



IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

Abiyselassie Gashaw
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Outline

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

Executive Summary

Summary of methodologies

- Data Collection through API
- Data Collection with Webscraping
- Data Wrangling
- Exploratory Data Analysis with SQL
- Exploratory Data Analysis with Data Visualization
- Interactive Visual Analytics with Folium
- Interactive Dashboard with Plotly Dash
- Machine Learning Predictions

Summary of all results

- Exploratory Data Analysis Result
- Interactive Analytics in Screenshots
- Predictive Analytics Result

Introduction

Project background and context

- SpaceX has been the most successful in making space travel affordable for everyone over decades accomplishing tasks like sending spacecraft to the International Space Station, providing satellite Internet access & sending manned missions to Space.
- It advertises on its website, Falcon 9 rocket launches, which are known for incorporating several reusable technologies that have revolutionized space travel by significantly reducing the cost of launching payloads into space.
- It launches with a relatively inexpensive cost of 62 million dollars; but other rocket providers cost upward of 165 million dollars each.
- Much of its significant cost savings is due to SpaceX's Falcon 9 innovative technology to reuse 1st stage.
- Therefore, if the '1st stage will land' can be determined, the cost of each launch can also be determined.

Problems to find answers

- Predict if the 1st stage of Falcon 9 will land successfully.
- Predict if SpaceX will reuse the 1st stage if the cost of each launch can be determined.
- Determine what attributes (payload mass, flight No., Orbit, etc.) are correlated with successful landings.
- Identify factors that affect successful 1st stage landing of Falcon 9.
- Detect ML model that perform the best for predicting the successful landing.

Section 1

Methodology

Methodology

Executive Summary

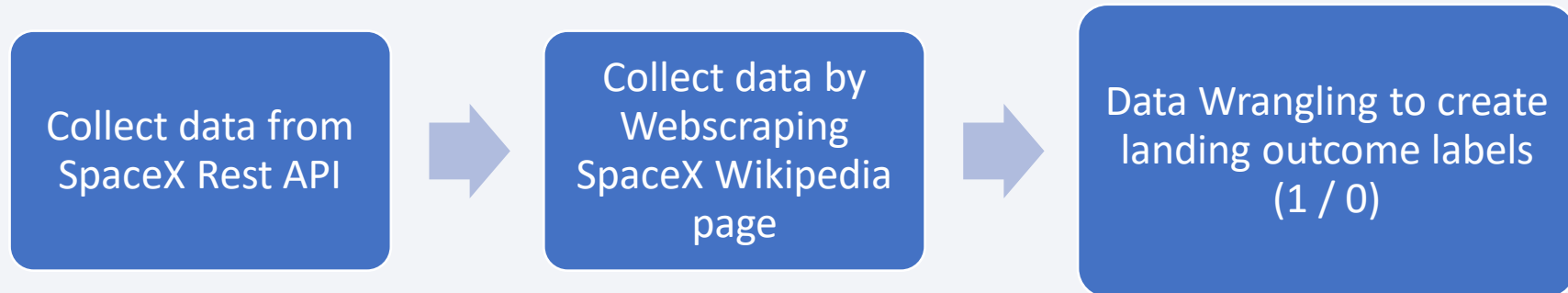
- **Data collection methodology:**
 - The data were collected from Public source using SpaceX API.
 - Falcon 9 launch records were webscraped from Wikipedia page using Beautiful Soup library.
- **Perform Data Wrangling:**
 - The data was wrangled to determine Success rate of Landing class & all successful landings were created.
- **Perform Exploratory Data Analysis (EDA) using Data Visualization and SQL**
 - Scatterplots, line plots & bar chart were plotted to identify factors for successful landing & yearly trend.
 - Total payload mass, Total No. of successful & failure mission outcomes, etc. were selected from SpaceXTBL
- **Perform Interactive Visual Analytics using Folium and Plotly Dash**
 - An Interactive Map with Folium was built to visualize a launch site proximity to geographical markers.
 - A Dashboard with Plotly Dash (Pie chart, Scatterplot, Payload range slider interaction) was built to visualize launch sites with most success & payload ranges.
- **Perform Predictive Analysis using Classification Models**
 - Successful landing outcomes were predicted by building ML models(LR, SVM, Decision Tree, KNN), tuning Hyperparameters & evaluating based on accuracy score and confusion matrix

Data Collection

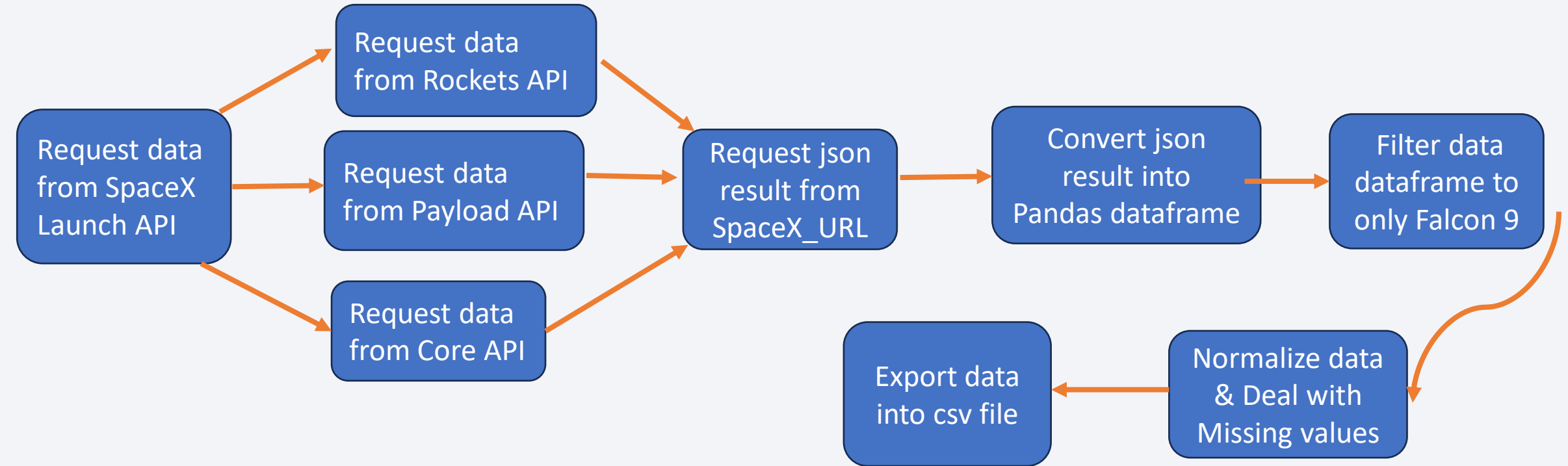
- The data were collected from Public source of SpaceX API &
 - Launch data was obtained using the get requests
- The data were collected by webscraping SpaceX Wikipedia page (updated on 9th June 2021)
 - Falcon 9 Launch data, another popular data source was obtained using Beautiful Soup library
- The data obtained from SpaceX API & Webscraping were wrangled
 - A landing outcome label was created from the Outcome column
 - Success rate of Landing class was determined

