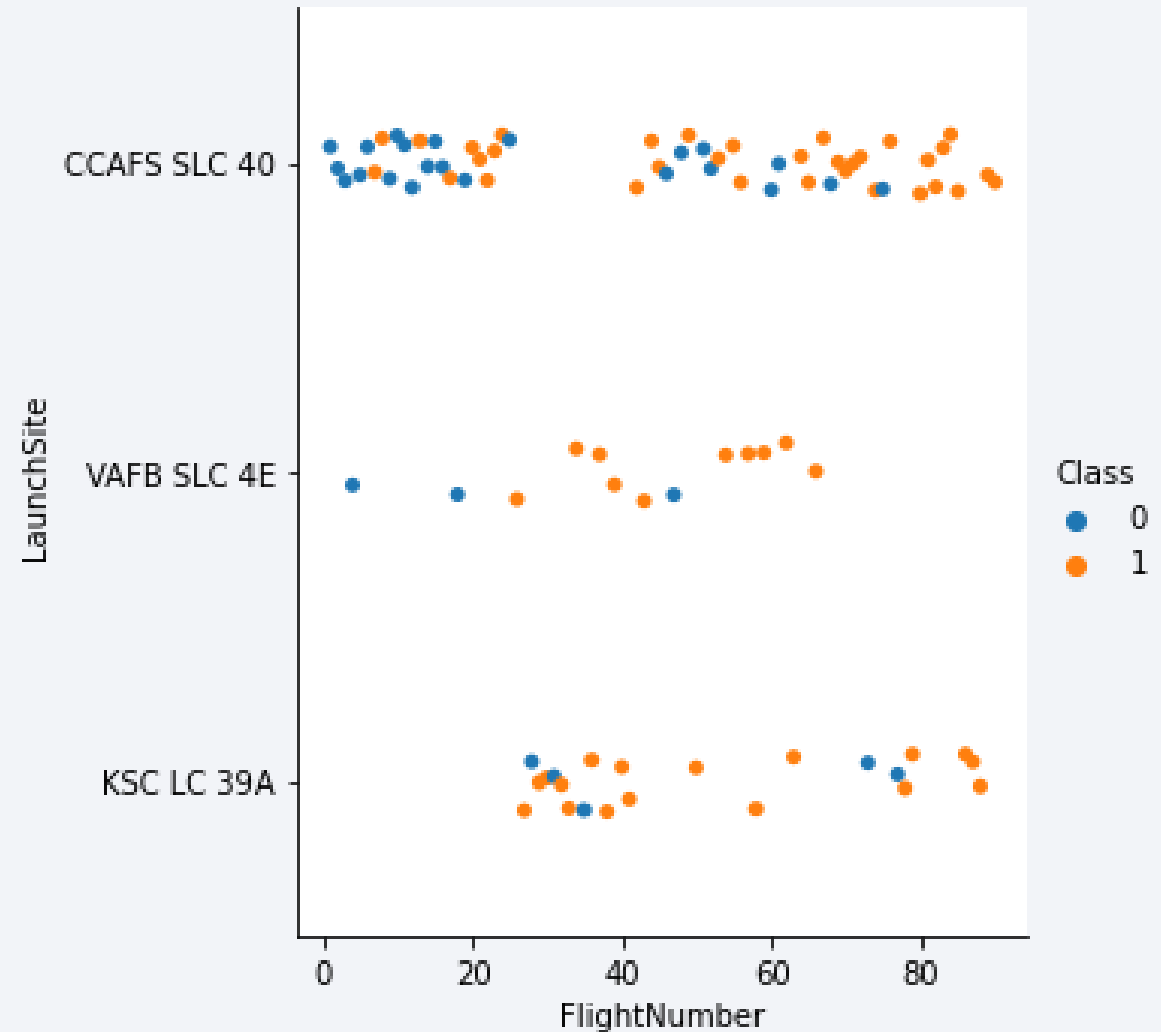
Section 2

# Insights drawn from EDA



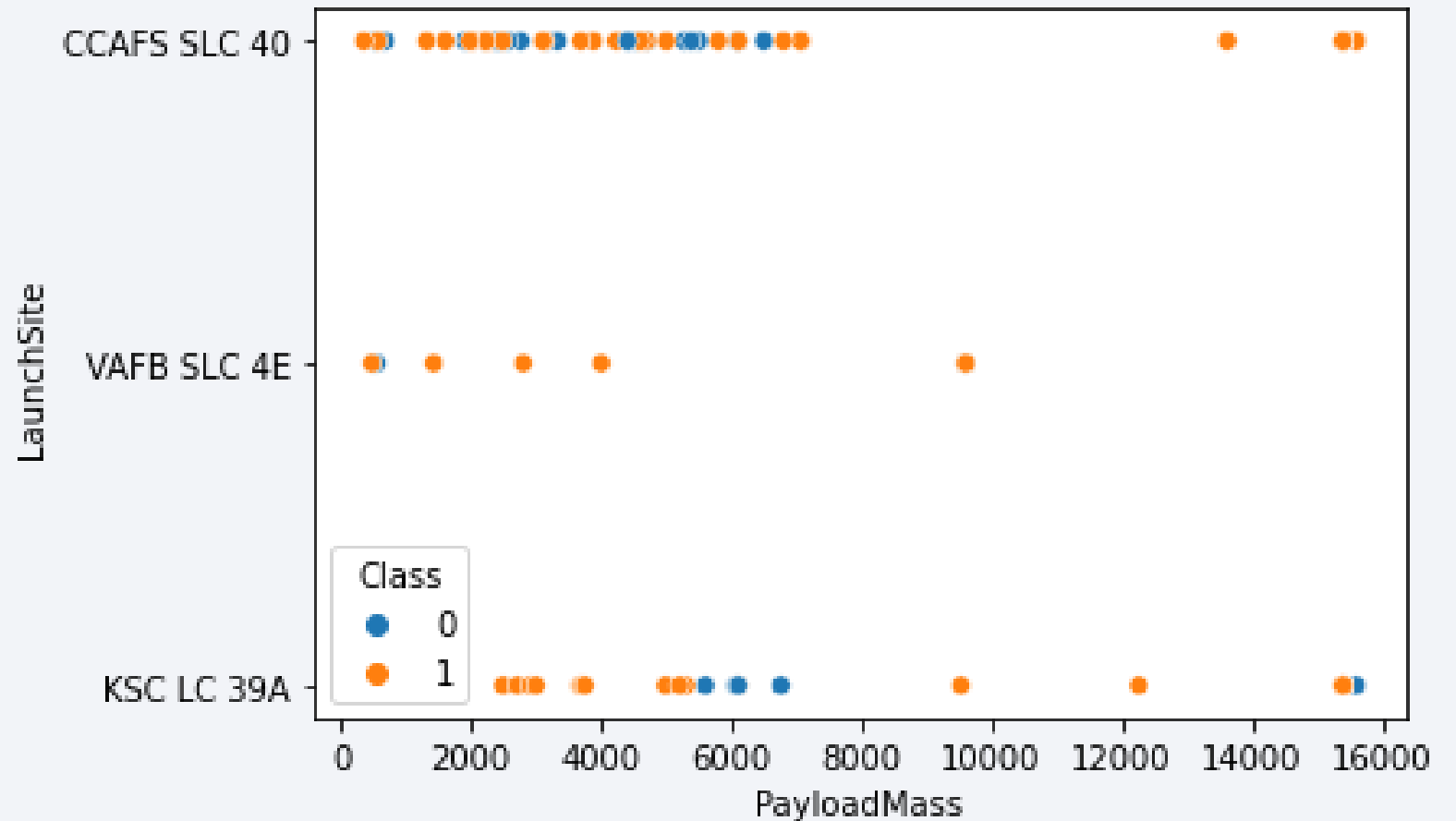
# Flight Number vs. Launch Site

- LaunchSite: CCAFS SLC 40 seems to have more failed landings.
- LaunchSite: KSC LC 39A has more successful landings.



