DEPI

SPACE_Y PROJECT

Our Team

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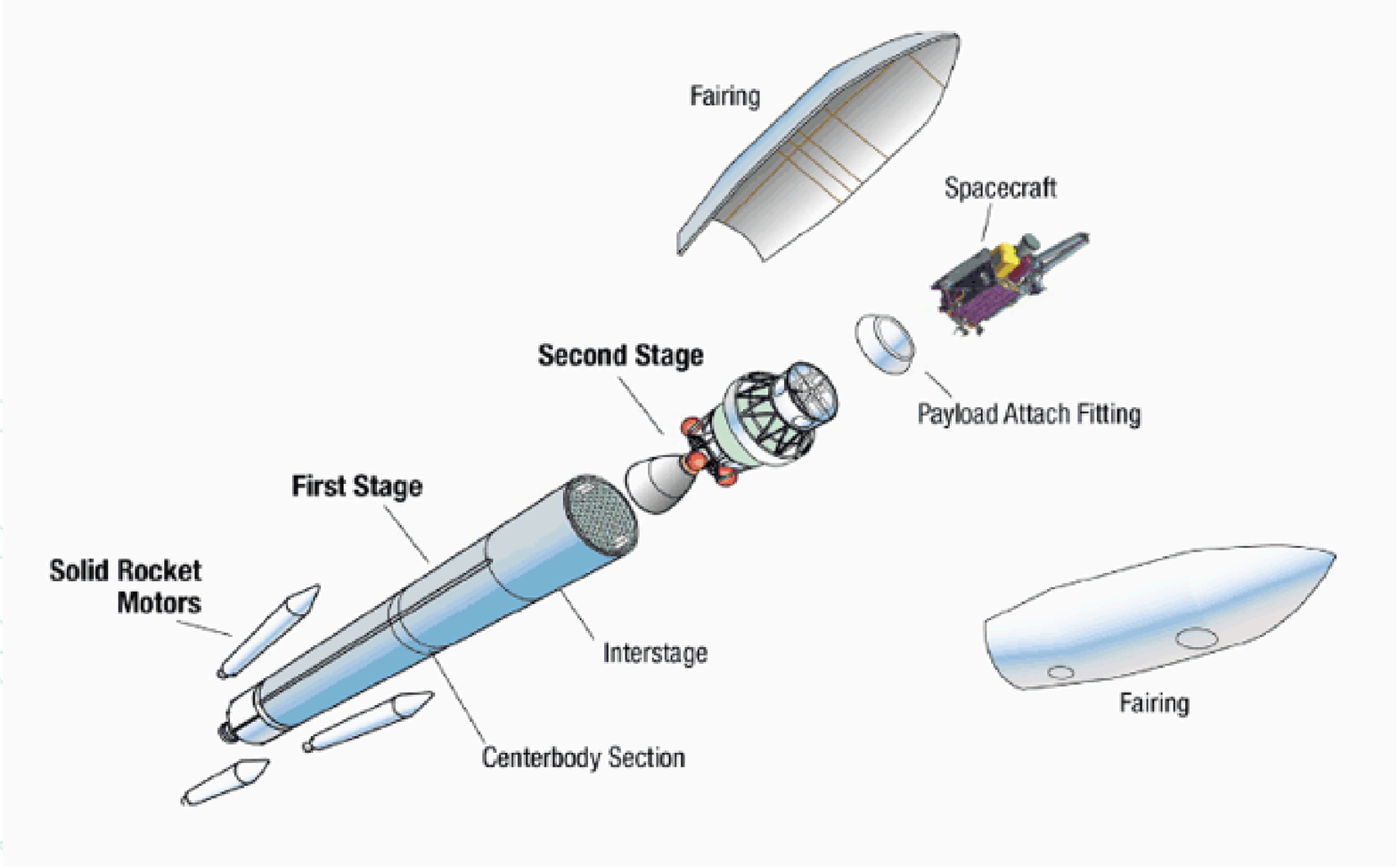
Scenario


Fairing
Spacecraft
Second Stage
Payload Attach Fitting
First Stage
Solid Rocket Motors
Interstage
Centerbody Section
Fairing

My Role in Data Wrangling

- My primary responsibility in the Space Y project was focused on Data Wrangling.
- Data Wrangling involves cleaning, organizing, and structuring raw data to make it usable for analysis and insights.
- I combined and prepared datasets from various SpaceX sources to ensure all the necessary data was available in a clear and structured format.

Why Data Wrangling is Critical

- Data Wrangling is a fundamental process because raw data often comes in complex and unstructured formats.
- By cleaning and organizing the data, we ensure that the project has a solid foundation for accurate analysis.
- I ensured that relevant data like rocket types, payload mass, and launch outcomes were easily accessible in one dataset.

Main and Supporting Datasets

- The core of my work involved processing a main dataset with past SpaceX launches.
  - Supporting datasets included:
    - Rockets dataset for booster details
    - Payload dataset for mass and orbit details
    - Launchpads dataset for site locations

Merging Data for Better Insights

- I used merge operations to integrate supporting datasets into the main launches dataset.
- New columns like BoosterVersion, LaunchSite, and PayloadMass were added to enrich the data.
- This process made it easier to analyze key factors affecting each launch.

Working with Nested Data

- Some datasets, like payloads and cores, had nested data (lists inside each launch).
- I used data explosion techniques to convert these lists into rows, making each payload and core analyzable individually.
- This allowed us to compute metrics like total payload mass for each launch.

Summary of My Contributions

Key Contributions:

- Cleaned and organized data from multiple SpaceX datasets
- Merged essential data points like booster versions, payload mass, and launch sites
- Handled complex data structures to make the data ready for analysis

Key Processing Steps Overview

Merging launch site information

This involves combining coordinates and names of launch sites for accurate data representation.

Processing payload data

In this step, mass and orbit information of the payload is processed to ensure compatibility with the mission.

Adding core information

Core details including block and serial numbers are integrated into the processing steps.

Creating landing outcome information

This step focuses on detailing the results of landing operations for future reference and analysis.

Converting landing pad codes

Landing pad codes are translated into actual names to facilitate better understanding and communication.

Filtering only Falcon 9 launches

This step ensures the analysis is concentrated solely on Falcon 9 data, eliminating irrelevant information.

Fixing date formats

Correcting date formats improves consistency and readability of the launch data.

Reordering and renaming columns for clarity

This process enhances the organization of the data, making it easier to interpret and analyze.

