



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 using an API
  - Data collection using web scraping
  - Data Wrangling
  - EDA using SQL
  - EDA using Python visualizations
  - Interactive maps
  - Machine learning to test Stage 1 reusability
- Summary of all results
  - Visualizations
  - Predictive analysis

# Introduction

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- Background
  - Space Y would like to compete with SpaceX, however understanding the cost and price of each launch is critical to the success of market entry
- Problem Statement
  - Will SpaceX reuse the first stage of their Falcon 9 rocket



Section 1

# Methodology

# Methodology

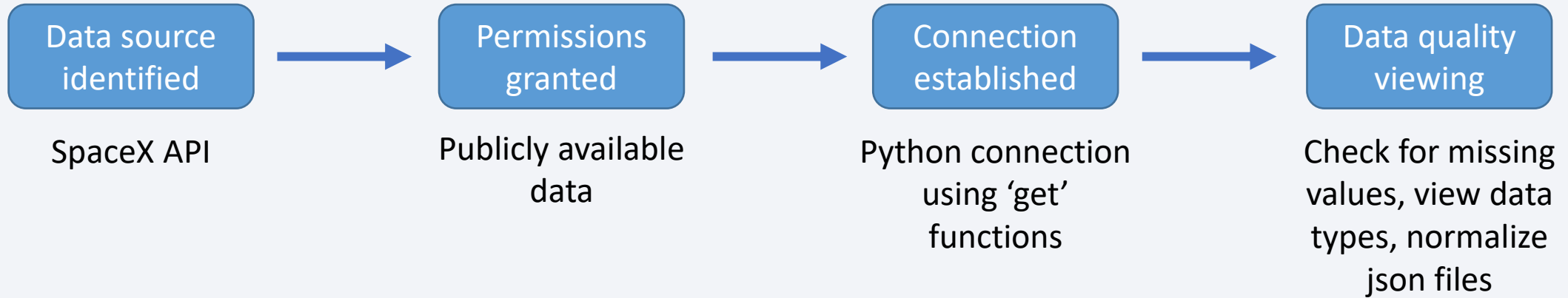
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## Executive Summary

- Data collection methodology:
  - Data was collected using publicly available SpaceX launch data. The data was sourced using an API through Python.
- Perform data wrangling
  - Using Python, separate data tables were parsed together, and successful landings were categorized alongside unsuccessful landings.
- 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

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# Data Collection – SpaceX API

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- SpaceX REST calls
  - def getBoosterVersion(data): for x in data['rocket']:
  - def getLaunchSite(data): for x in data['launchpad']
  - def getPayloadData(data): for load in data['payloads']:
  - def getCoreData(data): for core in data['cores']:
- Add the GitHub URL of the completed SpaceX API calls notebook (**must include completed code cell and outcome cell**), as an external reference and peer-review purpose

```
Def getXdata(data):  
For x in data['column']:  
  
spacex_url=https://api.spacexdata.com/v4/launches/past  
response = requests.get(spacex_url)  
print(response.content)  
  
data = pd.json_normalize(response.json())
```



# Data Collection - Scraping

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- Request the Falcon9 Launch Wiki page from its URL
- Add the GitHub URL of the completed web scraping notebook, as an external reference and peer-review purpose

Import required packages

'def' and 'get' functions to pull specific data from HTML

Define URL

Create a 'Beautiful Soup'

Convert to a dataframe

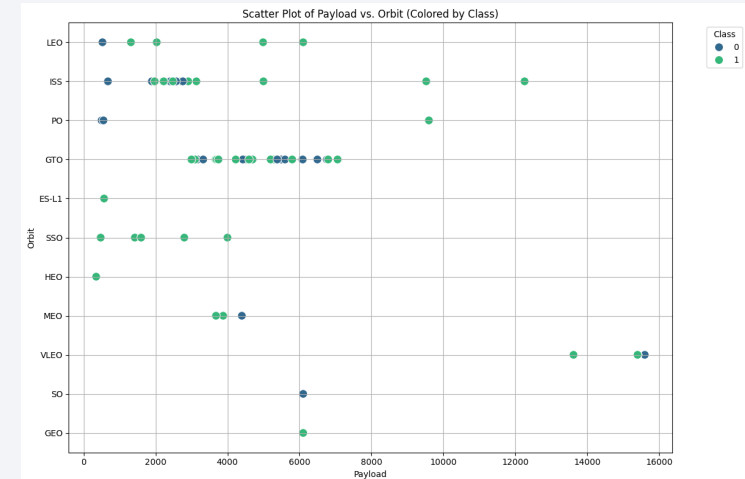
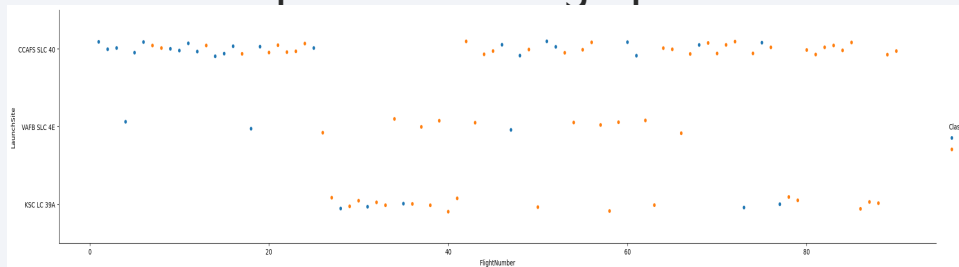
# Data Wrangling