Landing outcome labels were trained with

- 1 means the booster successfully landed
- 0 means it was unsuccessful

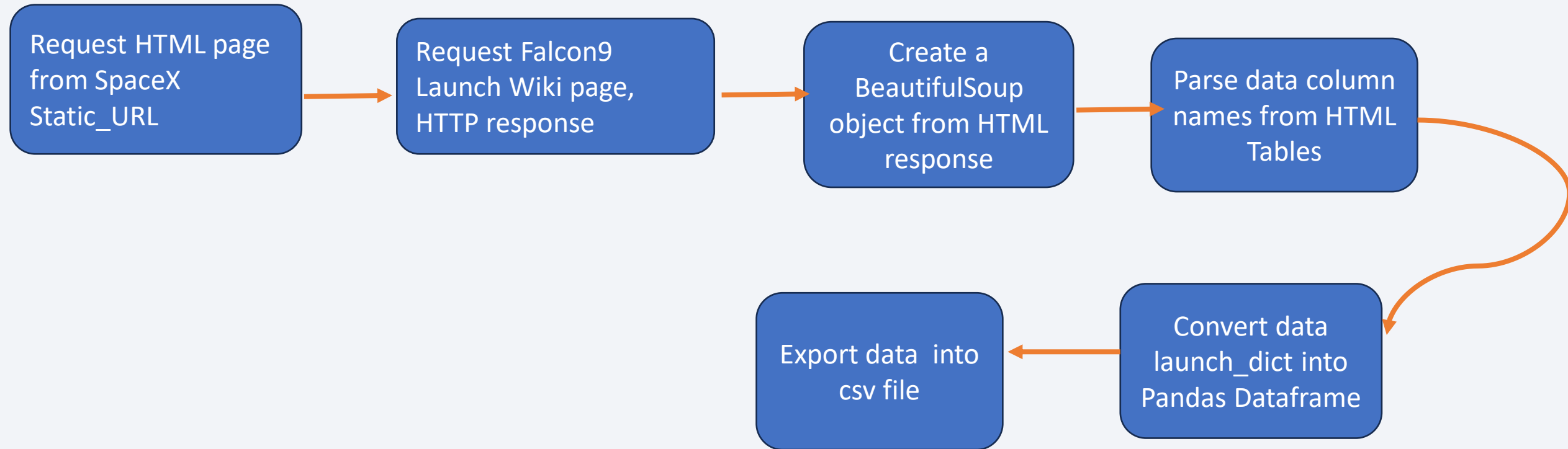


Data Collection – SpaceX API



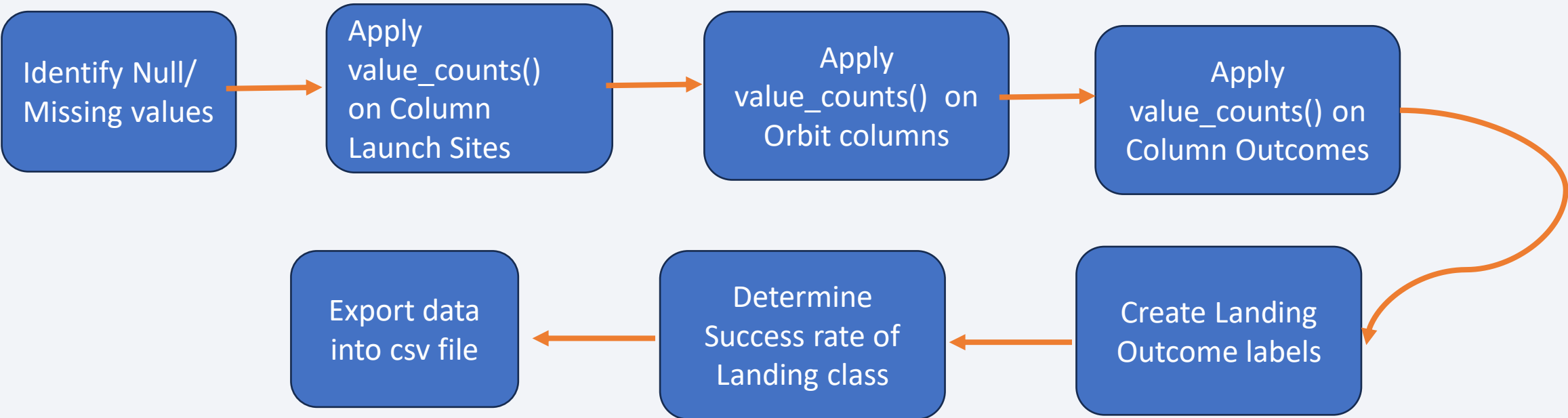
GitHub URL of the completed **SpaceX API calls** notebook
([https://github.com/abiyselassie22/testpro/blob/master/jupyter-labs-spacex-data-collection-api%20\(1\).ipynb](https://github.com/abiyselassie22/testpro/blob/master/jupyter-labs-spacex-data-collection-api%20(1).ipynb))

Data Collection - Scrapping



GitHub URL of the completed **web scraping** notebook
(<https://github.com/abiyselassie22/testpro/blob/master/jupyter-labs-webscraping.ipynb>)

Data Wrangling



GitHub URL of your completed **data wrangling** related notebooks
(https://github.com/abiyselassie22/testpro/blob/master/IBM-DS0321EN-SkillsNetwork_labs_module_1_L3_labs-jupyter-spacex-data_wrangling_jupyterlite.jupyterlite.ipynb)

EDA with Data Visualization

The following **attributes/features** can be used to determine if **Falcon 9's 1st stage** can be reused. **Scatterplots** were plotted to show if there is any relationship between

- Flight Numbers & Launch Site;
- Payload & Launch Site;
- Flight Number & Orbit type;
- Payload & Orbit type

✓ **1st stage** is more likely to land **successfully** when **Flight No.** (continuous launch attempts) increases.

Bar chart

- was plotted to show if there are any relationship between **success rate** & each **orbit type**

Line chart

- was drawn to show the **increasing yearly trend** of **launch success rates**.

GitHub URL of your completed EDA with data visualization notebook

([https://github.com/abiyselassie22/testpro/blob/master/IBM-DS0321EN-SkillsNetwork labs module 2 jupyter-labs-eda-dataviz.ipynb](https://github.com/abiyselassie22/testpro/blob/master/IBM-DS0321EN-SkillsNetwork%20labs%20module%20jupyter-labs-eda-dataviz.ipynb))

EDA with SQL

- Names of the **unique launch sites** were selected from the space mission **SPACEXTBL**
- **5 Records** of **launch sites** beginning with the string '**CCA**' were selected from **SPACEXTBL**
- Total payload mass carried by **boosters** launched by **NASA** (**CRS**) were selected
- Average payload mass carried by **booster version F9 v1.1** was selected.
- Date when the **first successful landing outcome** in ground pad was achieved was selected.
- Names of **boosters** with **success** in drone ship & payload mass **>= 4000 & <= 6000** were selected
- Total No. of **successful** & **failure mission outcomes** were selected
- Names of the **booster_versions** which have carried the maximum payload mass were selected.
- Count of **landing Failure** / **Success outcomes** b/n date **2010-06-04** & **2017-03-20** were ranked

GitHub URL of your completed **EDA** with **SQL** notebook

(https://github.com/abiyselassie22/testpro/blob/master/jupyter-labs-eda-sql-coursera_sqlite.ipynb)

Build an Interactive Map with Folium

- **map.add_child(circle)** & **map.add_child(marker)** objects were created & added to folium map to make the map look less congested & indicate points or areas of our interest.
- **MarkerCluster object** was added to simplify the folium map containing many markers having the same coordinate & group the multiple nearby markers into clusters based on proximity. This is used to mark all launch sites, success/failed launches for each site on an interactive map.
- **MousePosition object** was added on the folium map to get coordinate for a mouse over a point on the map & to easily find coordinates of any points of interests (e.g. Launch site).
- **Map.add_child(lines)** objects were used to add a polyline to the Folium map to visualize connections / draw distances between a launch site & their close proximities (selected coastline, railway, highway, city, etc.). This can be used to choose an optimal launch site.

GitHub URL of your completed **Interactive map with Folium map**

(https://nbviewer.org/github/abiyselassie22/testpro/blob/master/IBM-DS0321EN-SkillsNetwork_labs_module_3_lab_jupyter_launch_site_location.jupyterlite.ipynb)

Build a Dashboard with Plotly Dash

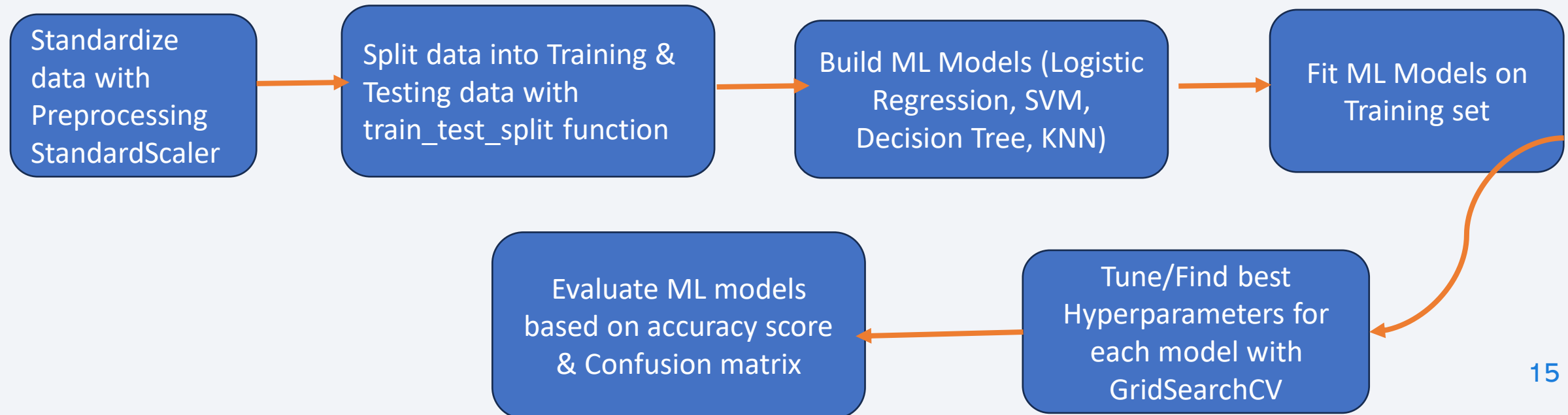
- **SpaceX Launch Records Dashboard** with **dropdown list** was created to let us select **all sites** and different launch sites.
- **Pie charts** for **all sites** & **selected sites** were rendered & added to the **SpaceX Launch Records Dashboard** to visualize the **launch success counts** for **all sites** & **entered sites**.
- **Payload range slider interaction** was rendered & added to the **SpaceX Launch Records Dashboard** to find if selection of **variable payloads** is correlated with the **mission outcome**.
- **Scatterplot chart** were rendered & added to the **SpaceX Launch Records Dashboard** to visualize how **payload** may be correlated with **mission outcomes** for **all sites** & **selected site(s)**.

GitHub URL of your **completed Plotly Dash** lab

(https://github.com/abiyselassie22/testpro/blob/master/spacex_dash_app.py)

Predictive Analysis (Classification)

GitHub URL of your completed **predictive analysis** lab,
https://github.com/abiyselassie22/testpro/blob/master/IBM-DS0321EN-SkillsNetwork_labs_module_4_SpaceX_Machine_Learning_Prediction_Part_5.jupyterlite.ipynb



Results

- **Exploratory Data Analysis results**
 - Visualization
 - SQL
- **Interactive Analytics Demo in screenshots**
 - Folium
 - Dashboard with Plotly Dash
- **Predictive Analysis results**
 - Machine Learning Prediction
 - Classification Accuracy
 - Confusion Matrix

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 blue and red, creating a sense of motion or data flow. A faint, light blue grid pattern is also visible, particularly in the lower-left quadrant. The overall effect is high-tech and digital.