# Payload vs. Launch Site

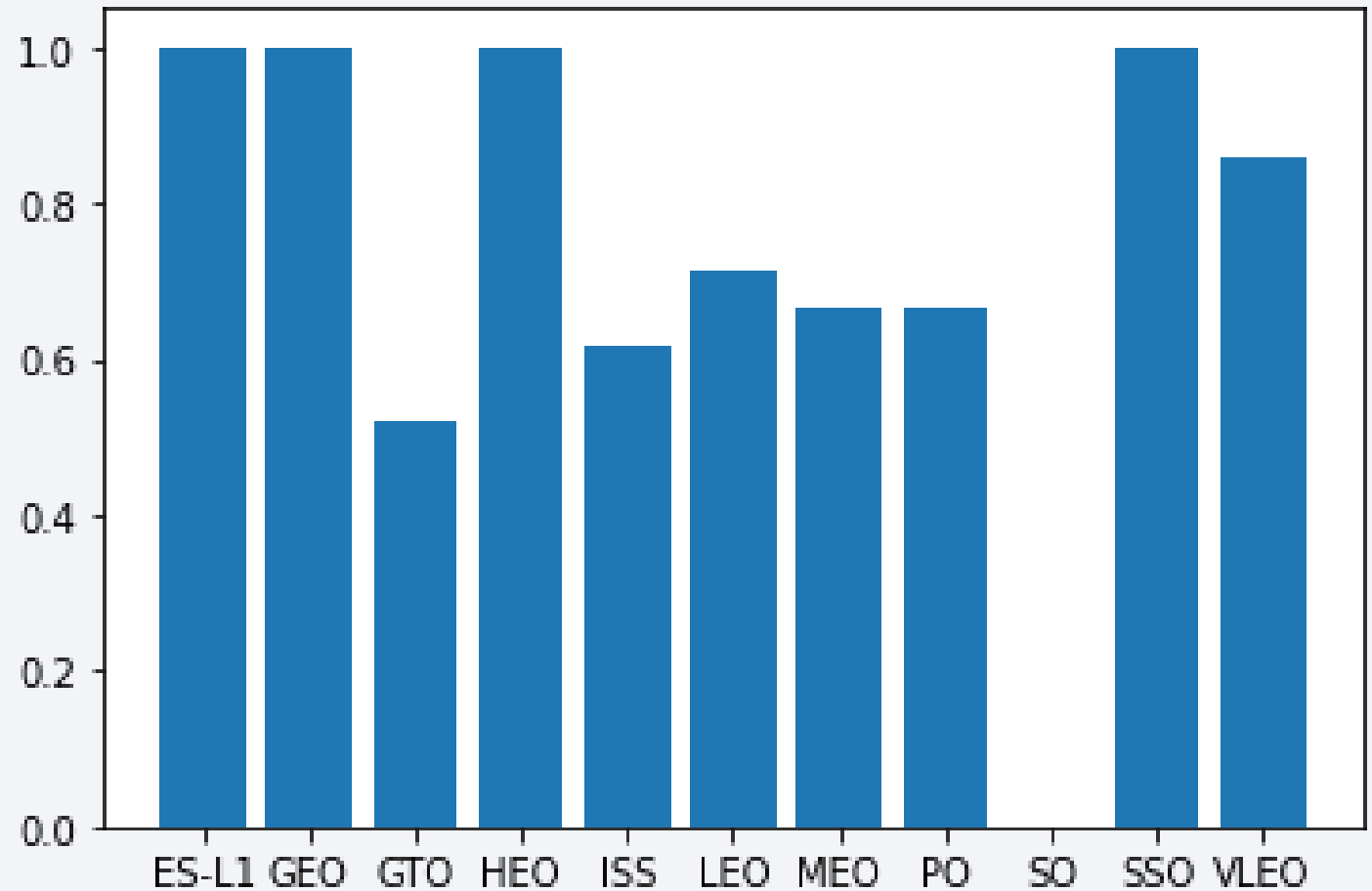
- Flights with payload masses between 2000 and 4000 appear mostly from LaunchSite: CCAFS SLC 40. And it is observed that most of the flights from here make successful landings.