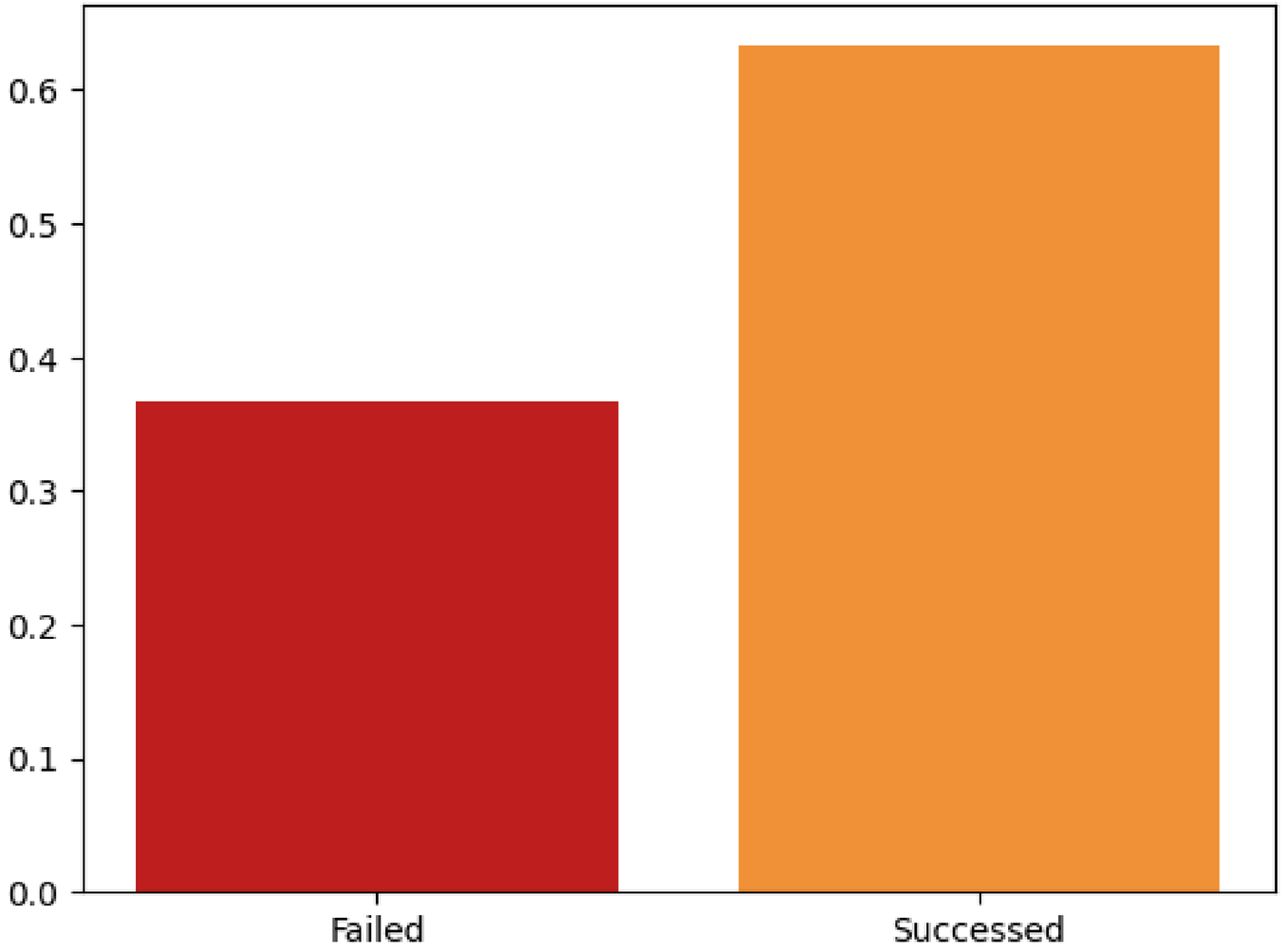
Handling missing payload mass data

Addressing gaps in payload mass data is crucial for accurate analysis of launch metrics.

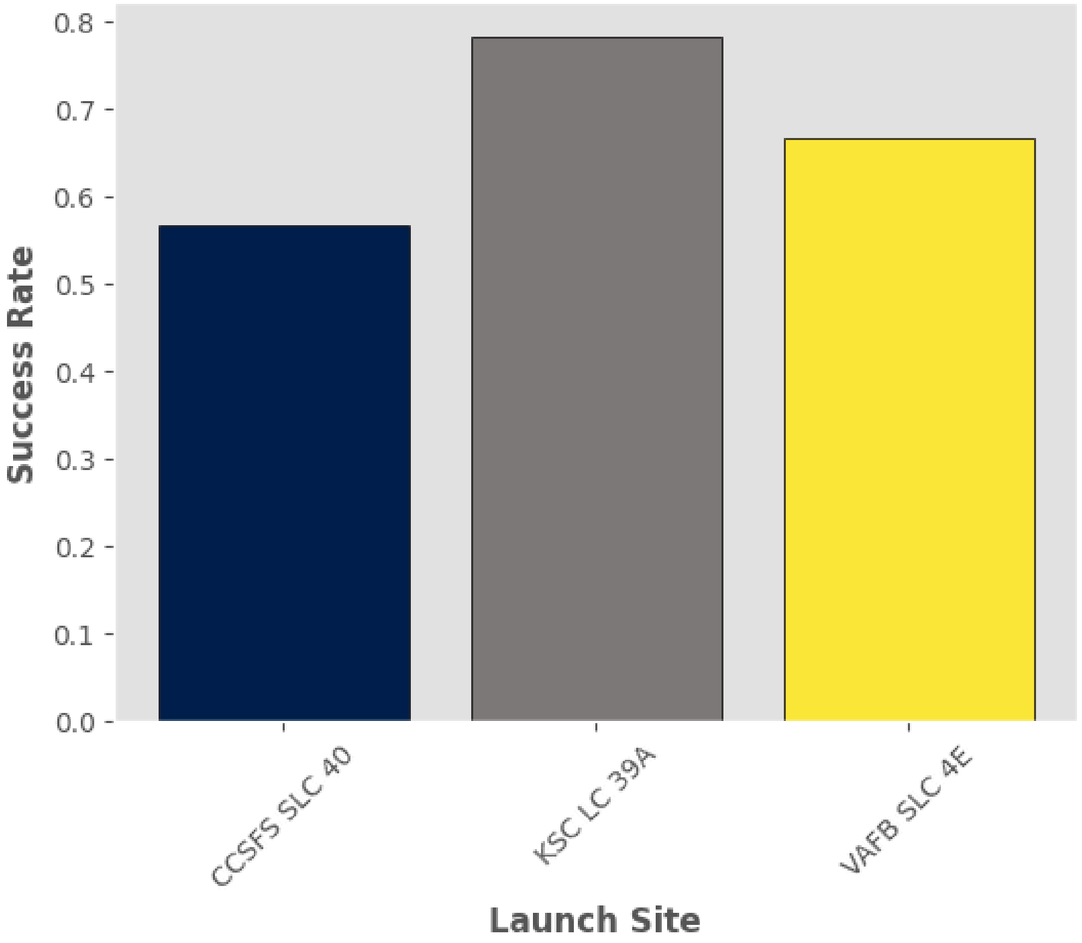
Final Dataset Overview

Comprehensive Information about Falcon 9 Launches
The end result is a clean dataset with comprehensive information about each Falcon 9 launch.

what is the percentage of the successful landings

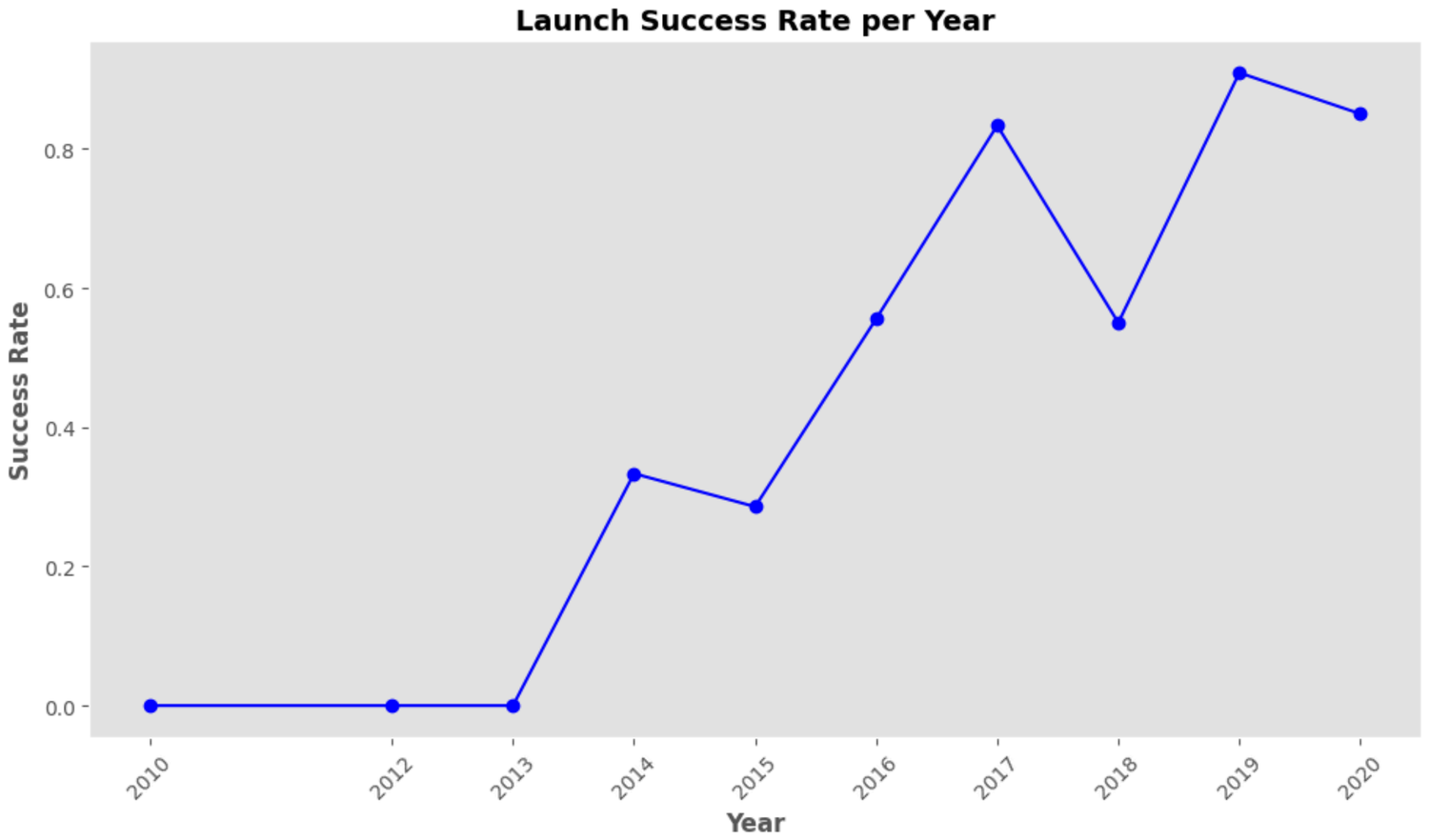


What is the success rate of each launching site ?