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- Data was processed and cleaned of inconsistencies
  - Exploratory visualizations were created for better understanding of the data
- Connect to data > Missing data viewed > success class created > data normalized > data frame created
- Add the GitHub URL of your completed data wrangling related notebooks, as an external reference and peer-review purpose

# EDA with Data Visualization

- Conducted an in-depth data analysis process using Python libraries for visualization such as Pandas and Seaborn
  - Scatter plots and bar graphs were utilized for EDA



- Add the GitHub URL of your completed EDA with data visualization notebook, as an external reference and peer-review purpose

# EDA with SQL

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- `%load_ext sql`
- `import csv, sqlite3 con = sqlite3.connect("my_data1.db") cur = con.cursor()`
- `%sql sqlite:///my_data1.db`
- `%sql create table SPACEXTABLE as select * from SPACEXTBL where Date is not null`
- `SELECT DISTINCT launch_site FROM SPACEXTABLE; WHERE launch_site LIKE 'CCA%' LIMIT 5;`
- Add the GitHub URL of your completed EDA with SQL notebook, as an external reference and peer-review purpose

# Build an Interactive Map with Folium

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- Added circles around the launch sites and distinctive features to relate whether a successful launch occurred there
- This aided in visualizing which sites were optimal for a successful flight
- Add the GitHub URL of your completed interactive map with Folium map, as an external reference and peer-review purpose

# Build a Dashboard with Plotly Dash

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- Created an interactive dashboard which visualized launch site data by developing a callback function which was also used to create a pie chart
- This was done to better understand the data and the underlying problem statement
- Add the GitHub URL of your completed Plotly Dash lab, as an external reference and peer-review purpose



# Predictive Analysis (Classification)

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- Summarize how you built, evaluated, improved, and found the best performing classification model
- You need present your model development process using key phrases and flowchart
- Add the GitHub URL of your completed predictive analysis lab, as an external reference and peer-review purpose

# Results

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



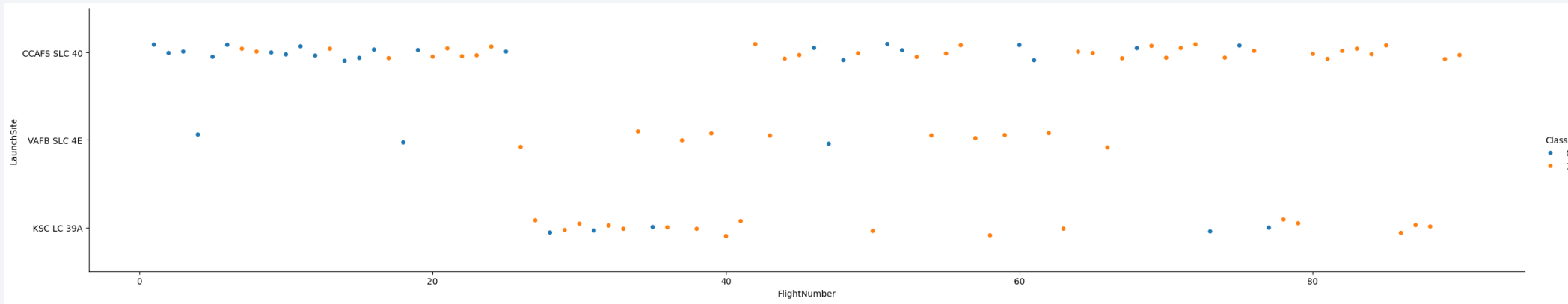
The background of the slide is an abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks in shades of red and cyan. A faint, light blue grid pattern is also visible, particularly in the lower half of the image. The overall effect is dynamic and technological.