# Success Rate vs. Orbit Type

---

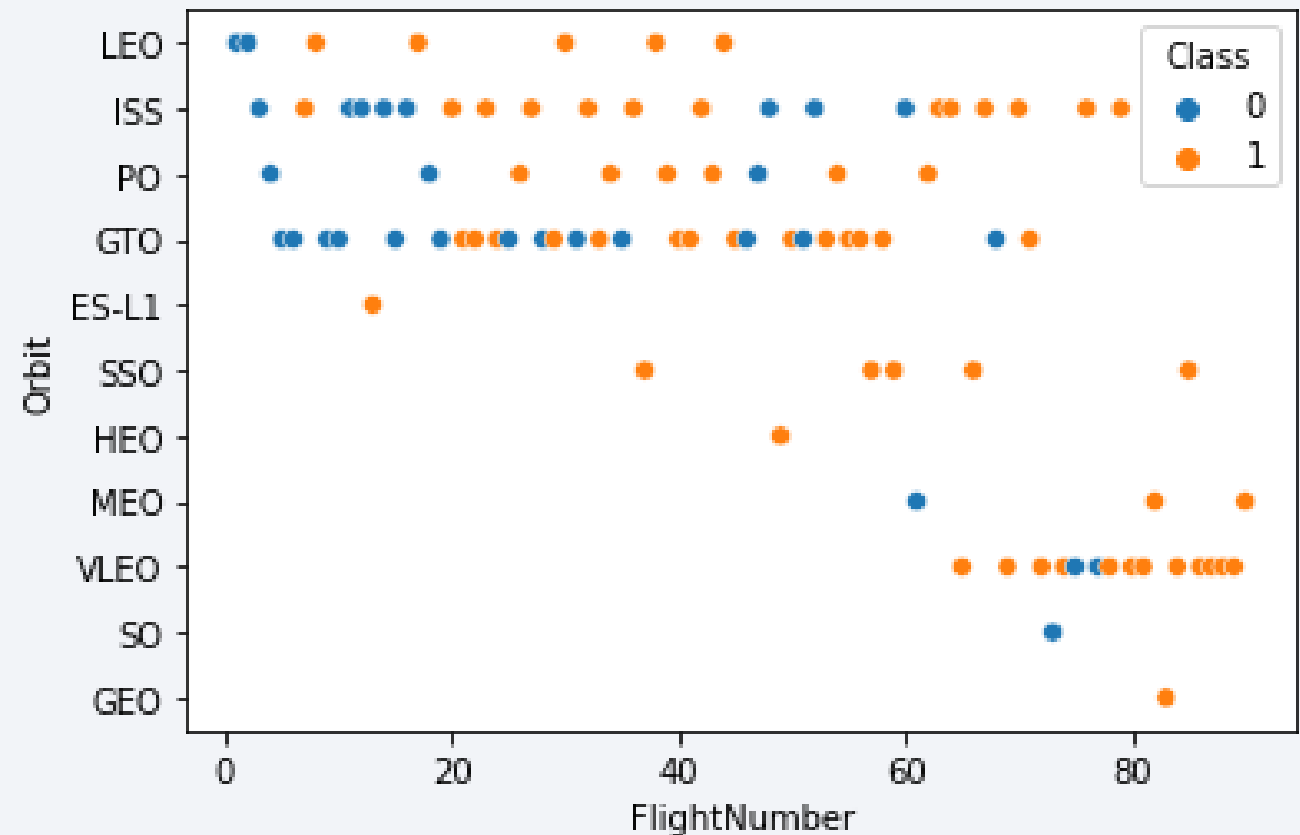
- Orbit Type: All flights with ES-L1, GEO, HEO and SSO appear to have a successful first landing.





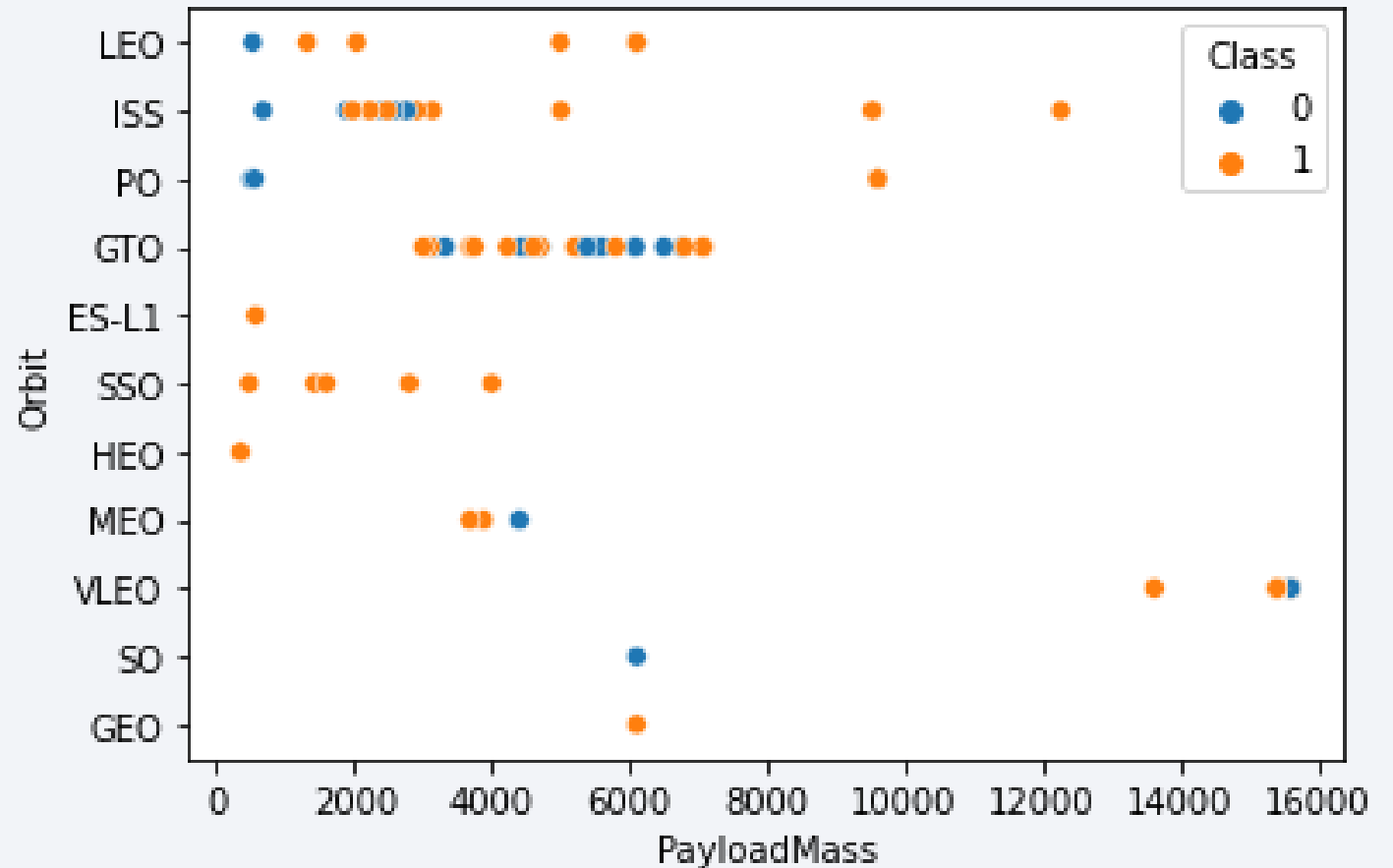
# Flight Number vs. Orbit Type

- It is seen that the number of successful landings increases as the flight numbers increase. At the same time, almost none of the first 50 flight number launches took place in orbits of the Orbit type (ES-L1, SSO, HEO, MEO, VLEO, SO, GEO).