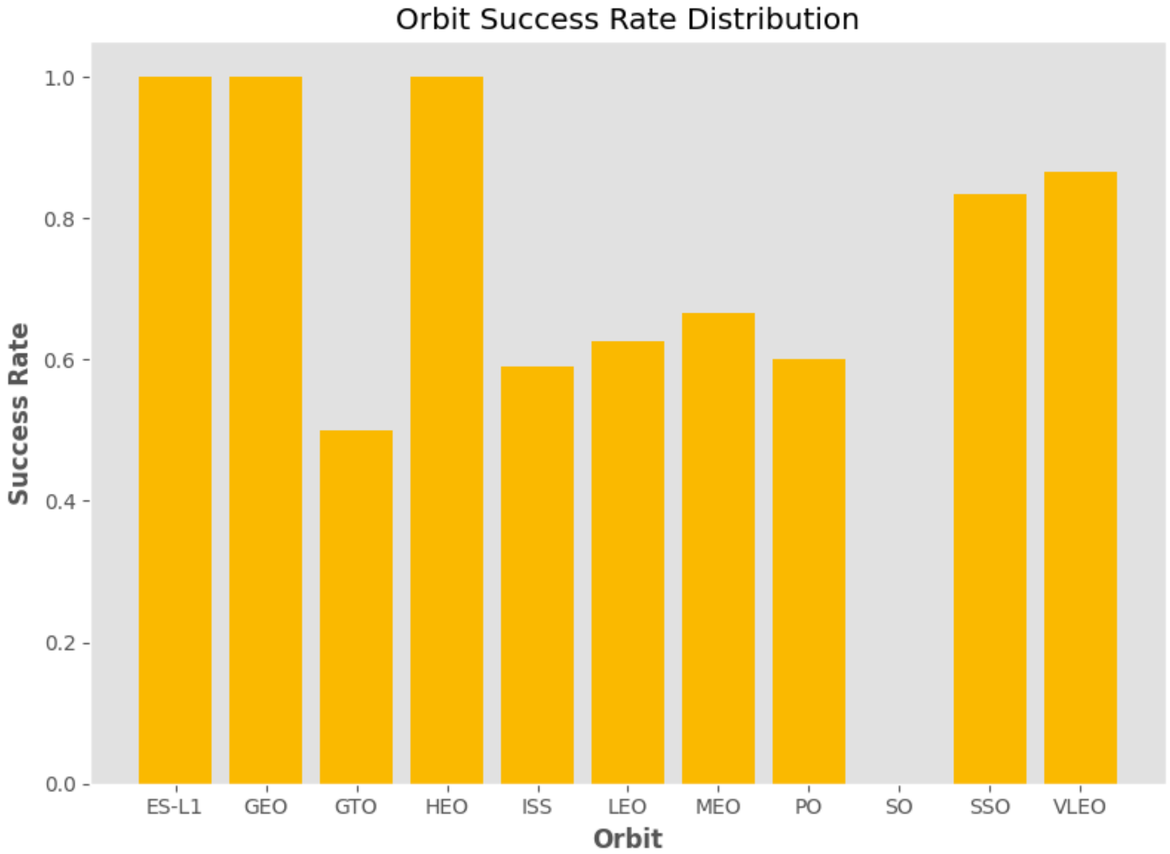
LaunchSite	sum	count	SuccessRate
CCSFS SLC 40	34	60	0.566667
KSC LC 39A	18	23	0.782609
VAFB SLC 4E	10	15	0.666667

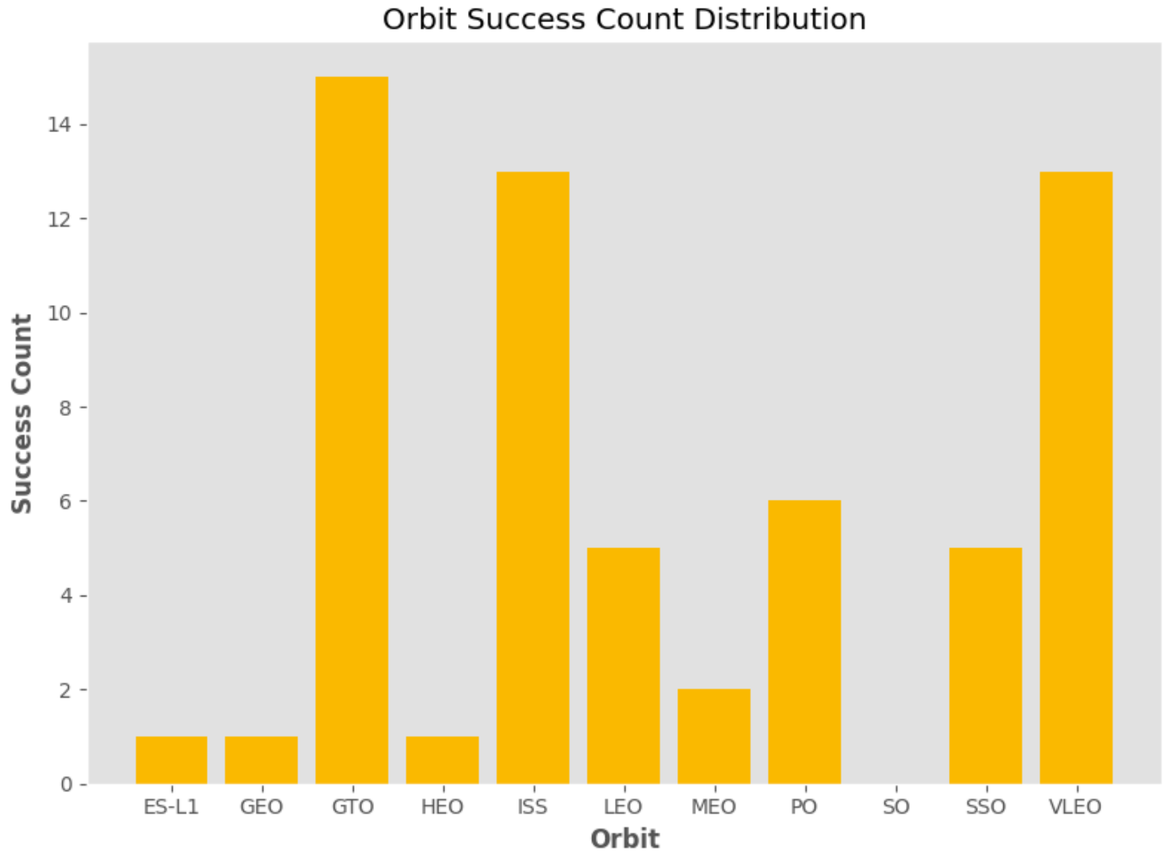
What is the success rate across the years ?