Section 2

Insights drawn from EDA

Flight Number vs. Launch Site

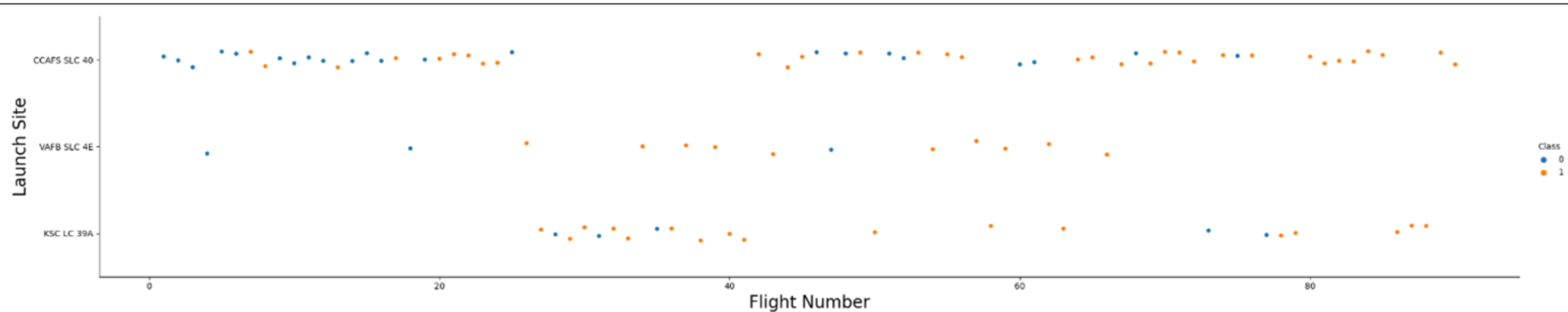


Figure 1. Scatterplot showing the relationship between **Flight Number & Launch Site**

- In **CCAFS LC-40 launch site**, the **Success** appears related to the **Nos. of flights**;
✓ as the No. of **Success** labeled **class 1** increases with increase in **Flight Nos.**
- While in **KSC LC-39A & VAFB SLC 4E launch sites**, **success** have no relationship with **Flight Nos.**.
✓ as the No. of their **Successes** show no difference when **Flight Nos.** increases.

Payload vs. Launch Site

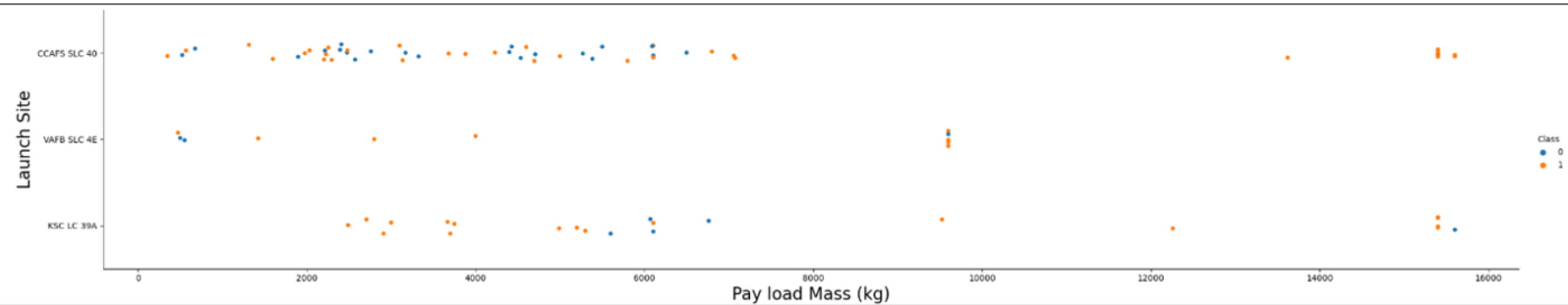


Figure 2. Scatterplot showing the relationship between **Payload** & **Launch Site**

The **Payload** vs. **Launch Site** **scatter point chart** shows that

- **Lighter weighted payloads** (1000-4000kg) has a **success rate** of **100%** for the **launch site VAFB-SLC**.
- ✓ there are no rockets launched for **heavy payload mass(>10000)** for **VAFB-SLC launch site**.
- **CCAFS LC-40** has a **success rate** of **60%**,
- ✓ but if the **mass** is above **> 10,000 kg**, the **success rate** is **100% (Class 1)**.

Success Rate vs. Orbit Type

- The plotted bar chart shows that
 - ✓ orbits ES-L1, GEO, HEO, & SSO have highest success rates, 100%
 - ✓ orbit VLEO, 90%.
 - ✓ Orbit, ISS, 60%
having lowest success rate

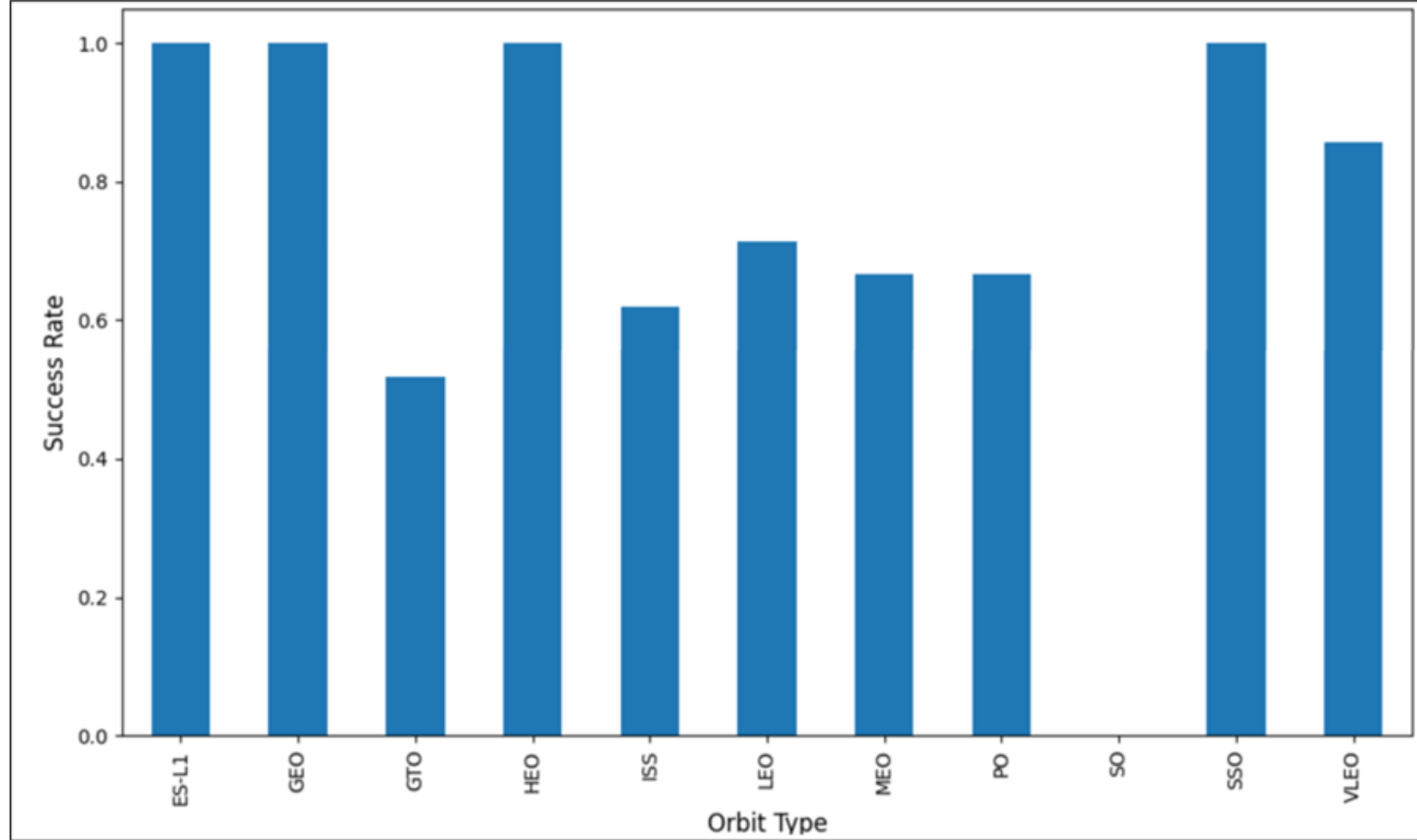


Figure 3. Bar chart showing the relationship between **Success Rate** & **Orbit Type**.

Flight Number vs. Orbit Type

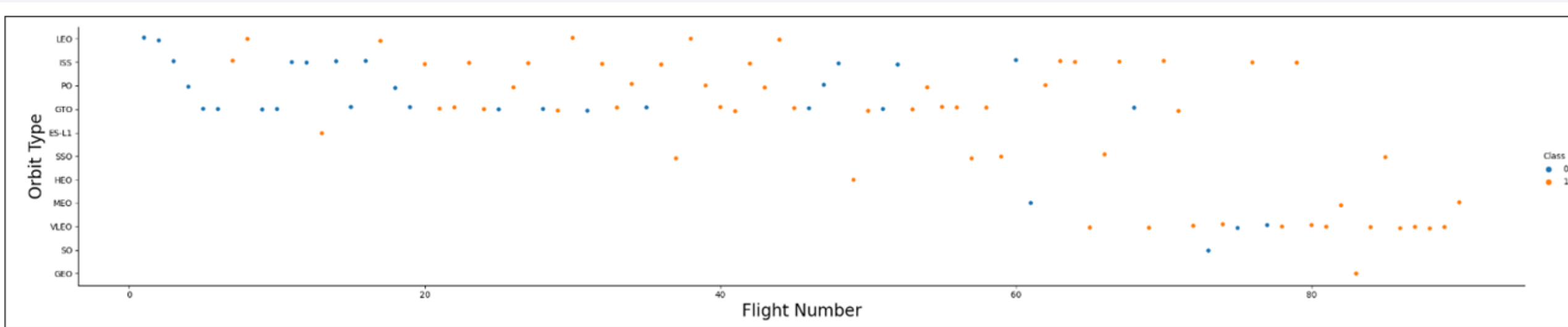


Figure 4. Scatterplot showing the relationship between **Orbit Type** & **Flight Number**.