Section 2

# Insights drawn from EDA

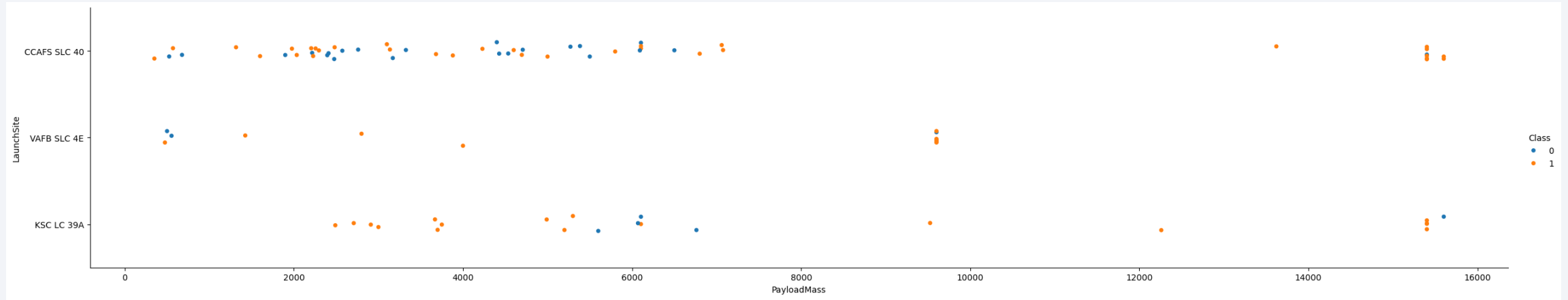


# Flight Number vs. Launch Site



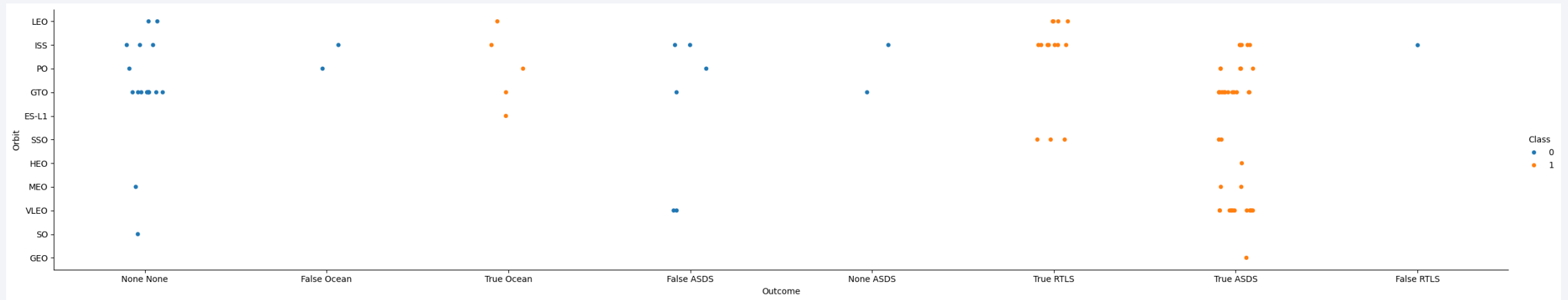
- This scatterplot shows the relationship between the place of launch and the flight number
- SLC-40 is heavily utilized as a launch site

# Payload vs. Launch Site



- This scatterplot shows the relationship between the payload size and the place of launch
- VAFB is seldom used and not at all at higher payloads
- Most payloads below 7000 happen at SLC-40

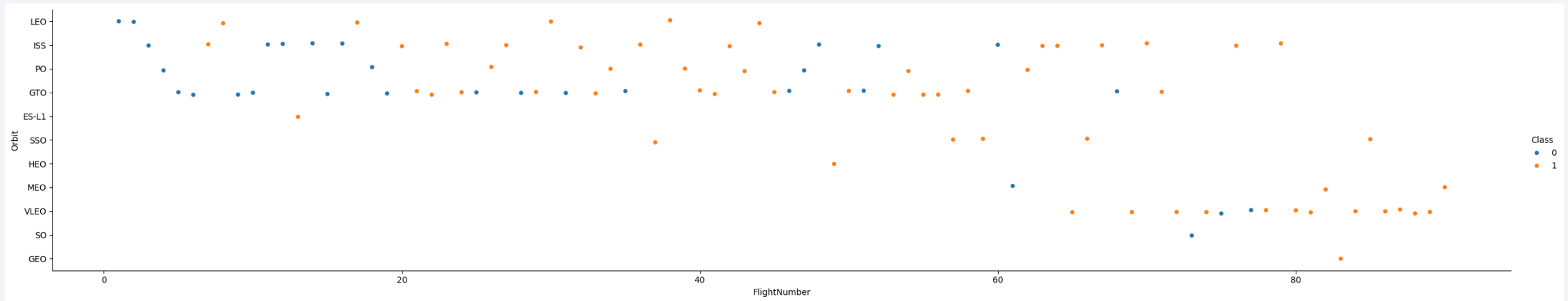
# Success Rate vs. Orbit Type



- This scatterplot shows the relationship between the success (0=unsuccessful) rate and the type of orbit the flight took.