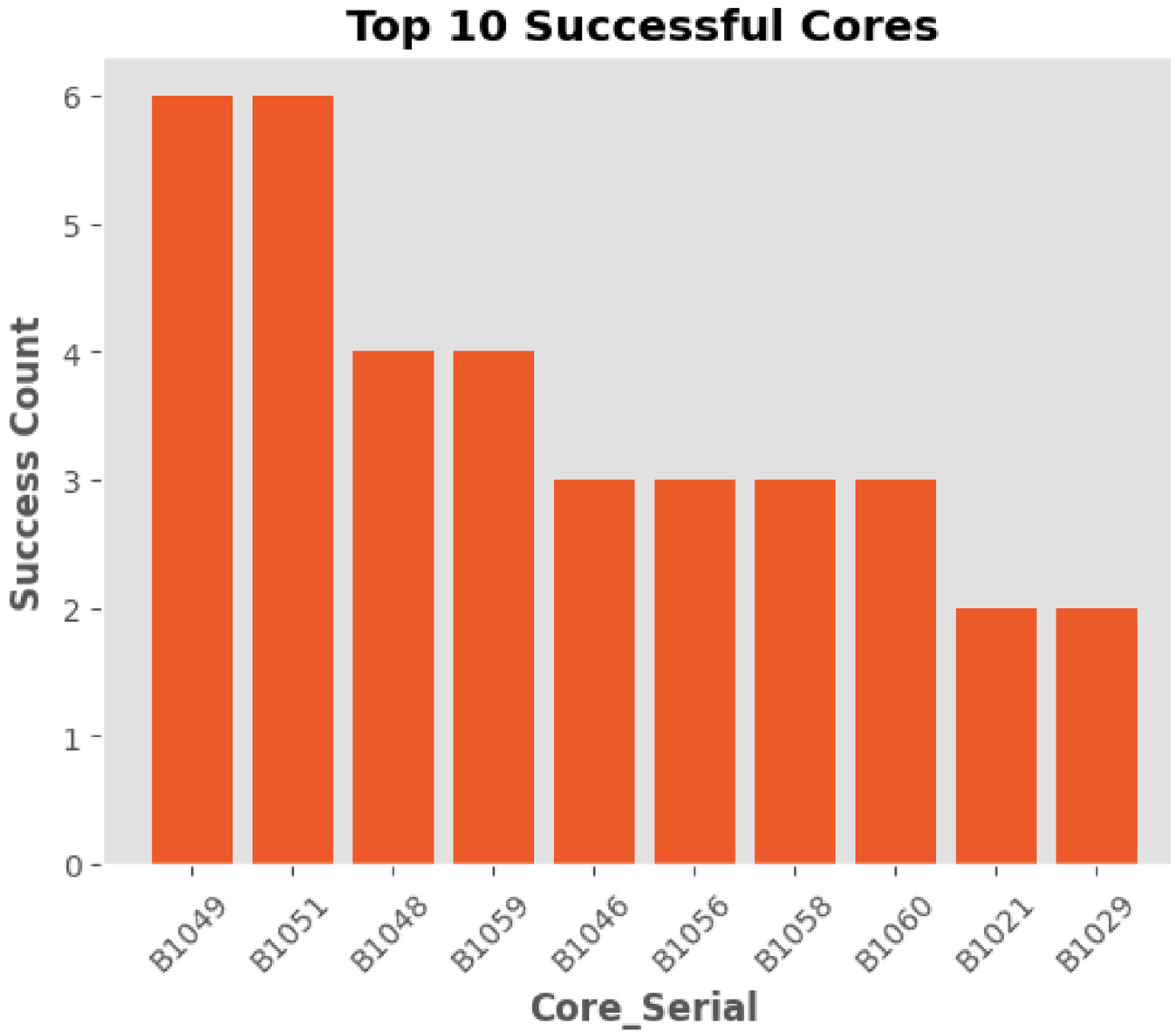
Date	Class
2010	**0.000000**
2012	**0.000000**
2013	**0.000000**
2014	0.333333
2015	0.285714
2016	0.555556
2017	0.833333
2018	0.550000
2019	0.909091
2020	0.850000

What Is the Correlation Between Orbit Type and Mission Success?



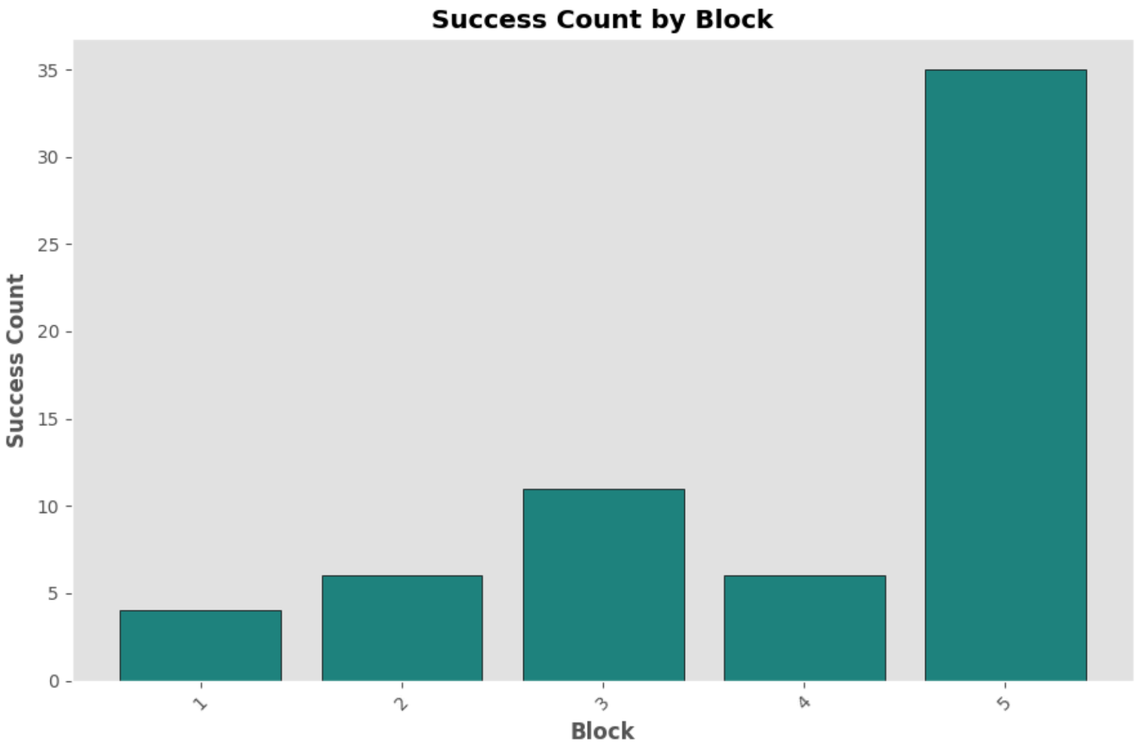


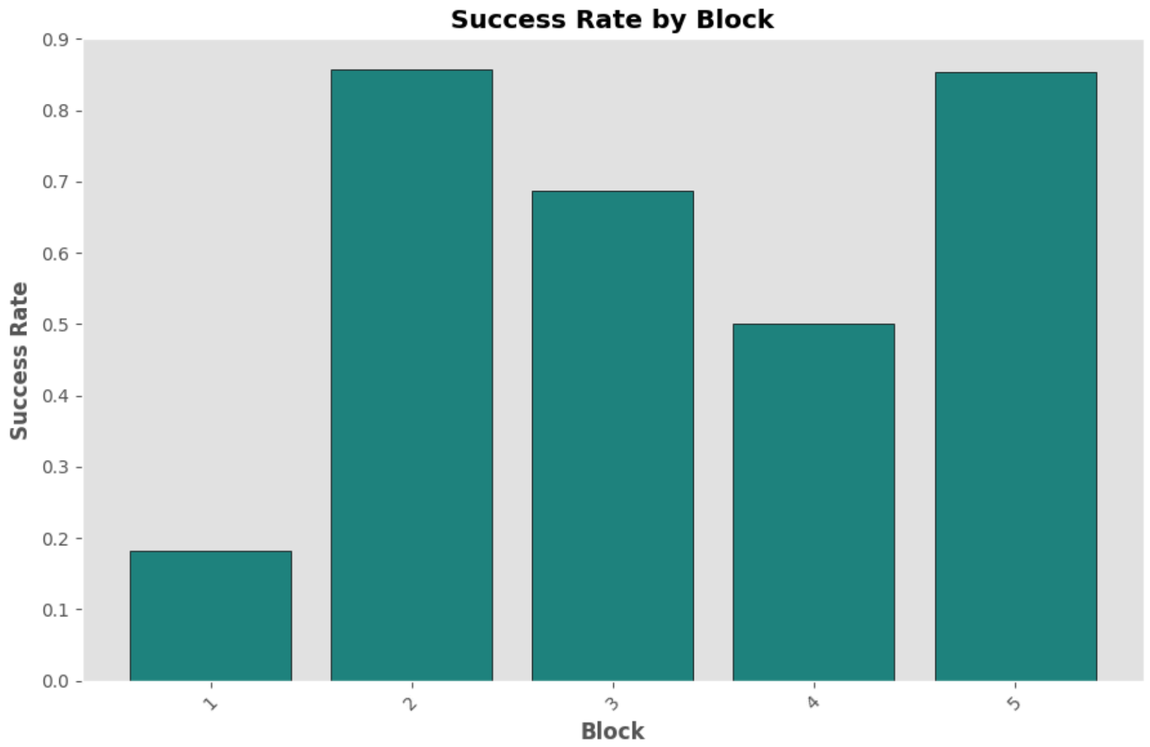
What are the top 10 types of cores that led to successful landings



Serial	Class
B1049	6
B1051	6
B1048	4
B1059	4
B1046	3
B1056	3
B1058	3
B1060	3
B1021	2
B1029	2