# Payload vs. Orbit Type

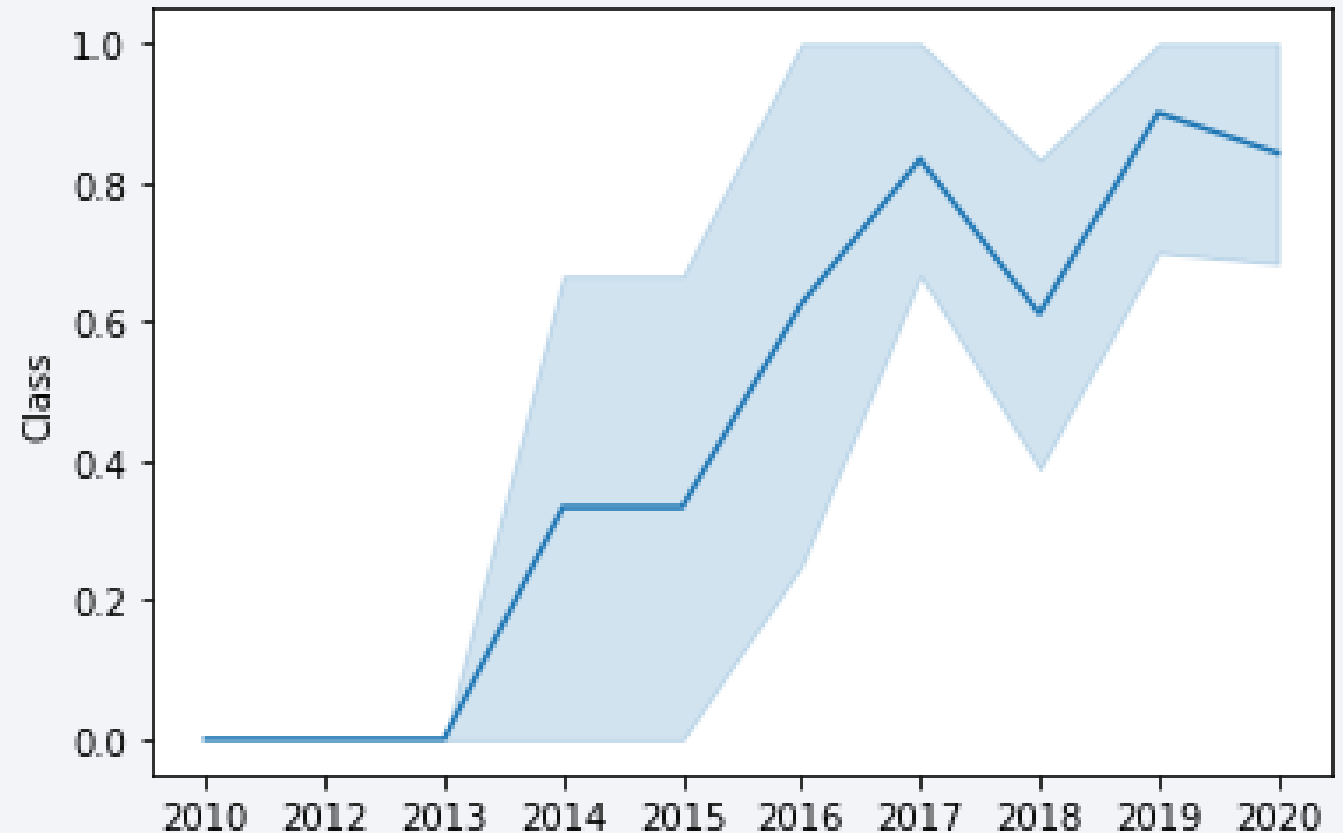
- It is seen that as the Payload Mass increases, the launches decrease.
- At the same time, all launches of the payload mass between 4000 and 8000 were made at the point of the Orbit-type GTO.



# Launch Success Yearly Trend

---

- It is very clear that as the years increase, the success rates increase.
- And none of the rockets launched in the first three years seem to have made their first successful landing.



# All Launch Site Names

---

- It can be seen that there are four initiation points.

```
1 %sql SELECT DISTINCT Launch_Site FROM SPACEXTBL
```

```
* sqlite:///my_data1.db  
Done.  
  
Launch_Site  
CCAFS LC-40  
VAFB SLC-4E  
KSC LC-39A  
CCAFS SLC-40
```

# Launch Site Names Begin with 'CCA'

- It seems that launch sites have only five starting points with CCA.

```
[ ] 1 %sql SELECT * FROM SPACEXTBL 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
04-06-2010	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
08-12-2010	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)
22-05-2012	07:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
08-10-2012	00:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
01-03-2013	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 total mass of payload carried by boosters launched by NASA (CRS) appears to be 45596 kg.

```
1 %sql SELECT SUM("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE "CUSTOMER" = 'NASA (CRS)'
```

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

```
Done.
```

```
SUM("PAYLOAD_MASS__KG_")
```

```
45596
```



# Average Payload Mass by F9 v1.1

---

- The average payload mass carried by the booster version F9 v1.1 appears to be 2534.6666666666665 kg.

```
[ ] 1 %sql SELECT AVG("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE "Booster_Version" LIKE '%F9 v1.1%'
```

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

```
Done.
```

```
AVG("PAYLOAD_MASS__KG_")
```

```
2534.6666666666665
```

# First Successful Ground Landing Date

---

- We see that the first successful landing was on the first day of May 2017.

```
[ ] 1 %sql SELECT MIN("DATE") FROM SPACEXTBL WHERE "Landing _Outcome" LIKE '%Success%'
* sqlite:///my_data1.db
Done.
MIN("DATE")
01-05-2017
```

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

---

- Rockets that landed successfully on your drone ship:
- F9 FT B1022,
- F9 FT B1026,
- F9 FT B1021.2,
- F9 FT B1031.2

```
1 %sql SELECT "BOOSTER_VERSION" FROM SPACEXTBL WHERE "LANDING _OUTCOME" = 'Success (drone ship)' \
2 AND "PAYLOAD_MASS_KG_" > 4000 \
3 AND "PAYLOAD_MASS_KG_" < 6000;
```

```
* 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

---

- We see that the total successful landing is 100 and the total failed landing is 1.

```
1 %sql SELECT \
2 (SELECT COUNT("MISSION_OUTCOME") FROM SPACEXTBL WHERE "MISSION_OUTCOME" LIKE '%Success%') AS SUCCESS, \
3 (SELECT COUNT("MISSION_OUTCOME") FROM SPACEXTBL WHERE "MISSION_OUTCOME" LIKE '%Failure%') AS FAILURE
```

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

SUCCESS	FAILURE
100	1

# Boosters Carried Maximum Payload

---

- Below we see the names of the launchers carrying the maximum load mass.

```
[ ] 1 %sql SELECT DISTINCT "BOOSTER_VERSION" FROM SPACEXTBL \
    2 WHERE "PAYLOAD_MASS__KG_" = (SELECT max("PAYLOAD_MASS__KG_") FROM SPACEXTBL)
```

```
* 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

---

- In 2015, we see the results of the rockets that landed unsuccessfully.