- The **Scatterplot** shows that the **Success** appears related to the Nos. of **flights** in the **LEO orbit**;
✓ as the No. of **Success** labeled **class 1** increases with increase in **Flight Nos.**
- while there seems to be no relationship between **success rate** & **Flight Nos.** in **GTO orbit**.
✓ as the No. of its **Success** shows no difference when **Flight Nos.** increases.

Payload vs. Orbit Type

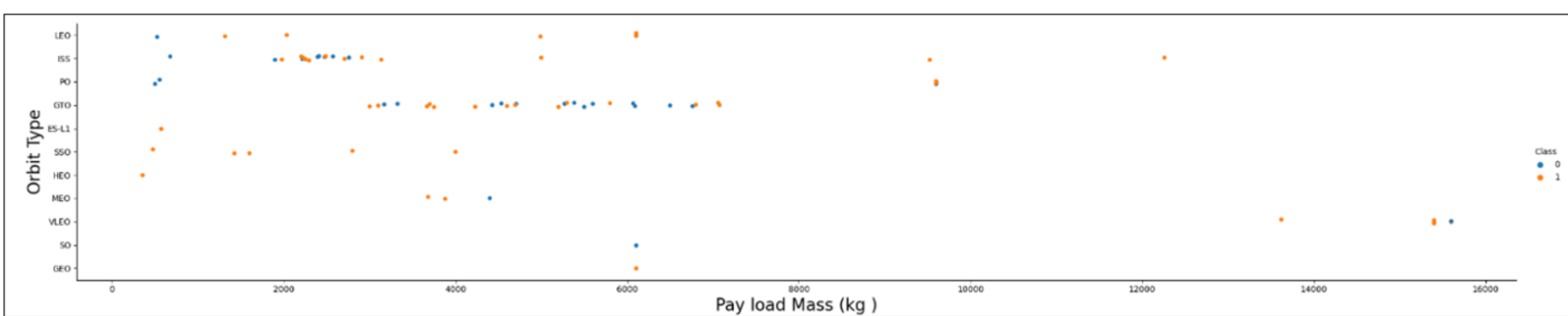


Figure 5. Scatterplot showing the relationship between **Payload** & **Orbit Type**.

- The **scatterplot** shows that with **heavy payloads**, the **successful landing** / **positive landing rate** are more for **Polar (PO)**, **LEO** & **ISS**.
- However, **successful landing** / **positive landing rate** for **GTO** cannot be distinguished as both positive **landing rate** & **negative landing rate** (**unsuccessful mission**).

Launch Success Yearly Trend

The **line chart**

- shows that the **Launch success rate** since **2013** has kept **increasing** over time till **2020**.

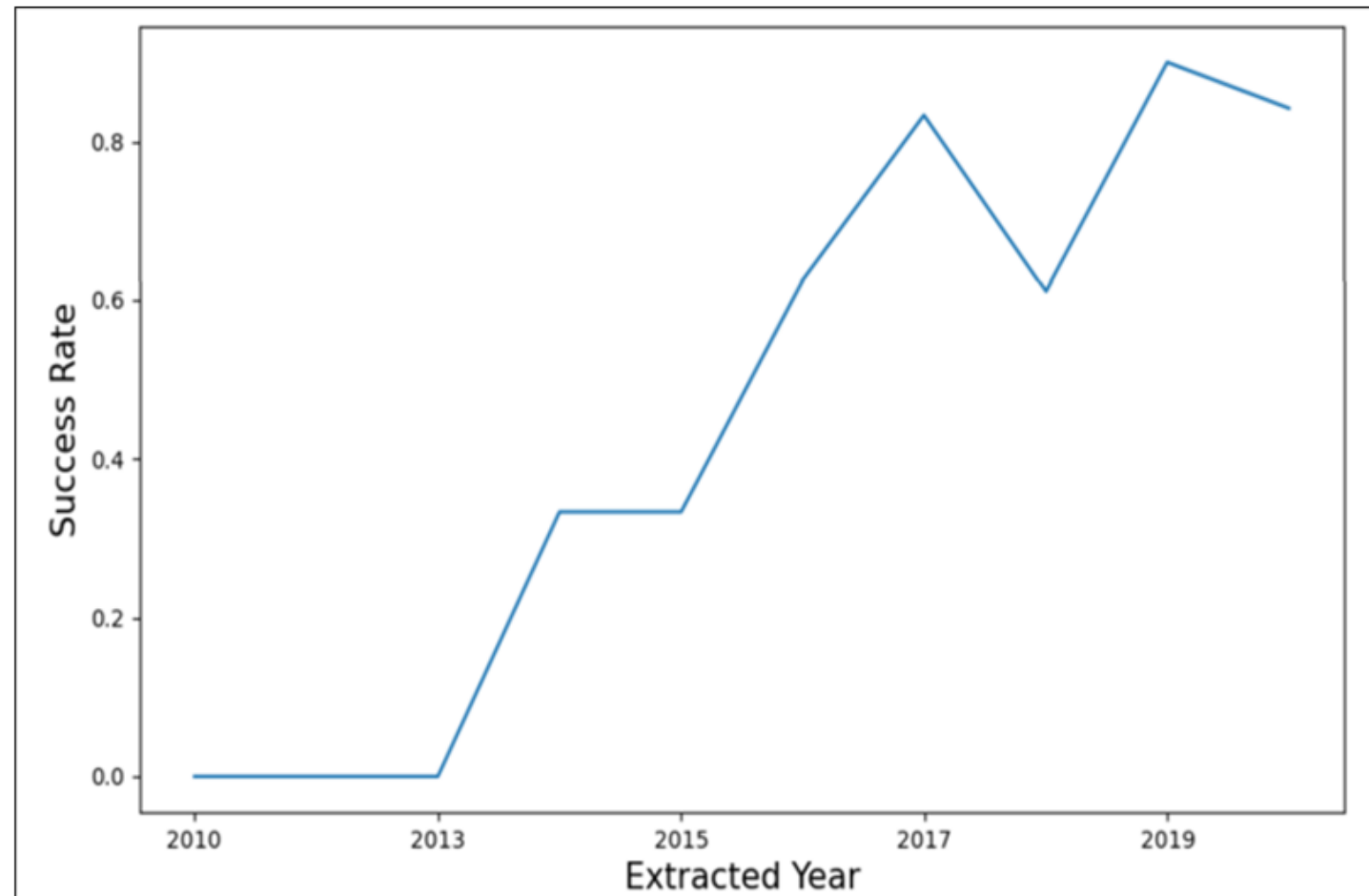


Figure 6. Line chart showing Launch Success Yearly Trend.

All Launch Site Names

Names of **unique launch sites** in space mission

- were distinctly selected from **SpaceX Table** as **CCAFS LC-40**, **VAFB SLC-4E**, **KSC LC-39A**, & **CCAFS SLC-40** in the query result on the Right.

```
[8]: %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'

```
[12]: %sql SELECT * FROM SPACEXTBL WHERE Launch_Site LIKE '%CCA%' LIMIT 5;
```

* sqlite:///my_data1.db
Done.

[12]:

| Date | Time (UTC) | Booster_Version | Launch_Site | Payload | PAYLOAD_MASS_KG_ | Orbit | Customer | Mission_Outcome | Landing_Outcome |
|------------|------------|-----------------|-------------|---|------------------|-----------|-----------------|-----------------|---------------------|
| 2010-04-06 | 18:45:00 | F9 v1.0 B0003 | CCAFS LC-40 | Dragon Spacecraft Qualification Unit | 0 | LEO | SpaceX | Success | Failure (parachute) |
| 2010-08-12 | 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 | 07:44:00 | F9 v1.0 B0005 | CCAFS LC-40 | Dragon demo flight C2 | 525 | LEO (ISS) | NASA (COTS) | Success | No attempt |
| 2012-08-10 | 00:35:00 | F9 v1.0 B0006 | CCAFS LC-40 | SpaceX CRS-1 | 500 | LEO (ISS) | NASA (CRS) | Success | No attempt |
| 2013-01-03 | 15:10:00 | F9 v1.0 B0007 | CCAFS LC-40 | SpaceX CRS-2 | 677 | LEO (ISS) | NASA (CRS) | Success | No attempt |

5 Records where **launch sites** begin with the string **'CCA'**

- were selected from **SpaceX Table** as shown in the query result above.

Total Payload Mass

```
[12]: %sql SELECT SUM(PAYLOAD_MASS__KG_) AS Total_Payload_Mass, Customer FROM SPACEXTBL WHERE Customer = 'NASA (CRS)';
```

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

```
Done.
```

```
[12]: Total_Payload_Mass  Customer
      -----
          45596  NASA (CRS)
```

The **total payload mass** carried by **boosters** launched by **customer NASA (CRS)**

- was selected from **SpaceX Table** and calculated as **45596** kg in the query result above.

Average Payload Mass by F9 v1.1

```
[13]: %sql SELECT AVG(Payload_Mass__KG_) AS Average_Payload_Mass FROM SPACEXTBL WHERE Booster_Version LIKE '%F9 v1.1%';
* sqlite:///my_data1.db
Done.
[13]: Average_Payload_Mass
      2534.6666666666665
```

The **average payload mass** carried by **booster version F9 v1.1**

- was selected from **SpaceX Table** & displayed as **2534.67** in the query result above.

First Successful Ground Landing Date

```
[15]: %sql SELECT MIN(Date),Landing_Outcome FROM SPACEXTBL WHERE Landing_Outcome LIKE '%ground pad%';
```

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

```
Done.
```

```
[15]: MIN(Date)    Landing_Outcome
```

```
2015-12-22  Success (ground pad)
```

The **Date** when the **1st successful** achievement of **landing outcome** took place in **ground pad**

- was selected from **SpaceX Table** as **22 August 2015** in the query result above.

Successful Drone Ship Landing with Payload between 4000 and 6000

```
[15]: %sql SELECT Booster_Version,Landing_Outcome, PAYLOAD_MASS__KG_ FROM SPACEXTBL WHERE Landing_Outcome='Success (drone ship)'AND PAYLOAD_MASS__KG_>=4000 AND PAYLOAD_MASS__KG_<6000
```

* sqlite:///my_data1.db
Done.