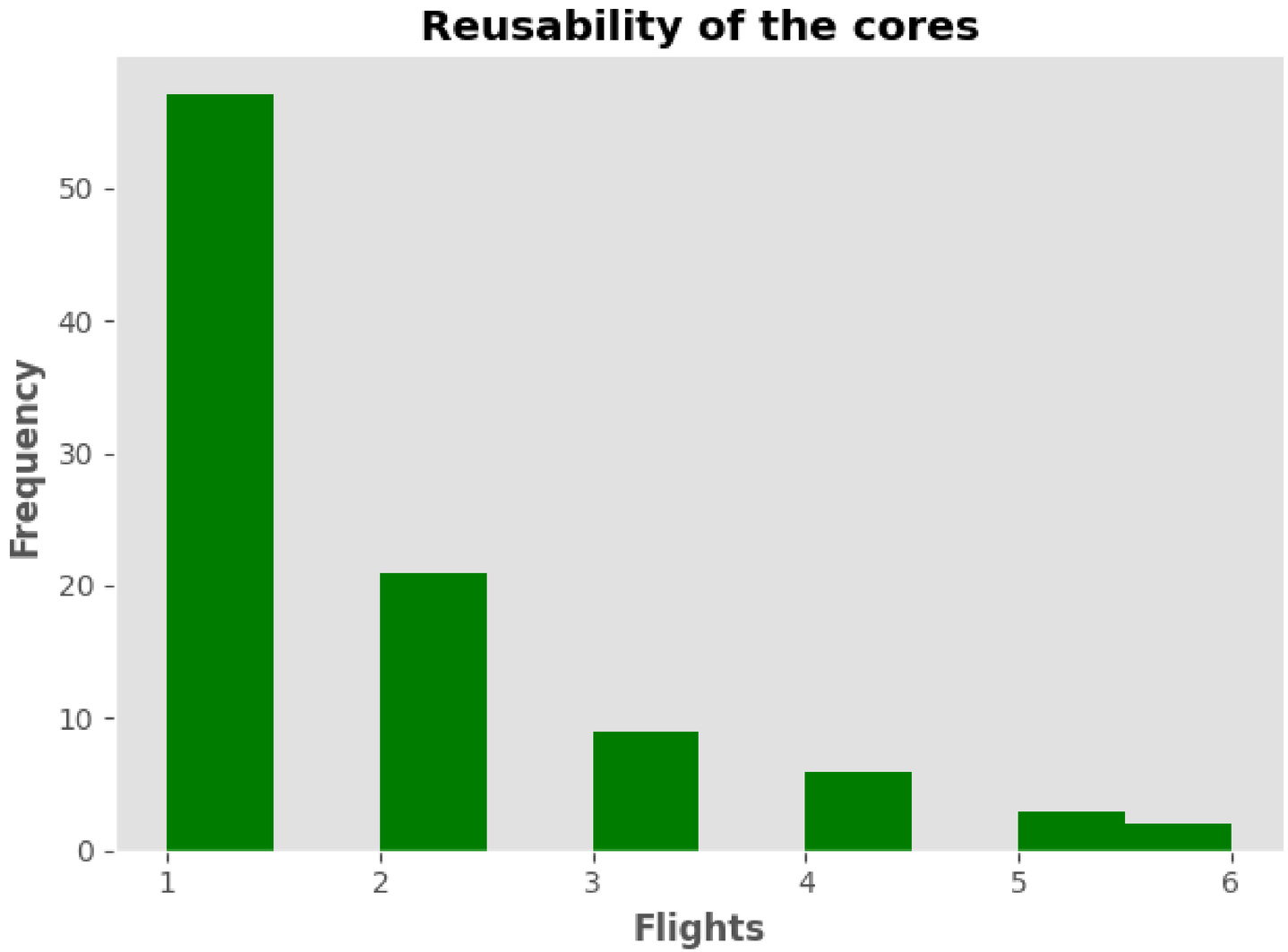
What are the blocks of these cores ?





this means if we want to launch a core it has to be of block 5

is there any reusable cores?



Interactive Visualization

interactive Maps using folium

Dashboard using Dash

Map using folium

Objectives

- determine the location of launchsites
- Launch site Success Count
- Calculate Proximities

Map using folium

Elements Used

- Map object
- Plugins
  - Marker Cluster
  - Mouse Position
- Features
  - DivIcon

Map using folium

Operation

- Creating the Map object using initial location which is taken to be the mean of all sites location
- Creating 3 Circles for each LaunchSite with marker on It in form of text which is the name of the LS
- add a marker Cluster
- add markers for success with green color and ones for failures with red color.

Map using folium

Operation

- Create a mouse postion object and add it to the map.
- Create a distance function
- Use mouse locations to determine proximity locations
- Use Distance function to calculate proximity distance

Does it Work ? !

DashBoard Using Dash

Objectives

- Display the Success rate of Each site using pie chart.
- Payloadmass Distrbution with Success / Failure acording to Booster version.

DashBoard using Dash

Elements

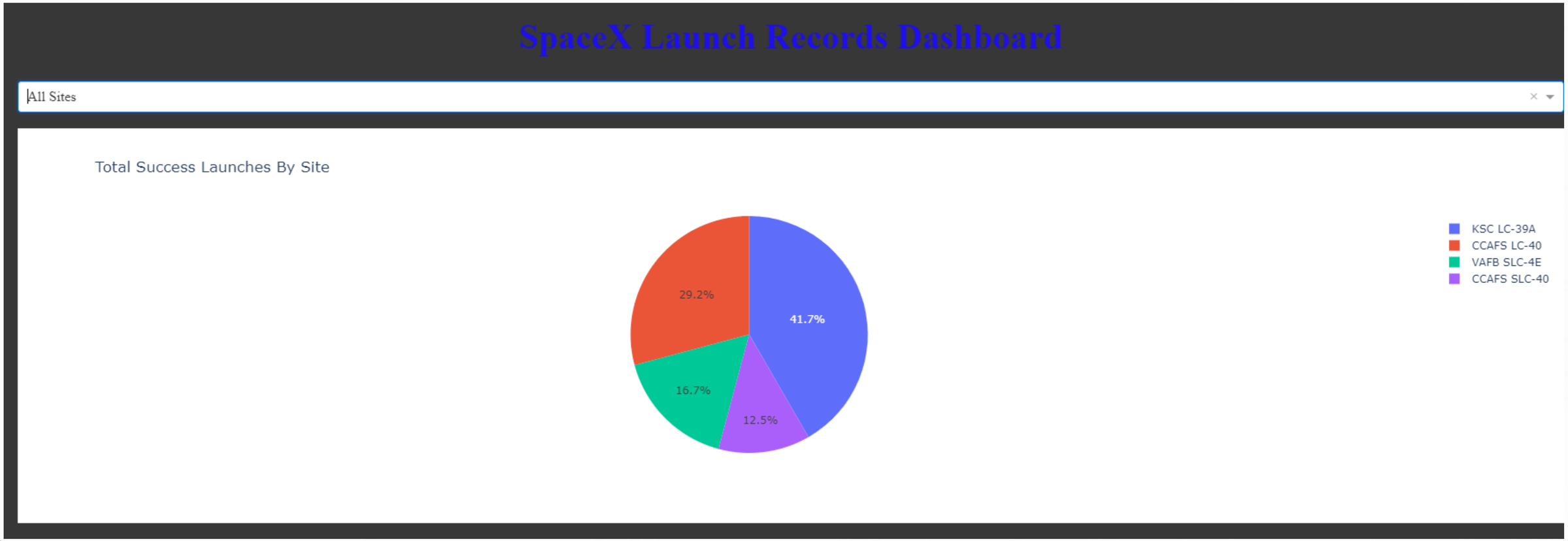
- Dash Library ,HTML ,dcc
- Input / OutPut for Callback
- plotly express for plotting