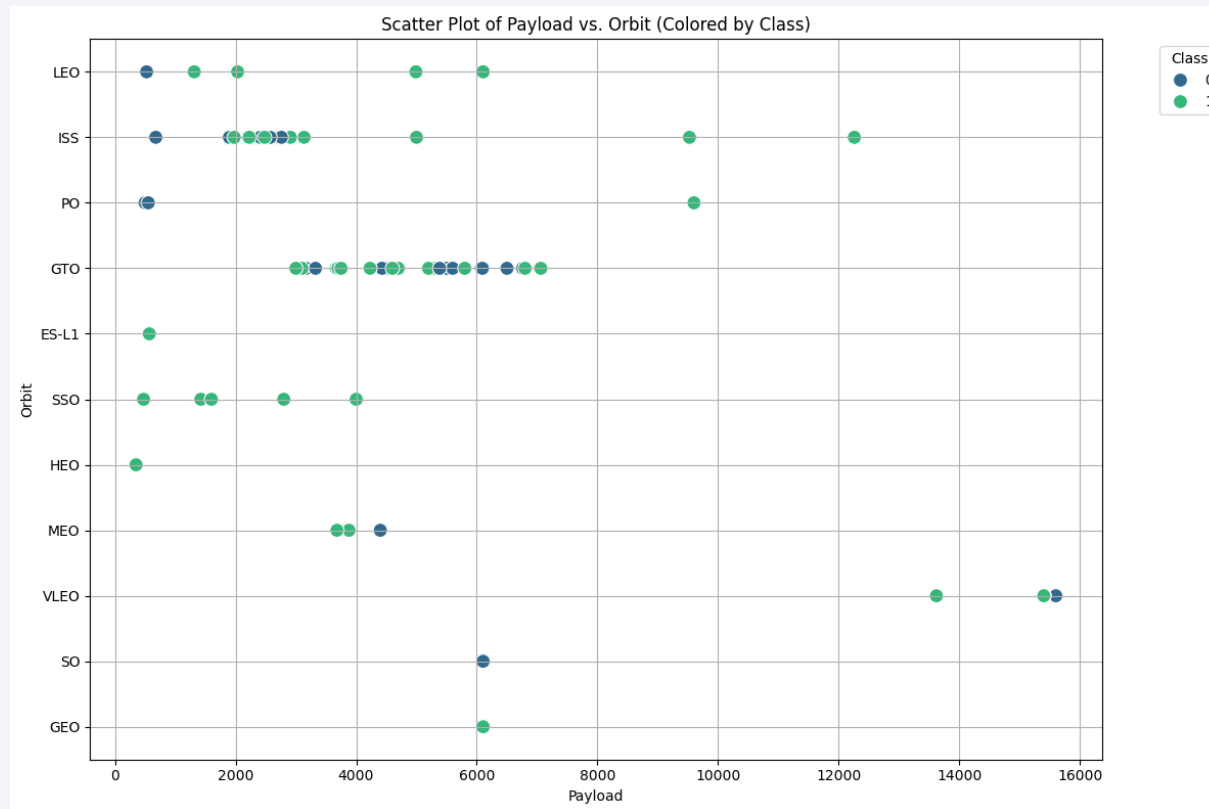


# Flight Number vs. Orbit Type



- This scatterplot shows the relationship between the flight number and type of orbit.
- In earlier flights only certain types of orbits were attempted with a mixed bag for whether the flight was successful

# Payload vs. Orbit Type



- This scatterplot shows the relationship between payload and type of orbit.
- The VLEO orbit is used with heavy payloads and SSO has all successful missions but on the lighter side of the payload weight

# All Launch Site Names

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- %  
%sql
- SELECT DISTINCT launch\_site
- FROM SPACEXTABLE;

Launch\_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

# Launch Site Names Begin with 'CCA'

- `%%sql`
- `SELECT *`
- `FROM SPACEXTABLE`
- `WHERE launch_site LIKE 'CCA%'`
- `LIMIT 5;`

Date	Time (UTC)	Booster_ Version	Launch_ Site	Payload	PAYLOA D_ MASS _KG_	Orbit	Custome r	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

# Total Payload Mass

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- %%%sql
- SELECT SUM(payload\_mass\_\_kg\_)
- FROM SPACEXTABLE
- WHERE customer = 'NASA (CRS)';

SUM(payload\_mass\_\_kg\_)

45596

# Average Payload Mass by F9 v1.1

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- %%%sql
- SELECT AVG(payload\_mass\_\_kg\_)
- FROM SPACEXTABLE
- WHERE Booster\_Version = 'F9 v1.1';

AVG(payload\_mass\_\_kg\_)

2928.4



# First Successful Ground Landing Date

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- %%sql
- SELECT MIN(Date)
- FROM SPACEXTABLE
- WHERE Mission\_Outcome = 'Success';

MIN(Date)

2010-06-04

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

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- %%%sql
- SELECT DISTINCT Booster\_Version
- FROM SPACEXTABLE
- WHERE Landing\_\_Outcome = 'Success (drone ship)'
- AND Payload\_Mass\_\_kg\_ > 4000
- AND Payload\_Mass\_\_kg\_ < 6000;