```
[15]:
```

| Booster_Version | Landing_Outcome | PAYLOAD_MASS__KG_ |
|-----------------|----------------------|-------------------|
| F9 FT B1022 | Success (drone ship) | 4696 |
| F9 FT B1026 | Success (drone ship) | 4600 |
| F9 FT B1021.2 | Success (drone ship) | 5300 |
| F9 FT B1031.2 | Success (drone ship) | 5200 |

Names of the boosters which have success in drone ship & have payload mass greater than 4000 but less than 6000 kg

- were selected from SpaceX Table & listed down in the sql query result above.

Total Number of Successful and Failure Mission Outcomes

```
[16]: %sql SELECT COUNT(Mission_Outcome), Mission_Outcome FROM SPACEXTBL WHERE Mission_Outcome = 'Success' OR Mission_Outcome LIKE '%Failure%' Group By Mission_Outcome;
* sqlite:///my_data1.db
Done.
```

```
[16]:
```

| COUNT(Mission_Outcome) | Mission_Outcome |
|------------------------|---------------------|
| 1 | Failure (in flight) |
| 98 | Success |

The total no. of **successful** & **failure mission outcomes**

- was selected from **SpaceX Table** &
- listed down as **1 Failure** & **98 Successful mission outcomes** in the sql query result above.

Boosters Carried Maximum Payload

The **names** of the **booster_versions** which have carried the **maximum payload mass**

- were selected from **SpaceX Table** & listed down in the sql query result below.

```
[17]: %sql SELECT Booster_Version, PAYLOAD_MASS_KG_ FROM SPACEXTBL WHERE PAYLOAD_MASS_KG_ = (select MAX(PAYLOAD_MASS_KG_) FROM SPACEXTBL);
```

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

```
Done.
```

```
[17]: Booster_Version  PAYLOAD_MASS_KG_
```

```
F9 B5 B1048.4      15600
```

```
F9 B5 B1049.4      15600
```

```
F9 B5 B1051.3      15600
```

```
F9 B5 B1056.4      15600
```

```
F9 B5 B1048.5      15600
```

```
F9 B5 B1051.4      15600
```

```
F9 B5 B1049.5      15600
```

```
F9 B5 B1060.2      15600
```

```
F9 B5 B1058.3      15600
```

```
F9 B5 B1051.6      15600
```

```
F9 B5 B1060.3      15600
```

```
F9 B5 B1049.7      15600
```

2015 Launch Records

```
[18]: %%sql
SELECT SUBSTR(Date, 1, 4) || '/' || SUBSTR(Date, 9, 2) || '/' || SUBSTR(Date, 6, 2) AS New_Date_Format,
Landing_Outcome, Booster_Version, Launch_Site FROM SPACEXTBL WHERE Landing_Outcome = 'Failure (drone ship)' AND SUBSTR(Date, 1, 4) = '2015';

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

[18]:

| New_Date_Format | Landing_Outcome | Booster_Version | Launch_Site |
|-----------------|----------------------|-----------------|-------------|
| 2015/01/10 | Failure (drone ship) | F9 v1.1 B1012 | CCAFS LC-40 |
| 2015/14/04 | Failure (drone ship) | F9 v1.1 B1015 | CCAFS LC-40 |

Launch records which will display the month names, **failure landing_outcomes** in **drone ship**, **booster versions**, & **launch_site** for the months in year **2015**

- were selected from **SpaceX Table** and listed down in the sql query result above.

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

```
[20]: %%sql
SELECT Landing_Outcome, COUNT(Landing_Outcome) AS COUNT, Rank() OVER (ORDER By COUNT(Landing_Outcome) DESC) FROM SPACEXTBL
WHERE Landing_Outcome = 'Failure (drone ship)' OR Landing_Outcome = 'Success (ground pad)' AND (Date between '2010-06-04' and '2017-03-20')
Group By Landing_Outcome
ORDER By COUNT DESC

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