DashBoard using Dash

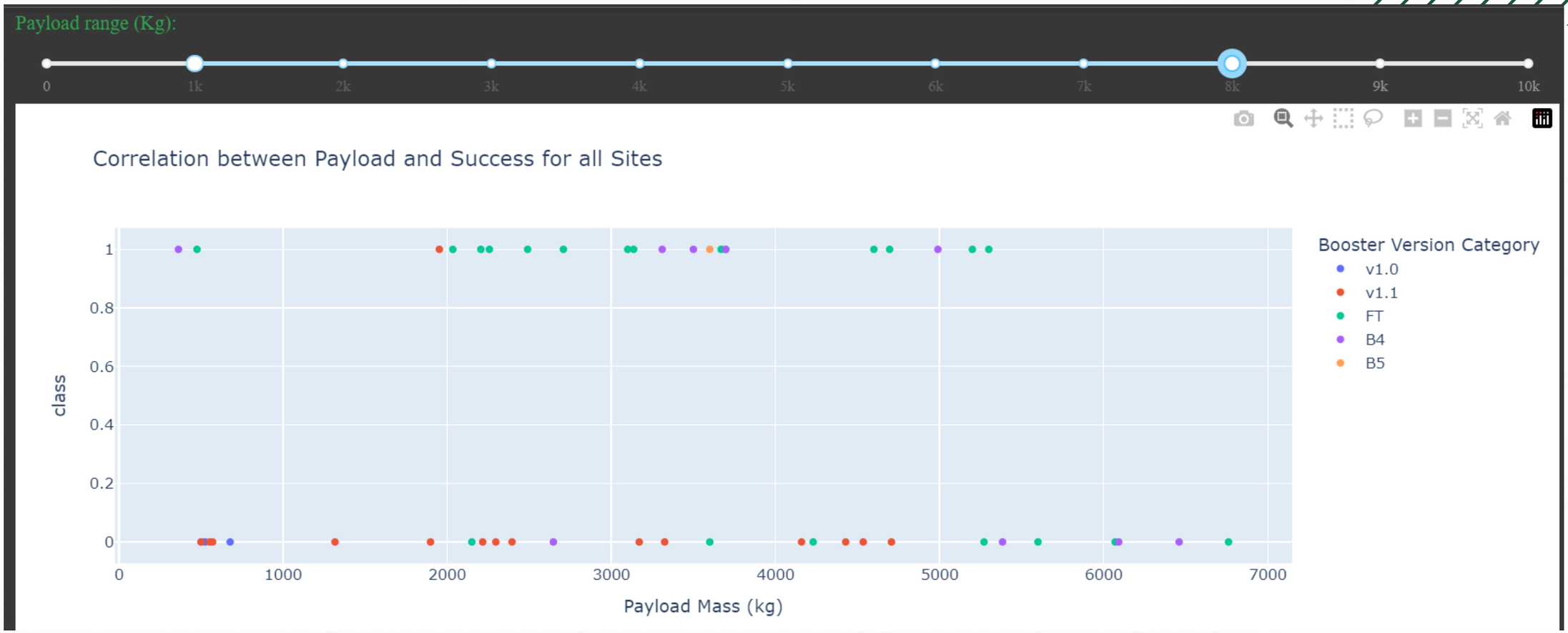
Operation

- Create a Dash Instant
- make layout
  - title
  - Dropdown box
  - slider
  - Graphs for PieChart and Scatter plot

DashBoard using Dash



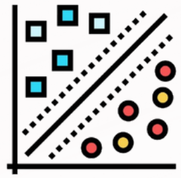
DashBoard using Dash



Does it Work ? !

ML Algorithms

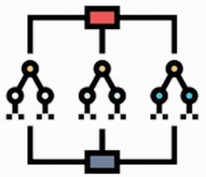
Logistic Regression



Support Vector Machine (SVM)

K-Nearest Neighbors (KNN)

Decision Tree



Random Forest

Preprocessing Phase

01 Split the data into features (X) and target (y)

02 Convert categorical features into numerical and use standardization

03 Split the data into train and test

04 Use Grid Search to find the best hyperparameters for each model

	Flight_Number	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	LandingPad	Block	ReusedCount	Serial	Longitude	Latitud
0	1	Falcon 9	5694.158673	LEO	CCSFS SLC 40	None - None	1	False	False	False	NaN	1.0	0	B0003	-80.577366	28.56185
1	2	Falcon 9	5694.158673	LEO	CCSFS SLC 40	None - None	1	False	False	False	NaN	1.0	0	B0004	-80.577366	28.56185
2	3	Falcon 9	525.000000	LEO	CCSFS SLC 40	None - None	1	False	False	False	NaN	1.0	0	B0005	-80.577366	28.56185
3	4	Falcon 9	800.000000	ISS	CCSFS SLC 40	None - None	1	False	False	False	NaN	1.0	0	B0006	-80.577366	28.56185
4	5	Falcon 9	677.000000	ISS	CCSFS SLC 40	None - None	1	False	False	False	NaN	1.0	0	B0007	-80.577366	28.56185

	Flight_Number	PayloadMass	Flights	GridFins	Reused	Legs	Block	ReusedCount	Longitude	Latitude	...	B1060	B1062	SLC 40	SLC 4E	LC 39A	JRTI-1	OCISLY	LZ-1
0	14	2395.0	1	1	0	1	1.0	0	-80.577366	28.561857	...	0	0	1	0	0	1	0	0
1	17	1898.0	1	1	0	1	1.0	0	-80.577366	28.561857	...	0	0	1	0	0	1	0	0
2	19	2477.0	1	1	0	1	1.0	0	-80.577366	28.561857	...	0	0	1	0	0	0	1	0
3	20	2034.0	1	1	0	1	1.0	0	-80.577366	28.561857	...	0	0	1	0	0	0	0	1
4	21	553.0	1	1	0	1	1.0	0	-120.610829	34.632093	...	0	0	0	1	0	0	0	0

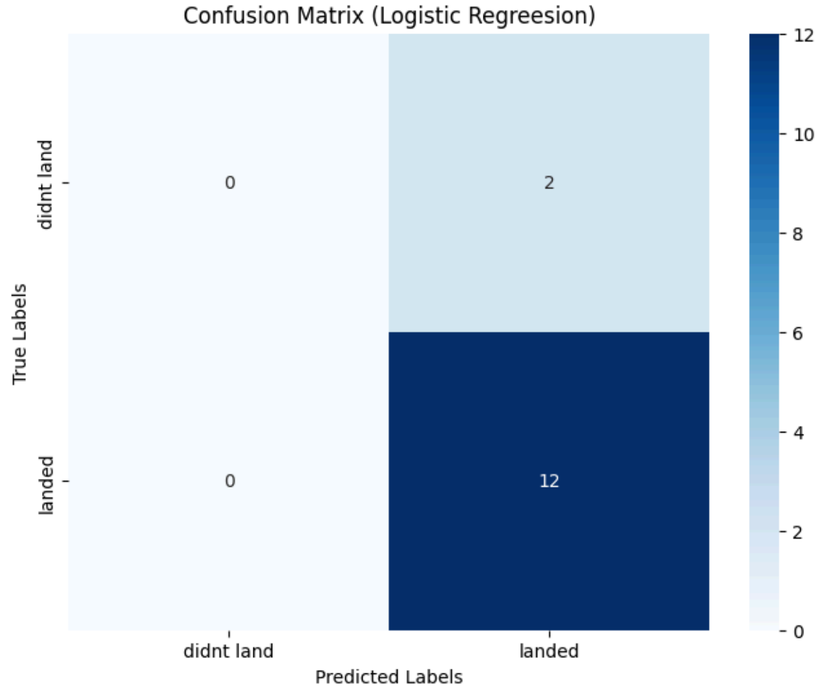
Logistic Regression

Best Hyperparameters

C: 0.1
Penalty: L2
Solver: lbfgs

Metrics

Training accuracy: 0.85
Testing accuracy: 0.86
Precision: 0.86
Recall: 1.00
F1 Score: 0.92



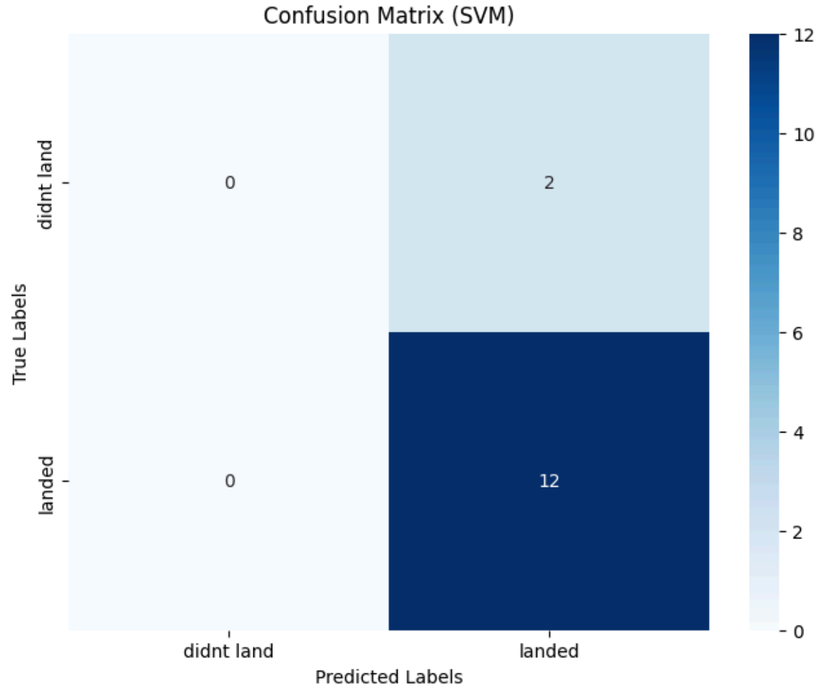
Support Vector Machine (SVM)