# Total Number of Successful and Failure Mission Outcomes

---

- %%sql
- SELECT Mission\_Outcome, COUNT(\*) as Total\_Count
- FROM SPACEXTABLE
- GROUP BY Mission\_Outcome;

Mission_Outcome	Total_Count
Failure (in flight)	1
Success	98
Success	1
Success (payload status unclear)	1

# Boosters Carried Maximum Payload

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- %%sql
- SELECT DISTINCT Booster\_Version
- FROM SPACEXTABLE
- WHERE Payload\_Mass\_\_kg\_ = (
  - SELECT MAX(Payload\_Mass\_\_kg\_)
  - FROM SPACEXTABLE

# 2015 Launch Records

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- %%sql
- SELECT
- substr(Date, 6, 2) AS Month,
- Landing\_\_Outcome,
- Booster\_Version,
- Launch\_Site
- FROM SPACEXTABLE
- WHERE substr(Date, 0, 5) = '2015'
- AND Landing\_\_Outcome = 'Failure (drone ship)'
- ORDER BY Month;

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

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- %%sql
- SELECT
- Landing\_\_Outcome,
- COUNT(\*) as Count\_of\_Outcomes
- FROM SPACEXTABLE
- WHERE Date BETWEEN '2010-06-04' AND '2017-03-20'
- GROUP BY Landing\_\_Outcome
- ORDER BY Count\_of\_Outcomes DESC;

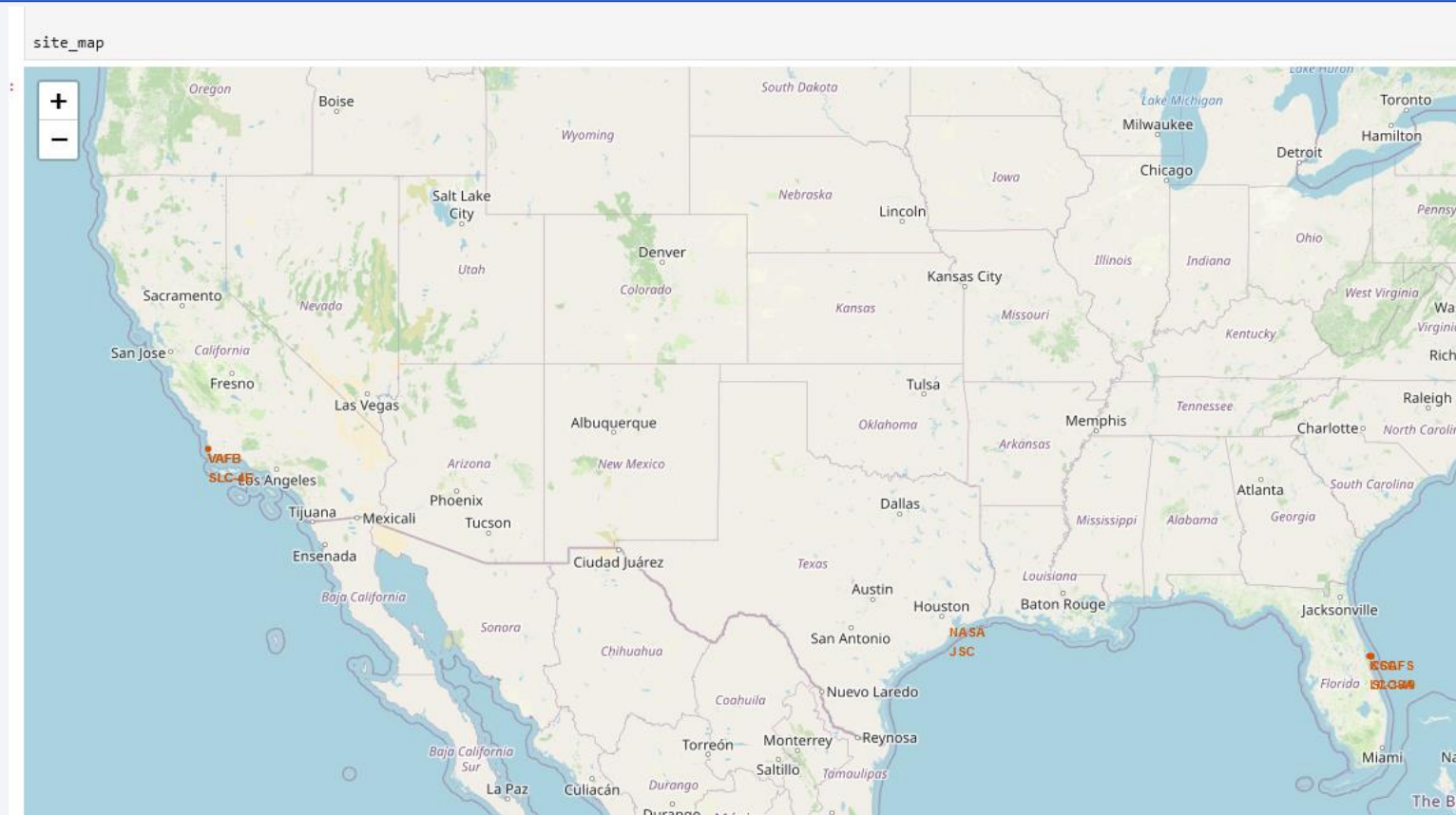


A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The background is a deep blue gradient.