```
[20]:
```

| Landing_Outcome | COUNT | Rank() OVER (ORDER By COUNT(Landing_Outcome) DESC) |
|----------------------|-------|--|
| Success (ground pad) | 5 | 1 |
| Failure (drone ship) | 5 | 1 |

Count of **landing outcomes** (such as **Failure (drone ship)** or **Success (ground pad)**) between the **date 2010-06-04 & 2017-03-20**

- was selected from **SpaceX Table** &
- ranked in **descending** order & counted as **5 Success (ground pad)** or **5 Failure (drone ship)** as shown in the sql query result above.

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

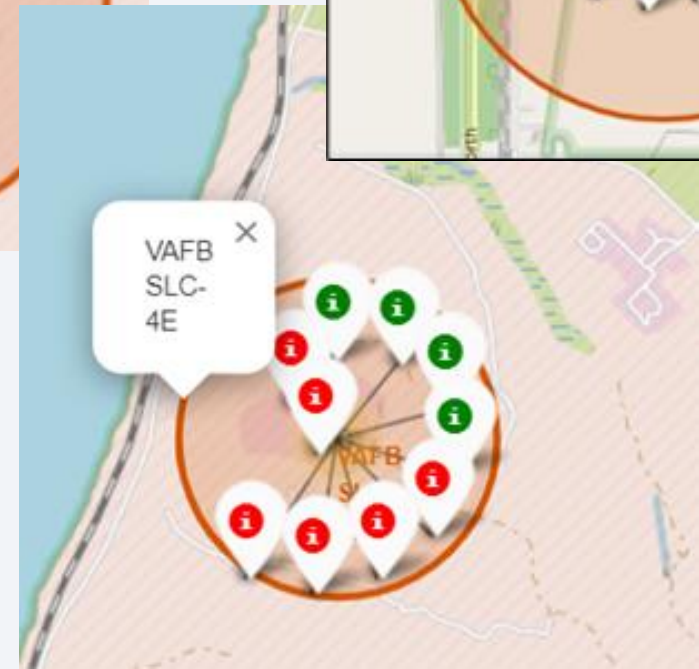
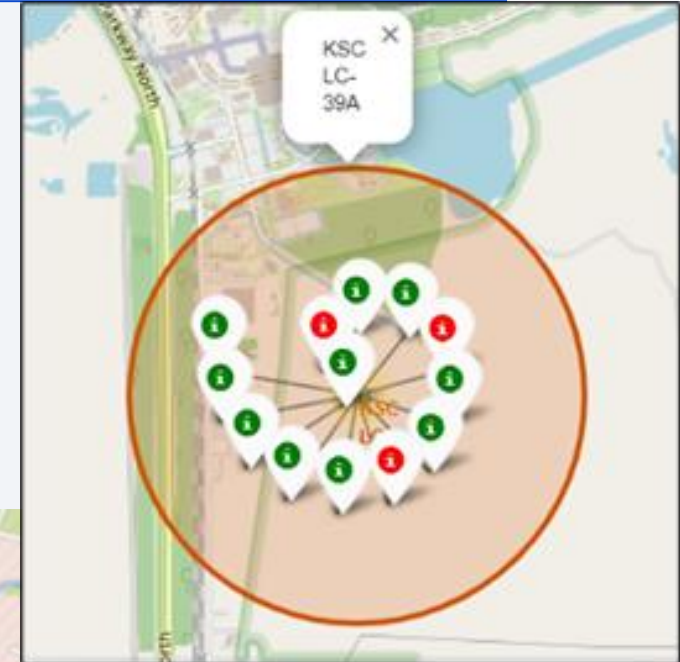
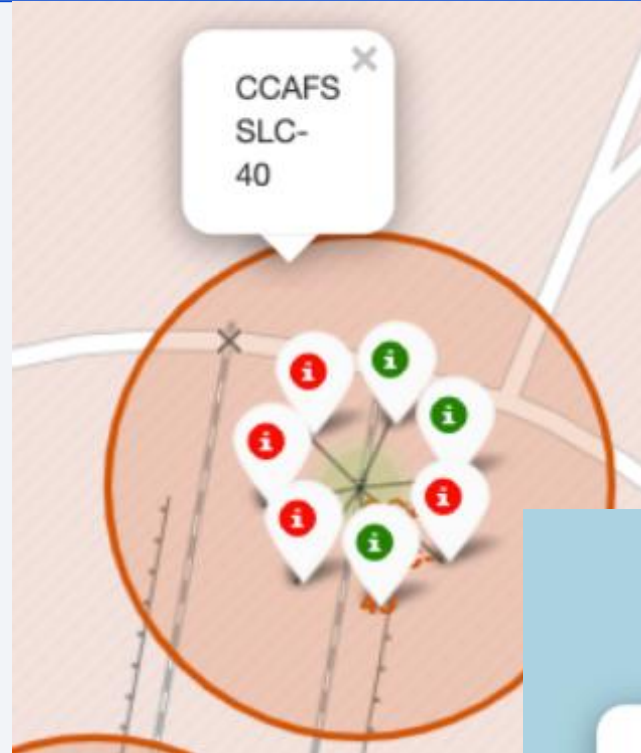
Folium Map showing all Launch Sites Marked



- All **launch sites** were created & added on the **site map**
 - ✓ using **folium.Circle** & **folium.Marker** objects as shown above.
- **Launch Sites** marked were **CCAFS LC-40**, **VAFB SLC-4E**, **KSC LC-39A**, & **CCAFS SLC-40**

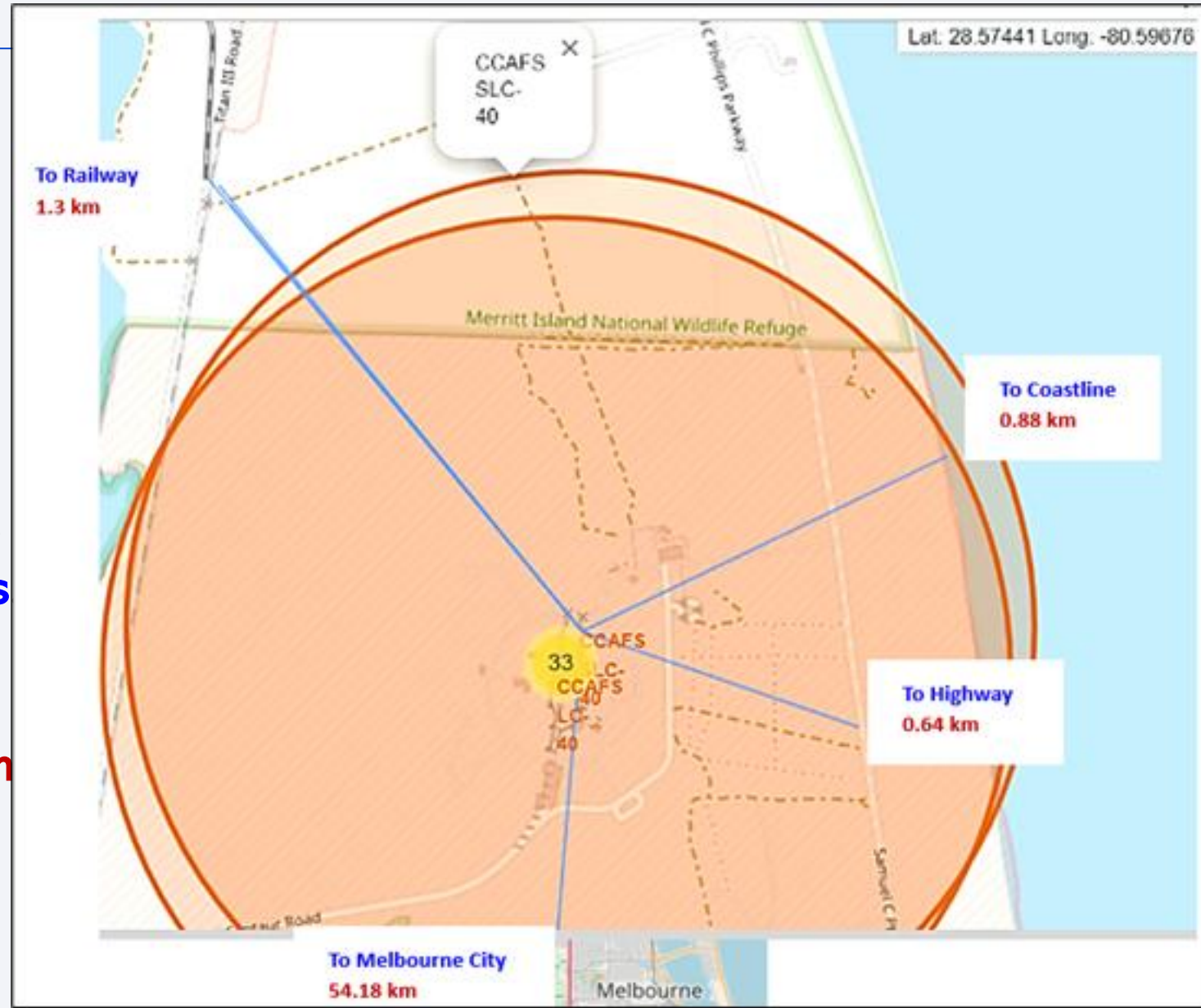
Folium Map showing marked success/failed launches for Launch Sites

- The **success/failed launches** for each **launch site** were created and marked on the **site map** using `MarkerCluster()` objects
- **Zooming** in on the **launch sites**, **Launch sites** are marked; where
 - ✓ each **green** marker represent a **launch** which was **successful** &
 - ✓ each **red** marker stands for a **failed launch** as shown in the screenshot map on the Right.
- **Launch Site KSC LC-39A** (more **green markers**)
 - ✓ shows relatively **higher success rate** than **CCAFS SLC-40** & **VAFB SLC-4E**



Folium Map showing distances b/n selected Launch Site to its proximities

- The **distances** between a **launch site CCAFS SLC-40** to its closest **coastline, highway, railway & city**
- were calculated by finding their **coordinates** on the **map**
 - ✓ first using **MousePosition** objects; &
 - ✓ then a **PolyLine** was drawn between the **CCAFS SLC-40** to its selected **proximities**
 - Launch site is in close proximity to the
 - ✓ **Highway (0.64 km) & Coastline (0.88 km)**
 - But farther away from
 - ✓ **City (Melbourne) (54.18 km).**

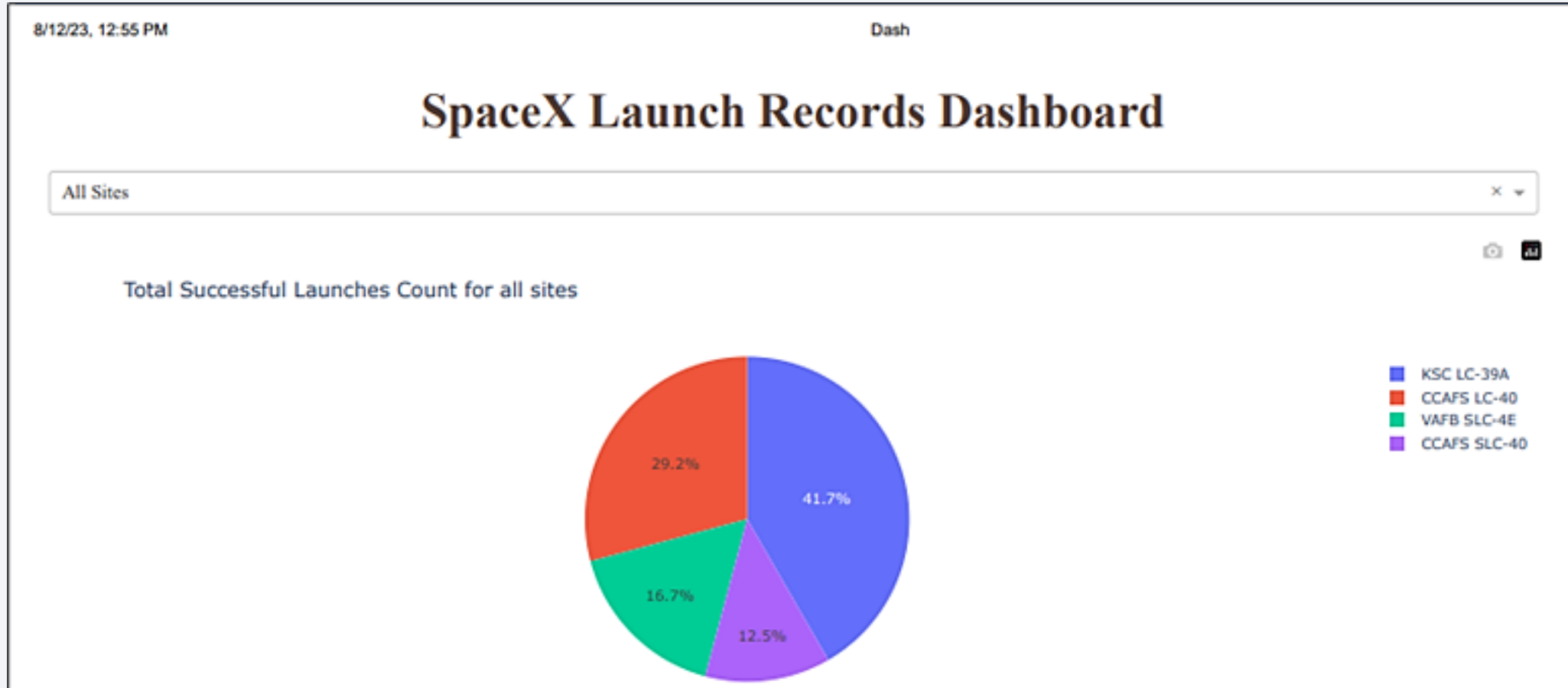




Section 4

Build a Dashboard with Plotly Dash

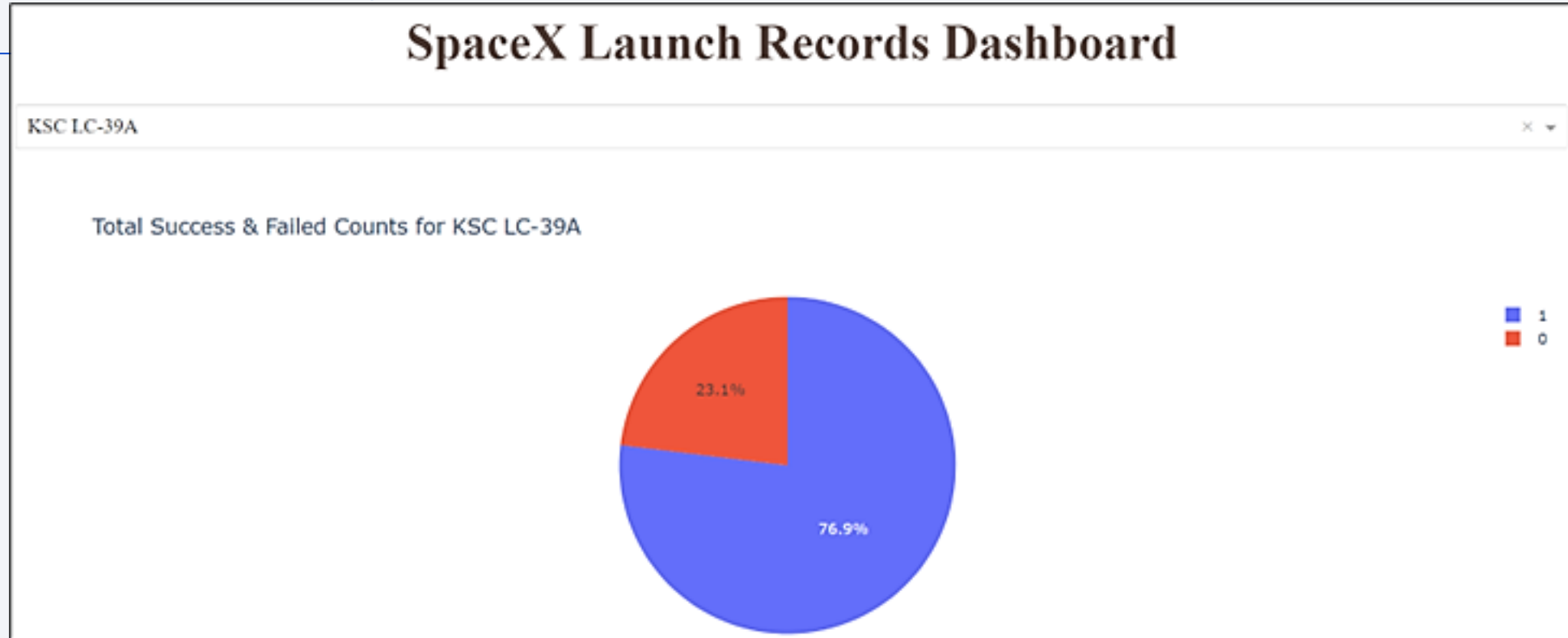
Dashboard showing Total Successful Launches Count for all Sites



The **Pie Chart** shows when **all sites** are chosen from the **Dropdown** menu.

- **KSC LC-39A (41.7%)** is the **launch site** with the **largest successful launches**.