```
1 %sql SELECT substr("DATE", 4, 2) AS MONTH, "BOOSTER_VERSION", "LAUNCH_SITE" \
2 FROM SPACEXTBL\
3 WHERE "LANDING _OUTCOME" = 'Failure (drone ship)' and substr("DATE",7,4) = '2015'
```

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

```
Done.
```

	MONTH	Booster_Version	Launch_Site
01		F9 v1.1 B1012	CCAFS LC-40
04		F9 v1.1 B1015	CCAFS LC-40



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

---

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.

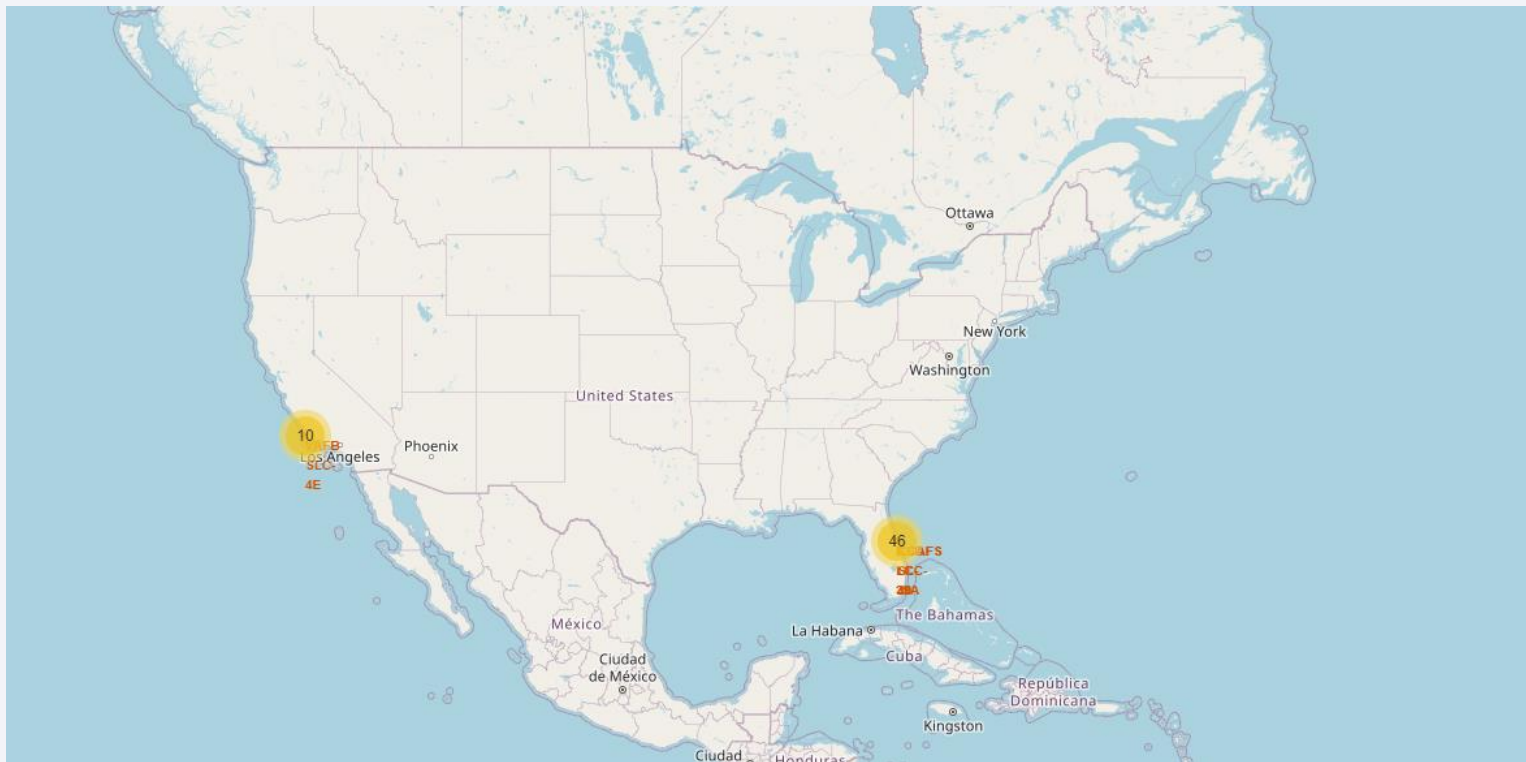
Section 3

# Launch Sites Proximities Analysis

# Site Map

---

- The map is as follows when a marker is added at each launch point in the dataset.