Section 3

# Launch Sites Proximities Analysis

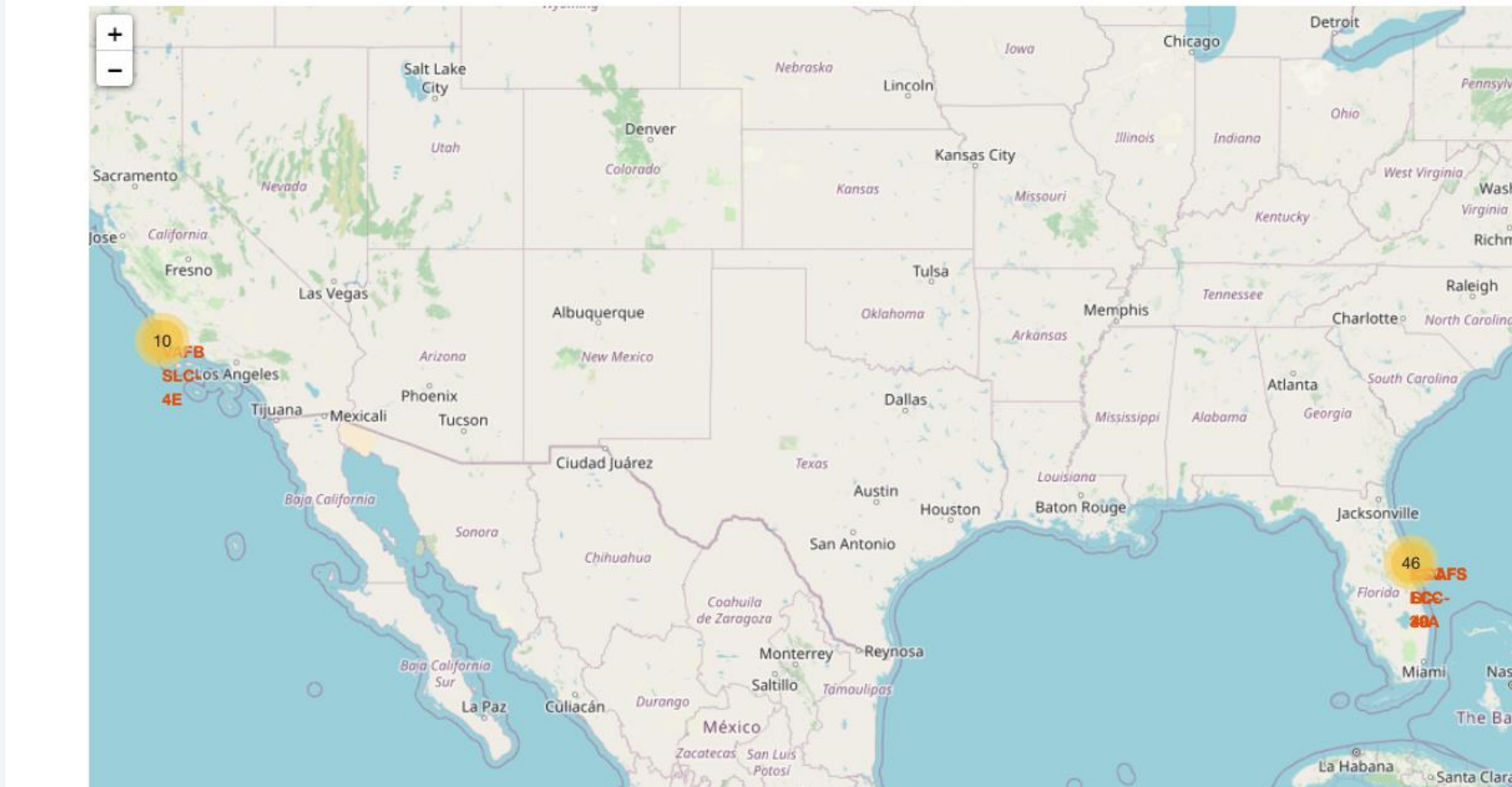
# Folium Map



- Launch sites occur in 2 main places, Southern California and The Atlantic side of Florida

# Folium Map Successful Missions

our updated map may look like the following screenshots:



- The majority of successful launches occurred in the Florida locations

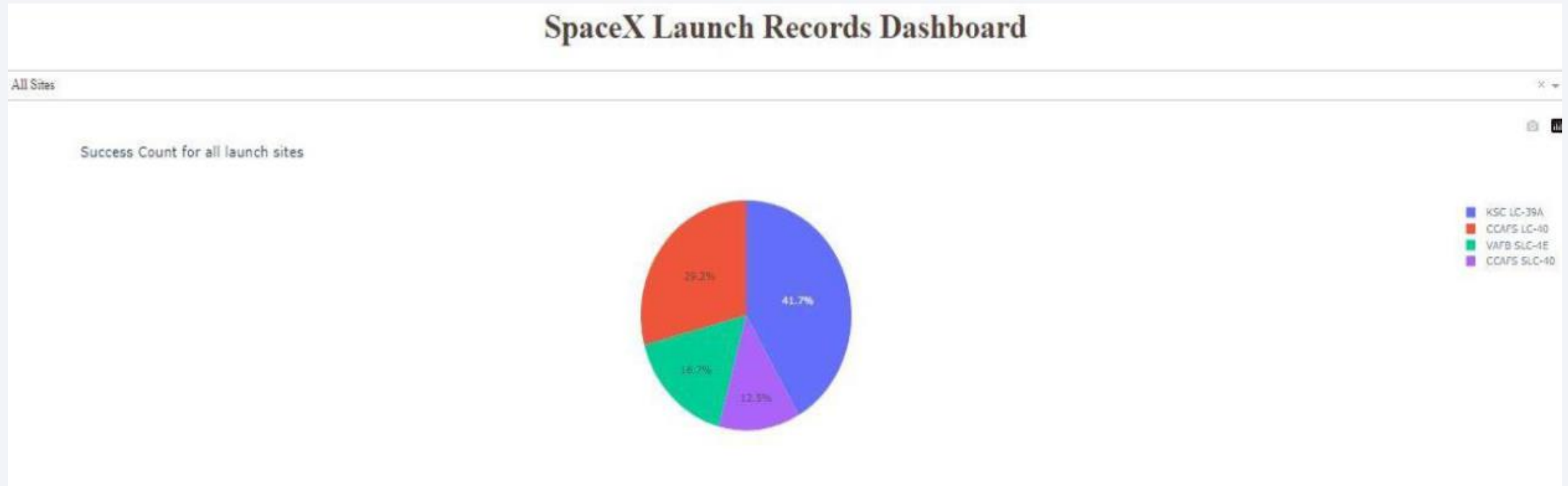




Section 4

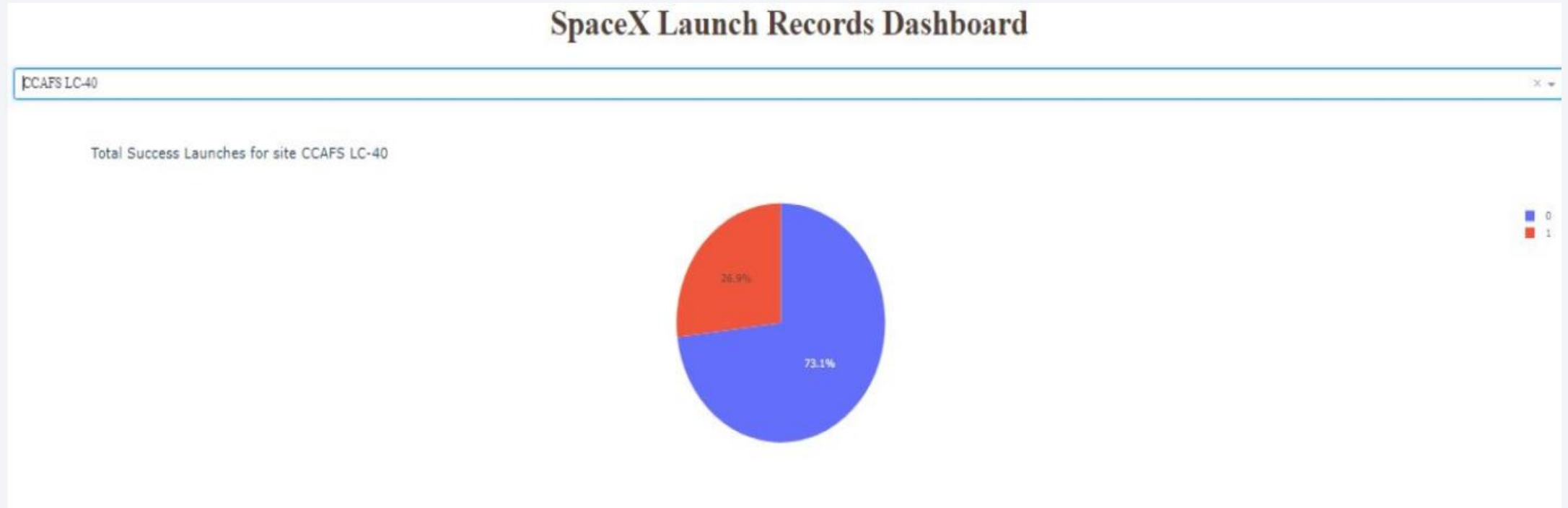
# Build a Dashboard with Plotly Dash

# Dashboard Screenshot 1



- Site KSC LC-39A has the highest launch success rate, 42%,
- CCAFS LC-40 has 29%
- VAFB SLC-4E has 17%
- CCAFS SLC-40 has 13%