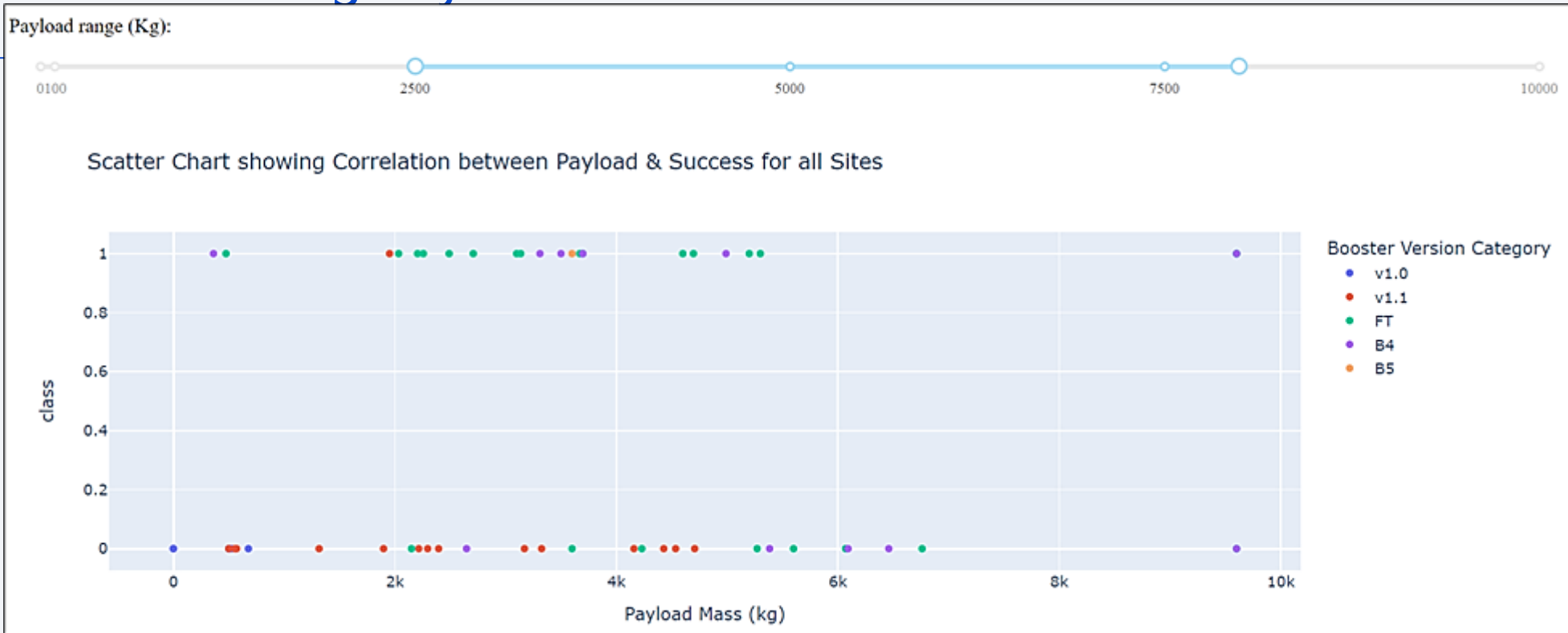
Dashboard showing Total Success & Failed Counts for KSC LC-39A



The **Pie Chart** shows when **launch site KSC LC-39A** is chosen from the **Dropdown** menu.

- **1** represents **successful launches** &
- **0** represents **failed launches**.
- **76.9%** of the **launches** done at **KSC LC-39A** are **successful launches**.
- **KSC LC-39A** is displayed as the **launch site** with **highest launch success ratio**.

Dashboard showing Payload vs. Launch Outcome Scatter Plot for all sites



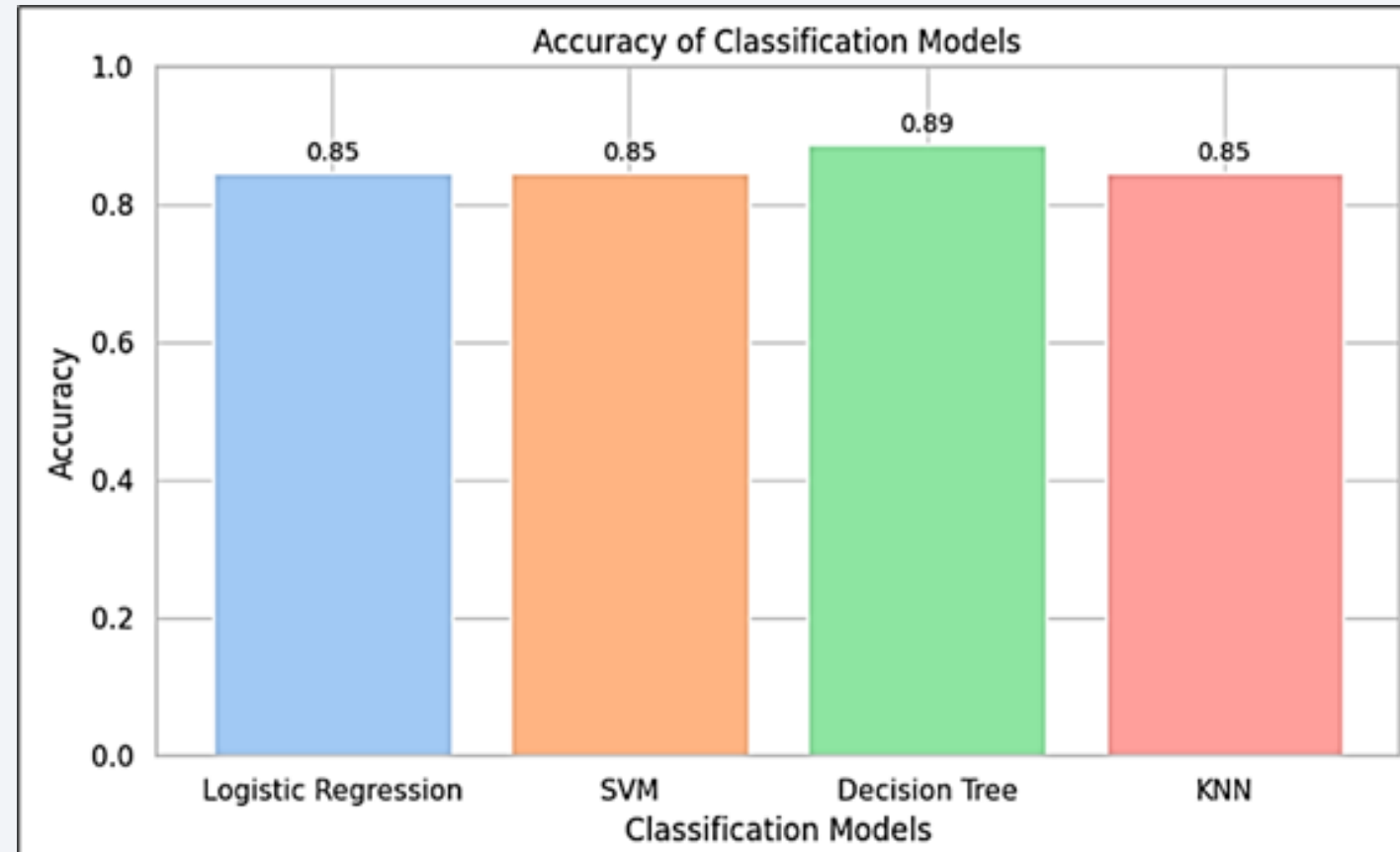
- The **Scatterplot** shows when **payload mass range** is selected b/n **2500 & 8000 kg** in the **range slider**.
- **Class 0** represents **failed launches** while **class 1** represents **successful launches**.
- The **payload range** of **20000-4000 kg** has the **highest launch success rate** whereas the **payload range** of **6000-8000 kg** has the **lowest launch success rate**.
- The **Booster version FT** tends to show the **highest launch success rate**
✓ with a greater number of **successful launches** reaching **class 1**.

Section 5

Predictive Analysis (Classification)

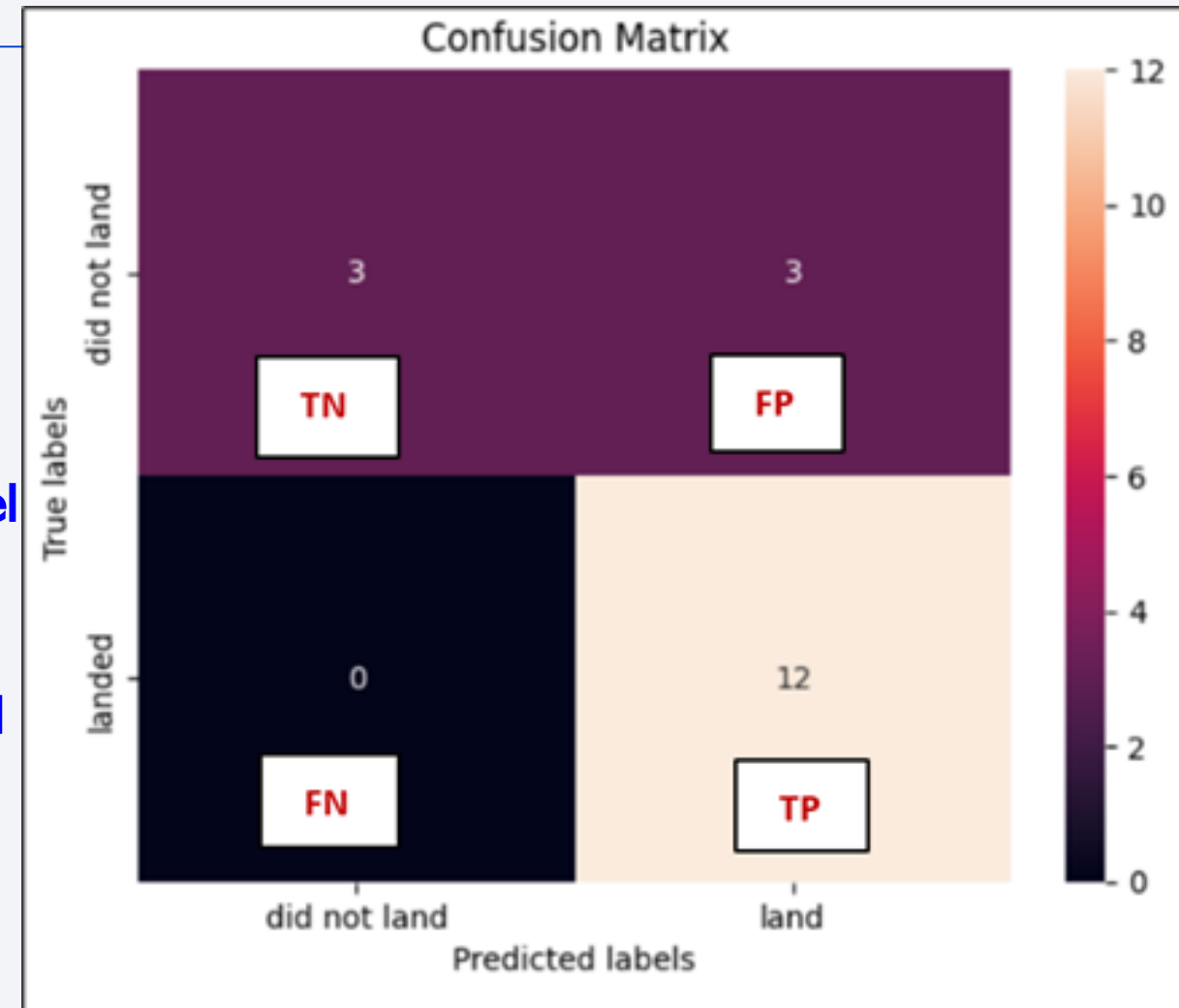
Classification Accuracy

- All the **Classification models** (**LR**, **SVM**, **Tree** and **KNN**)
- share the **same accuracy score (0.83)** on **test set**.
- **GridSearchCV.best_scores_**
✓ used to compare **classification models** as shown in the **Bar chart**
- **Decision tree** with the **tree_cv.best_score_** of **0.89** has the **highest classification accuracy**.



Confusion Matrix

- The **confusion matrix** plotted after **Tree's accuracy** is tested on the **test** data as shown on the Right.
- It shows **corrected** & **wrong predictions** of landings in comparison with the **actual** labels.
- The **test** set has only **18 rows/outcomes**
- **True positive (TP)** is the **outcome (12)** when the **model correctly predicts** that a **landing** occurred when it actually did **land**.
- **False positive (FP)** is the **outcome (3)** when the **model predicts** a **landing** when it didn't actually **land**.
- **False Negative (FN)** is the **outcome (0)** when **model incorrectly predicted** no landing when there was one.
- **True Negative (TN)** is the **outcome (3)** when **model correctly predicted** no landing.



Confusion Matrix (Continued)

The weighted **Average of F1-Score** (**0.81**)

- balances **precision** & **recall** gained from **Confusion Matrix** as shown in query result below.