# Color Labels

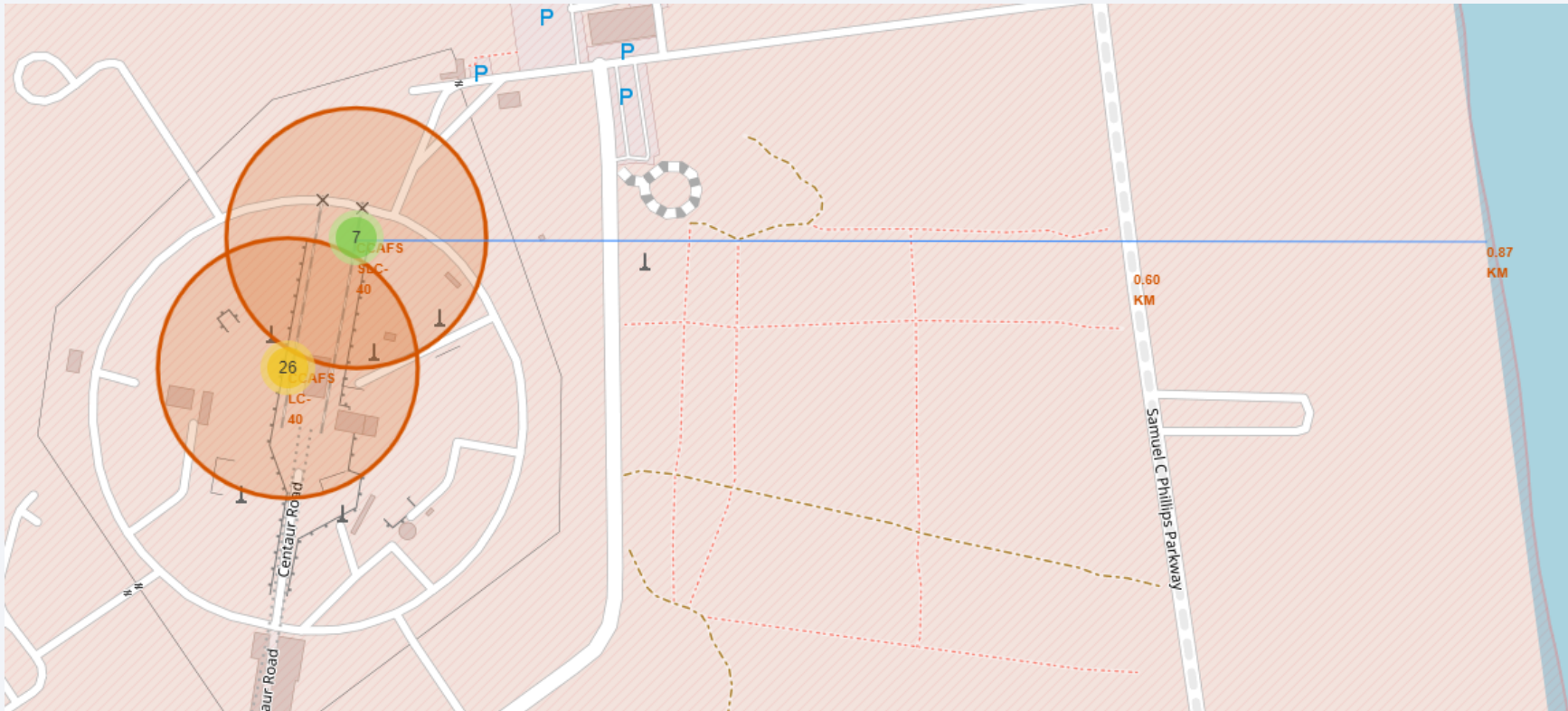
---

- We see that it is labeled with colored labels.



# Adding Distances to the Map

- We see the distance of the launch point to the nearest shore.





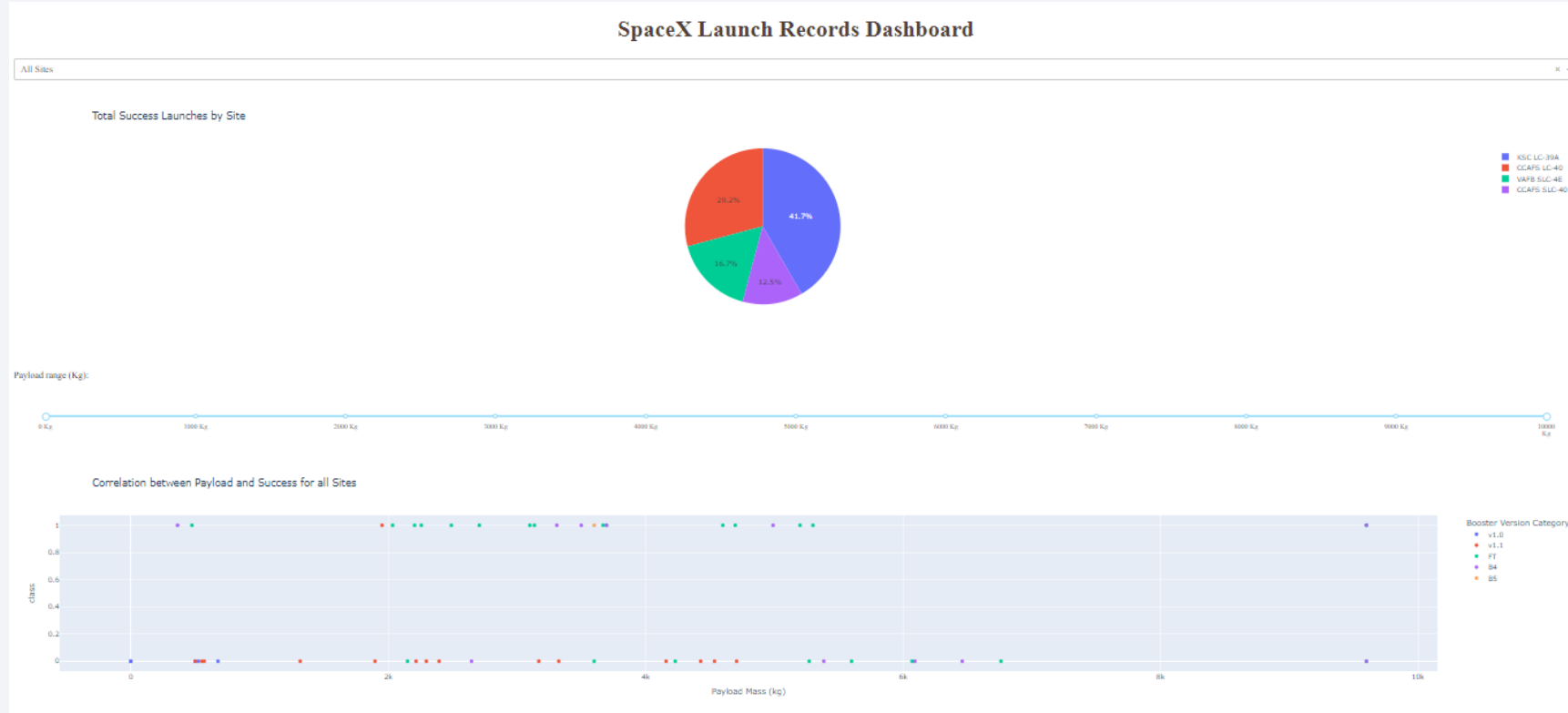
The background of the slide is a close-up, artistic photograph of a printed circuit board (PCB). The board is dark, and the intricate circuit traces are highlighted in a vibrant, glowing red. Numerous small, circular components, likely solder joints or micro-components, are visible along the traces, some of which also appear to be glowing. The overall effect is a high-tech, digital aesthetic.

Section 4