Best Hyperparameters

C: 0.001
gamma: 0.001
kernel: linear

Metrics

Training accuracy: 0.85
Testing accuracy: 0.86
Precision: 0.86
Recall: 1.00
F1 Score: 0.92



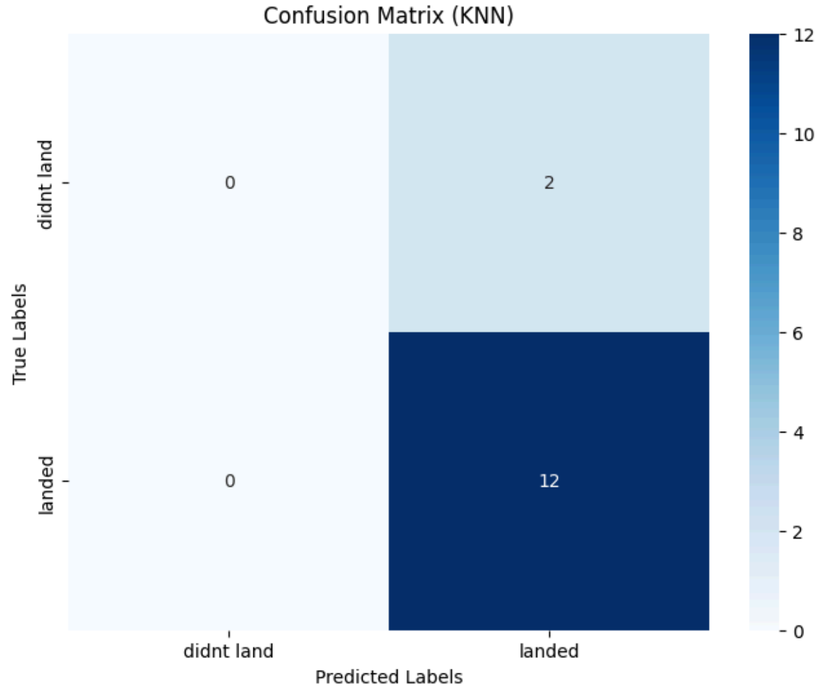
K-Nearest Neighbors (KNN)

Best Hyperparameters

algorithm: auto
n_neighbors: 3
p: 2

Metrics

Training accuracy: 0.85
Testing accuracy: 0.86
Precision: 0.86
Recall: 1.00
F1 Score: 0.92



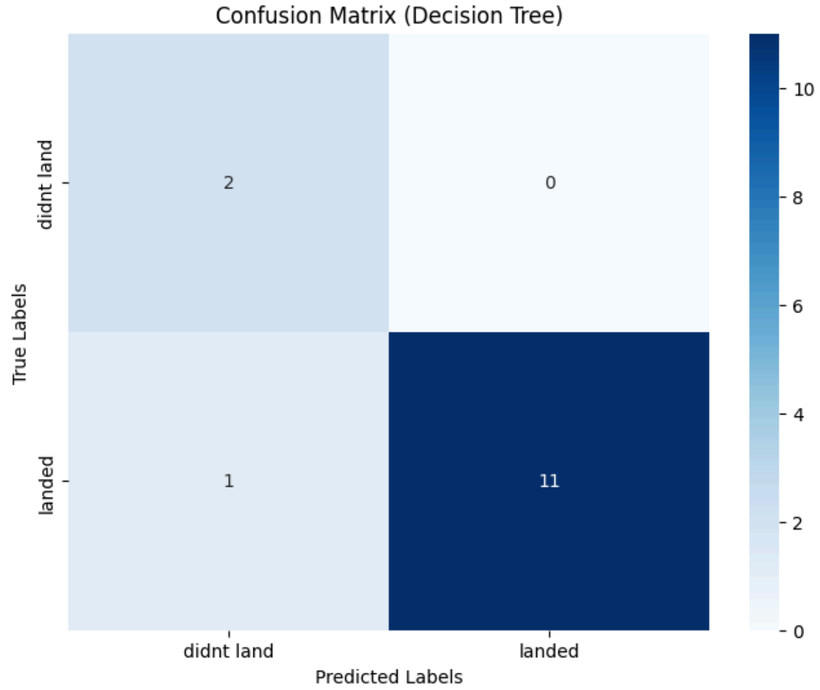
Decision Tree

Best Hyperparameters

criterion: entropy
max_depth: 6
max_features: sqrt
min_samples_leaf: 2
min_samples_split: 10
splitter: best

Metrics

Training accuracy: 0.9267
Testing accuracy: 0.93
Precision: 1.00
Recall: 0.92
F1 Score: 0.96



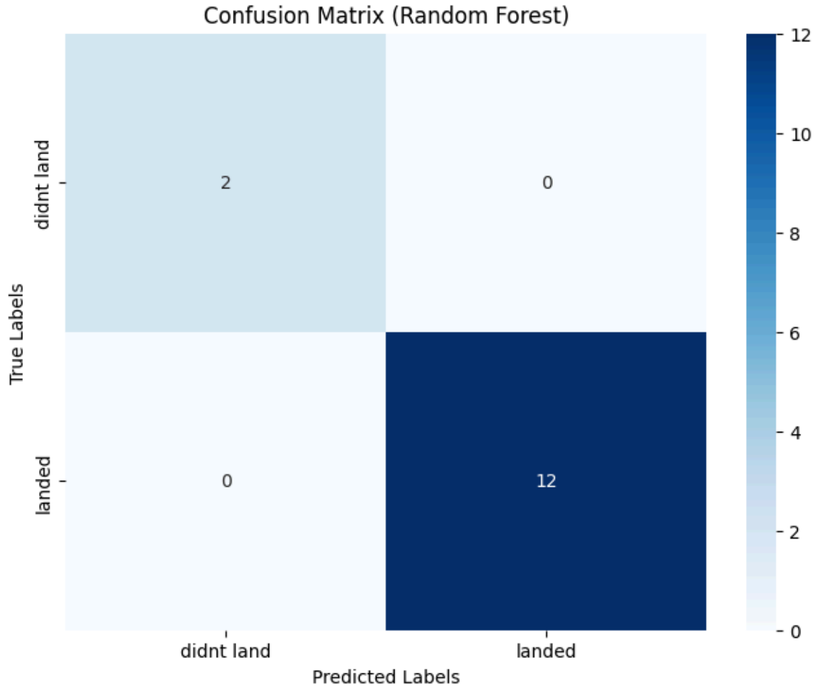
Random Forest

Best Hyperparameters

criterion: gini
max_depth: 10
max_features: sqrt
min_samples_leaf: 1
min_samples_split: 2
n_estimators: 100

Metrics

Training accuracy: 0.81
Testing accuracy: 1.00
Precision: 1.00
Recall: 1.00
F1 Score: 1.00