```
[35]: from sklearn.metrics import f1_score  
f1_score(Y_test, yhat, average='weighted')
```

```
[35]: 0.8148148148148149
```

```
[36]: from sklearn.metrics import classification_report, confusion_matrix  
import itertools  
cnf_matrix = confusion_matrix(Y_test, yhat)  
np.set_printoptions(precision=2)  
print (classification_report(Y_test, yhat))
```

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 1.00 | 0.50 | 0.67 | 6 |
| 1 | 0.80 | 1.00 | 0.89 | 12 |
| accuracy | | | 0.83 | 18 |
| macro avg | 0.90 | 0.75 | 0.78 | 18 |
| weighted avg | 0.87 | 0.83 | 0.81 | 18 |

Conclusions

- All models (LR, SVM, Decision Tree, KNN) perform best w same accuracy score (0.83) on test set.
- GridSearchCV_best scores_ (0.89) ranks 'Decision Tree' as the best model, outperforming the other models.

Confusion Matrix plays a good job in predicting outcomes with Average Accuracy F1-score (0.81)

- correctly predicted 12 (TP) of outcomes as landed &
 - none of them (FN=0) wrongly predicted as not landed,
- ✓ showing that landing is not costly & thus SpaceX can reuse the Falcon 9 1st stage.

The following attributes/factors are attributed to the successful 1st stage landing of Falcon 9.

- ES-L1, GEO, HEO & SSO Orbits display highest success rates, 100%
- KSC LC-39A (41.7%) has the highest successful launches of all the launch sites.
- Lighter payloads have higher launch success rates than heavier payloads for all launch sites.
- Heavier payloads perform better (100%) than lighter payloads for CCAFS LC-40.

Conclusions (continued)

- The **correlation** of **attributes** (payload mass, flight No., Orbit, etc.) with another also contributed to the successful landings.
- More **successful** landing rate was observed for **Polar** (**PO**), **LEO** & **ISS** orbits with heavy payloads.

The other factor that is responsible for **successful** landing of Falcon 9 **1st** stage is the fact that

- all **launch sites** are **optimally** located near the Equator, the coastline, highways & railways.
- **Launch sites** **closer** to the equator seems to gain a rotational speed boost due to the Earth's rotation which can be useful in terms of **fuel efficiency** & **payload capacity**.
- Their **proximity** to the coastline (oceans) allows them to have a safe flight over the water during the **launch**, minimizing risk to populated areas.
- Their **accessibility** to **railways** & **highways** facilitate easy transportation of equipment & payloads.
- **Launch success rate** has shown an **increasing** yearly trend since **2013** (year when **reusable** technology was started). This predicts future successful landings of **Falcon 9 1st** stage as more promising.

Appendix

```
[34]: models = ['Logistic Regression', 'SVM', 'Decision Tree', 'KNN']
GridSearchCV.best_scores_ = [logreg_cv.best_score_, svm_cv.best_score_, tree_cv.best_score_, knn_cv.best_score_]
# Set the style using Seaborn
sns.set(style="whitegrid")

# Create a bar plot
plt.figure(figsize=(8, 5))
plt.bar(models, GridSearchCV.best_scores_, color=sns.color_palette("pastel"))

# Adding labels and title
plt.xlabel('Classification Models')
plt.ylabel('Accuracy')
plt.title('Accuracy of Classification Models')
plt.ylim(0, 1) # Set the y-axis limits between 0 and 1

# Display the accuracy values on top of the bars
for i, v in enumerate(GridSearchCV.best_scores_):
    plt.text(i, v + 0.01, f'{v:.2f}', ha='center', va='bottom', fontsize=10)

# Show the plot
plt.tight_layout()
plt.show()
```

Thank you!