# Build a Dashboard with Plotly Dash

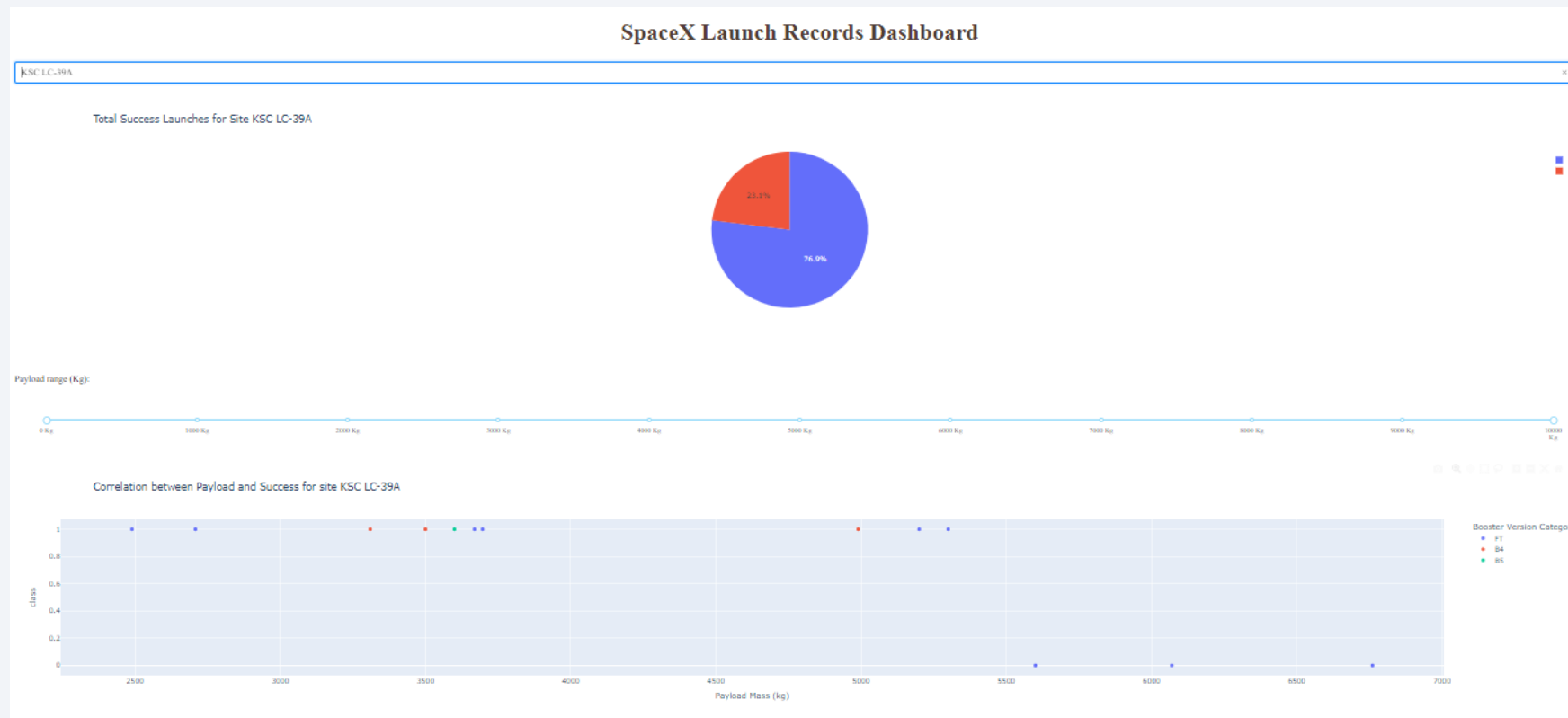
# Tüm Siteler Panosu

- fdg



# Highest Launch Success Rate

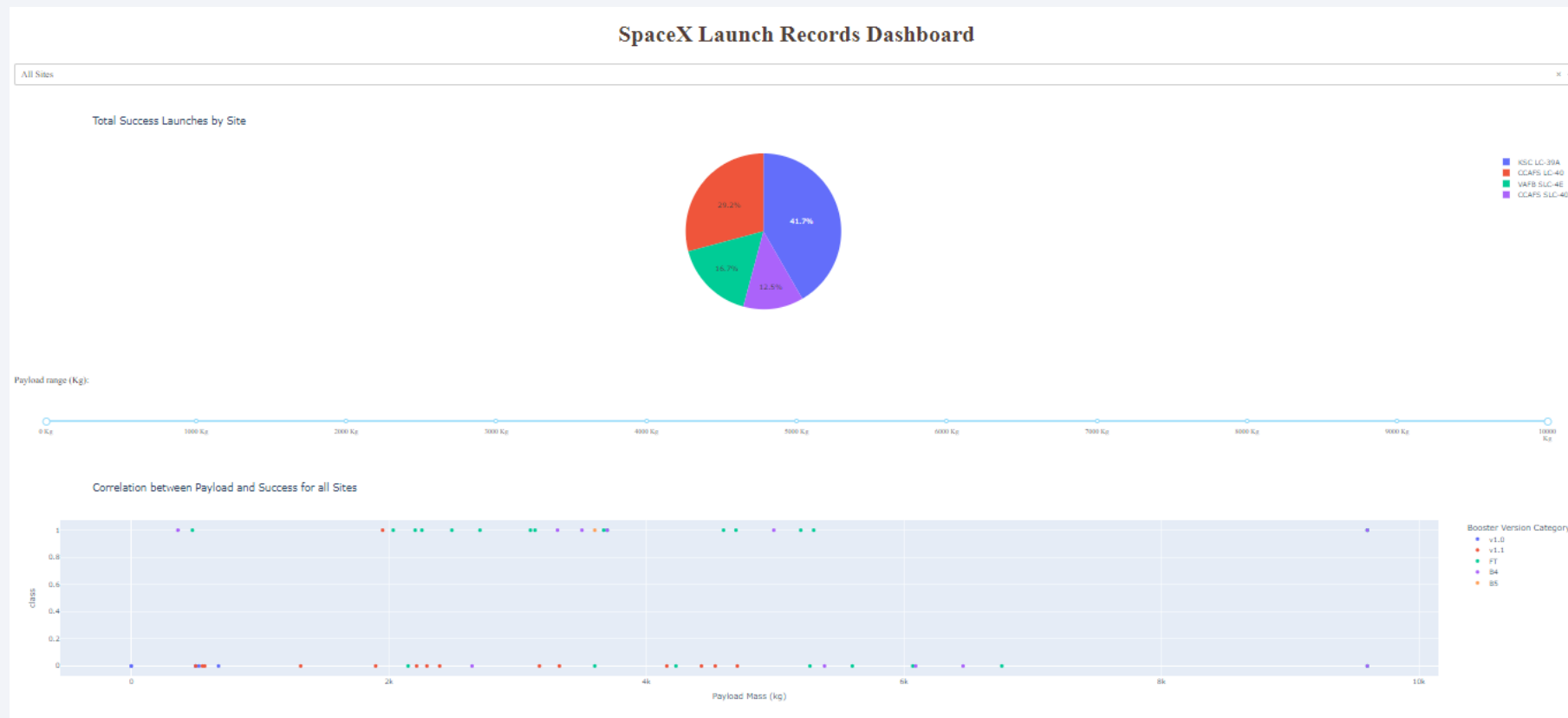
- In the Dashboard below, we see the point with the highest launch point.





# All Cities Distribution Chart

- We see the scatterplot graph at different values.



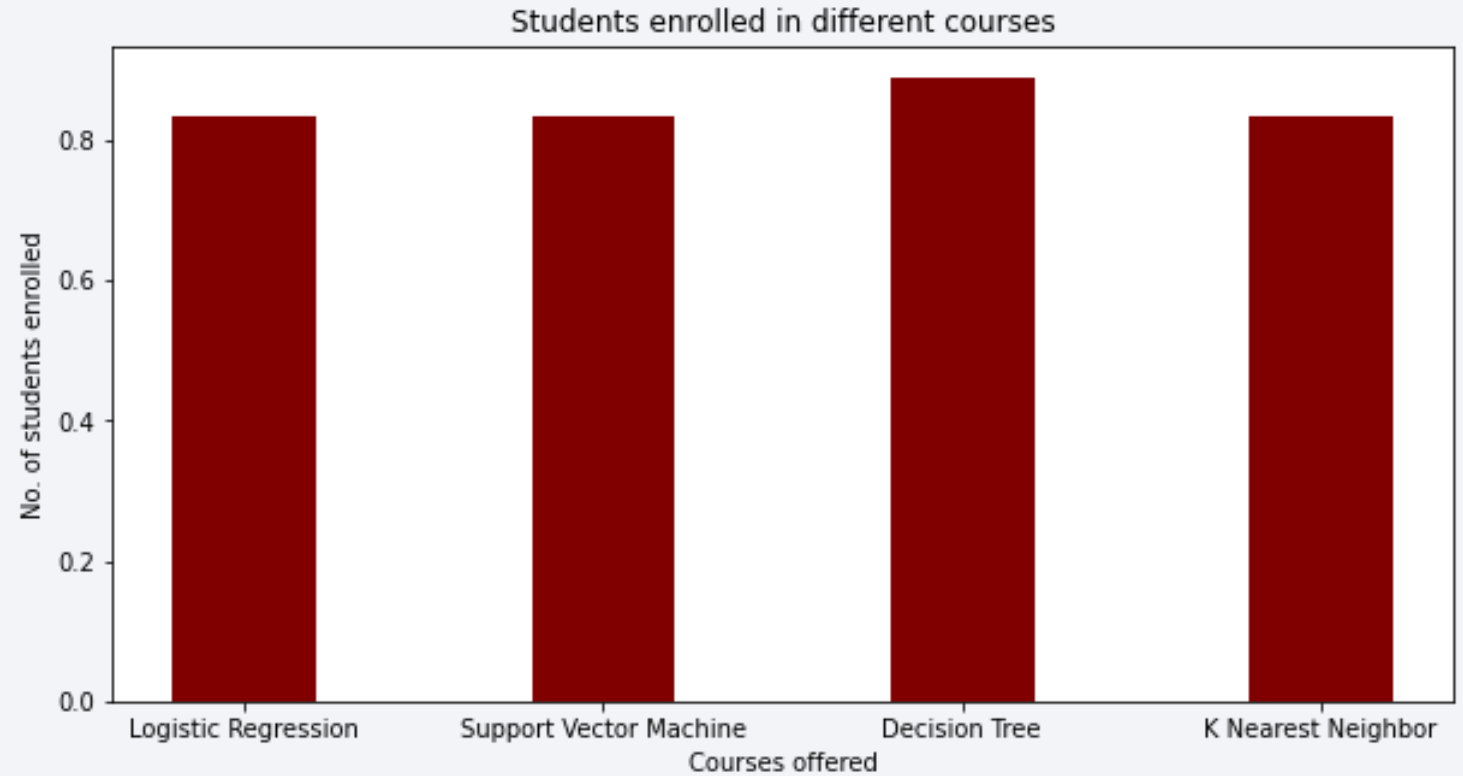
Section 5

# Predictive Analysis (Classification)

# Classification Accuracy

---

- Best Model: Decision Tree



# Confusion Matrix

- Best Model: Decision Tree
- Decision Tree Confusion Matrix



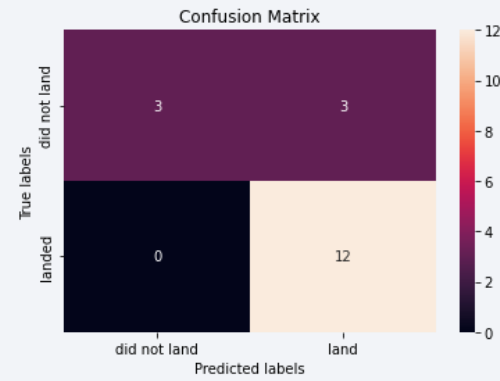
# Conclusions

---

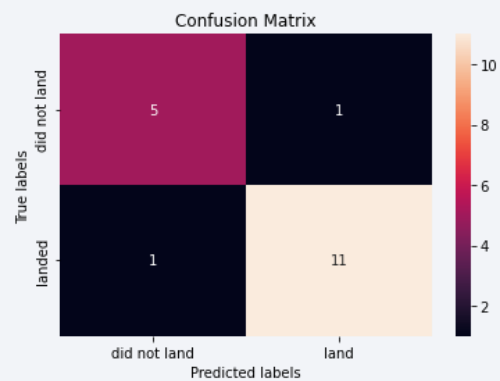