ML FLOW & UI

Connect to the ml flow server

log model to ml flow

UI using Gradio

- !pip install pyngrok
- !pip install mlflow
- !pip install gradio

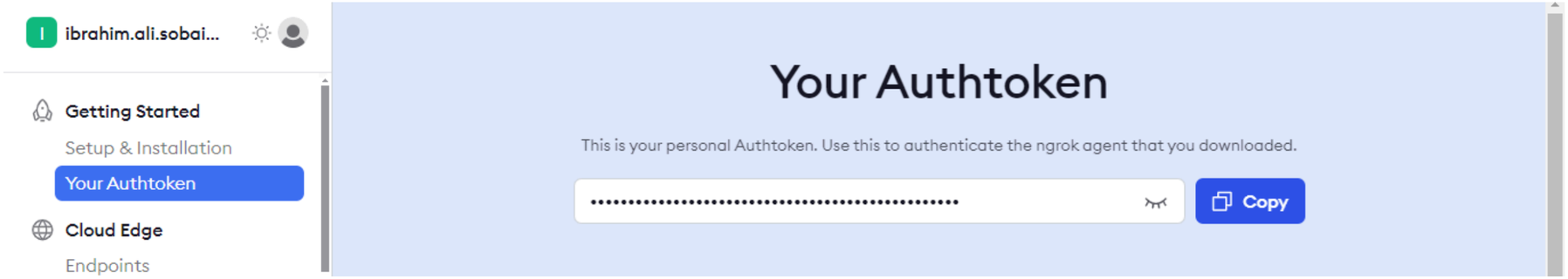
Upload launches_data_ML

- Connect to the mlflow server

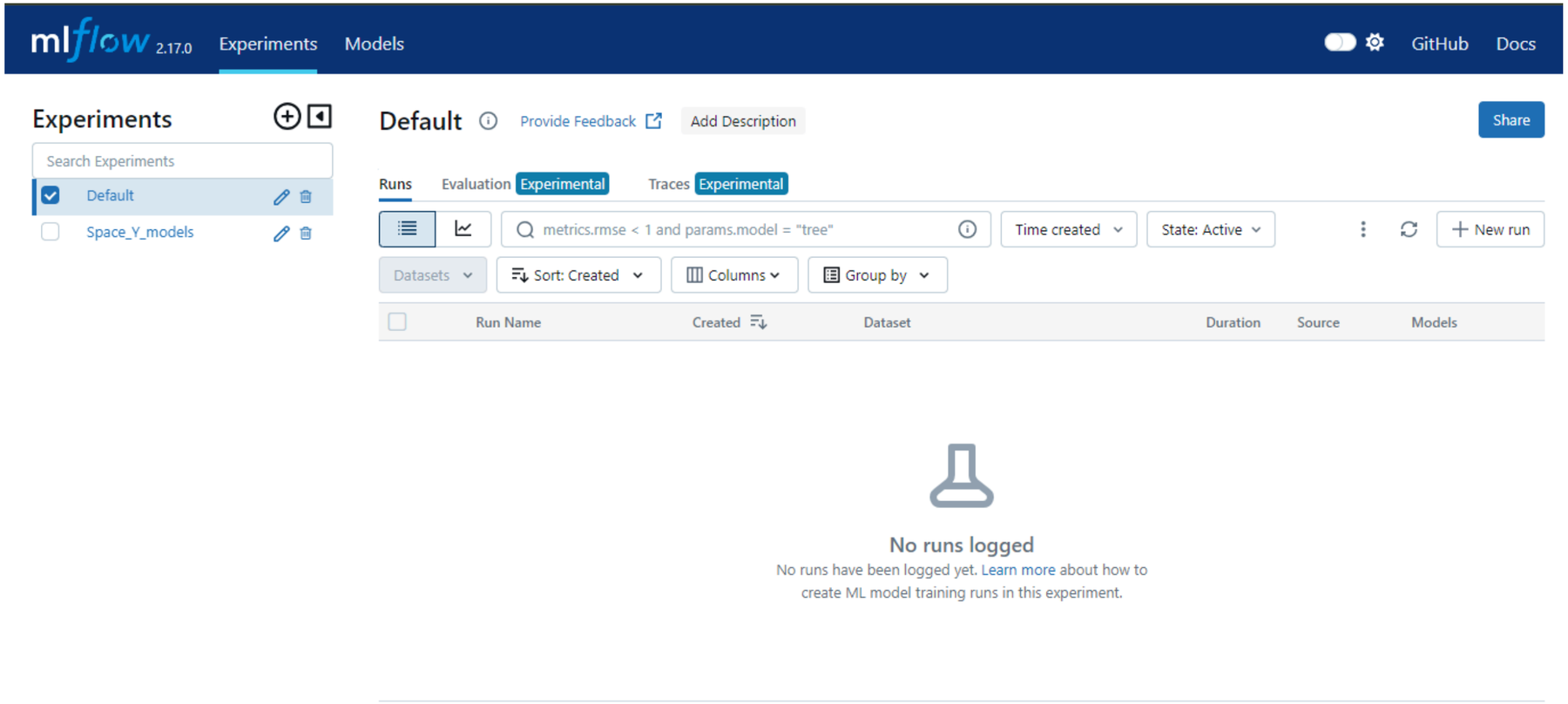
```
mlflow.set_experiment("Space_Y_models")
```

Space_Y_models

Enter your authtoken



connect agent to the server



log model to ml flow

- Log hyperparameters
- Log metrics
- Log the model

UI using Gradio

since our inputs are a lot we can categorize them in 5 categories(tabs)

- **Basic flight info** : payloadmass , number of flughts
- **Rocket Features** : griffins , legs,reused , reused count
- **Launch and Lading info** : longitude , latitude, launchsize,LandingPad
- **Orbital info** : orbit
- **Booster serial numbe**r : serial

Basic Flight Info | Rocket Features | Launch and Landing Info | Orbital Info | Booster Serial Numbers

Payload Mass (kg)

Number of Flights

Prediction

Submit

Basic Flight Info | Rocket Features | Launch and Landing Info | Orbital Info | Booster Serial Numbers

- [ ] Grid Fins
- [ ] Reused
- [ ] Legs

Block (1-5)

1

Reused Count

0

Prediction

Submit

Basic Flight Info | Rocket Features | Launch and Landing Info | Orbital Info | Booster Serial Numbers

Longitude

.

Latitude

.

Launch Site

CCSFS SLC 40

Landing Pad

JRTI-1

Prediction

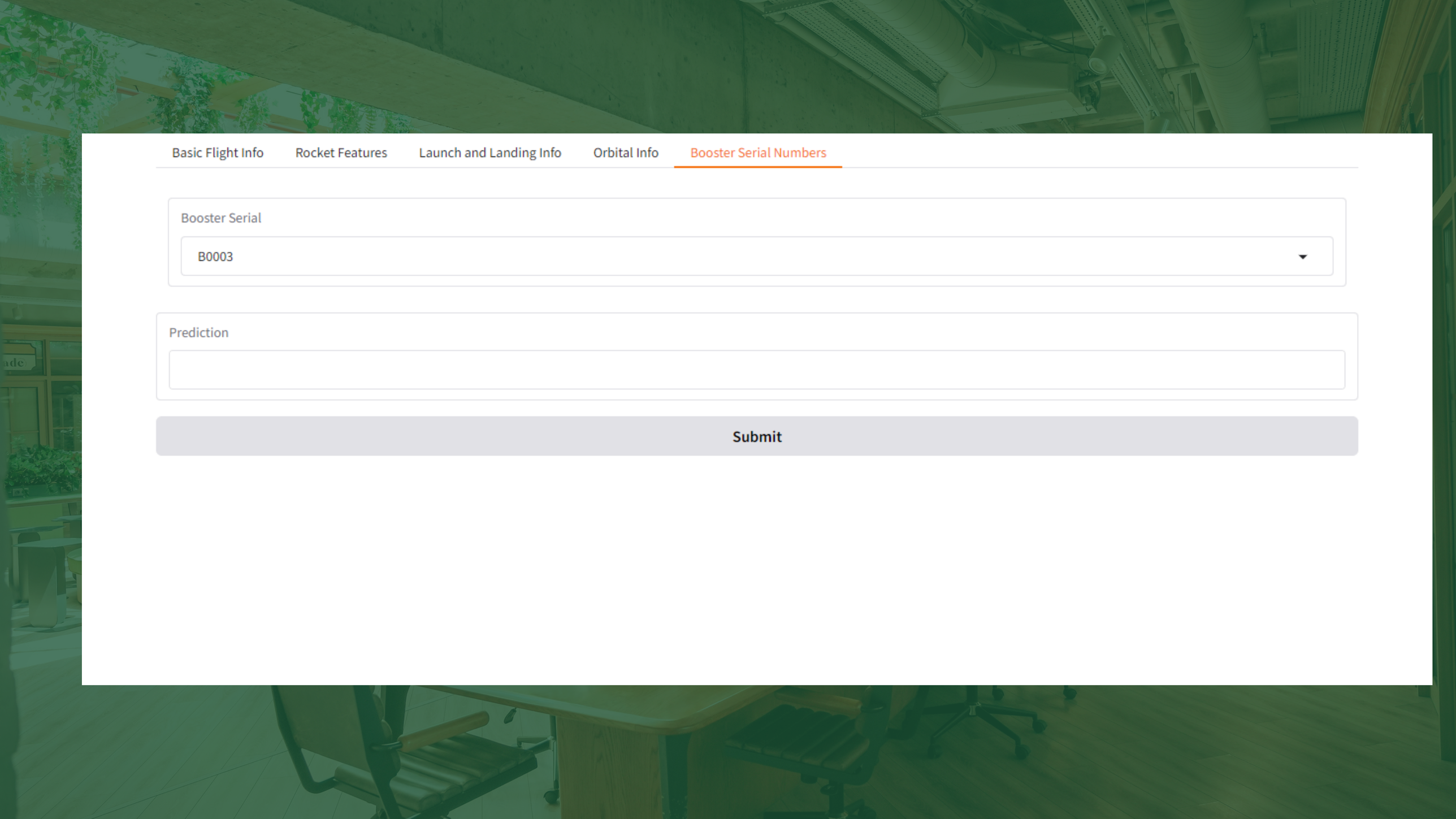
Basic Flight Info | Rocket Features | Launch and Landing Info | Orbital Info | Booster Serial Numbers

Orbit

LEO

Prediction

Submit


Basic Flight Info
Rocket Features
Launch and Landing Info
Orbital Info
Booster Serial Numbers
Booster Serial
B0003
Prediction
Submit