# Dashboard Screenshot 2



- Site CCAFS LC-40 had a success ratio of 73%

## <Dashboard Screenshot 3>

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- Replace <Dashboard screenshot 3> title with an appropriate title
- Show screenshots of Payload vs. Launch Outcome scatter plot for all sites, with different payload selected in the range slider
- Explain the important elements and findings on the screenshot, such as which payload range or booster version have the largest success rate, etc.

Section 5

# Predictive Analysis (Classification)



# Classification Accuracy

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Best Score for Decision Tree Classifier:

0.8732142857142857 Best Score for Support Vector

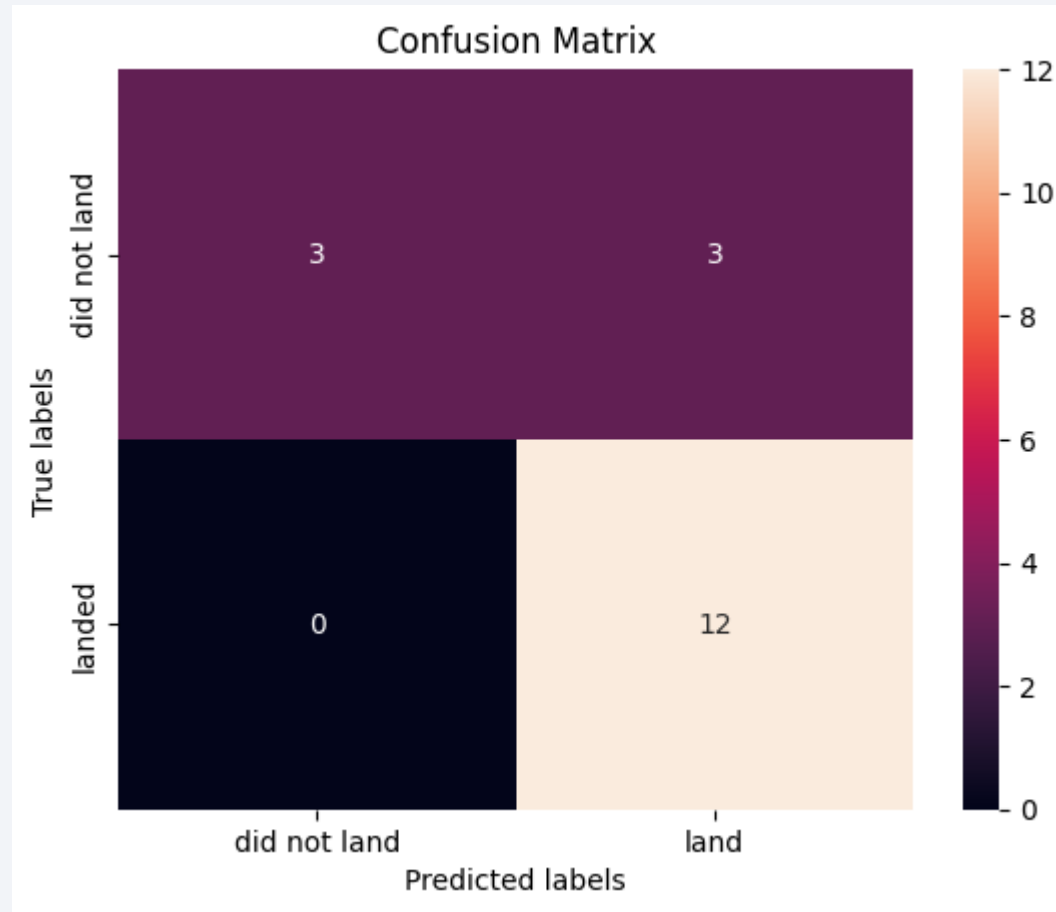
Machine: 0.8482142857142856 Best Score for KNN:

0.8482142857142858 Best Score for Log Reg:

0.8464285714285713

# Confusion Matrix

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# Conclusions

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- The Decision Tree produced the highest accuracy in predicting whether SpaceX will reuse the 1<sup>st</sup> stage of the rocket.
- Success rate overall increased from 2013 to 2020
- Success was mixed between launch sites and orbits, results were inconclusive here

Thank you!