## Logistic Regression



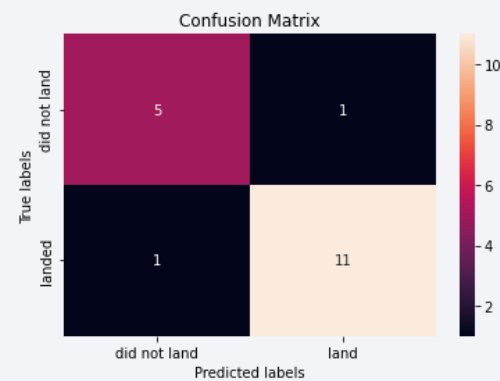
## Support Vector Machine



## Decision Tree



## KNN



# Appendix

---

```
model_name = ["Logistic Regression",
              "Support Vector Machine",
              "Decision Tree",
              "K Nearest Neighbor"]

ho_as = [logreg_ho, svm_ho, tree_ho, knn_ho]

model = pd.DataFrame({"Model" : model_name,
                     "Accuracy Score (Hyperparameter Optimization)": ho_as})
model
```

```
fig = plt.figure(figsize = (10, 5))

columns = model["Model"]
values = model["Accuracy Score (Hyperparameter Optimization)"]
# creating the bar plot
✓plt.bar(columns, values, color = 'maroon',
         width = 0.4)

plt.show()
```

Thank you!

