



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

# Executive Summary

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## Data was:

- Extracted from SpaceX API or Wikipedia
- Wrangled into a suitable format
- Explored using:
  - Graphs
  - SQL queries
  - Interactive dashboards
- Split into training and testing data
- Created ML models using training data
- Evaluated models using test data

## Results:

- All models performed equally well with 83% accuracy
- Models have false positives but not false negatives
- Limited size of dataset makes testing accurately difficult, and further data would give more confidence to results

# Introduction

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- SpaceX offers rocket launches at a significantly lower cost than competitors
- They are able to do this by re-using part of the rocket
- They can only reuse the rocket part if this part lands successfully
- If we can predict if a rocket part will land successfully then this can be used by competitors to identify cost savings



Section 1

# Methodology

# Methodology

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## Executive Summary

- Data collection methodology:
  - Data collected either via API or web scraping Wikipedia
- Perform data wrangling
  - Outcomes categorized as successful or 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
  - Built, tuned and evaluated classification models

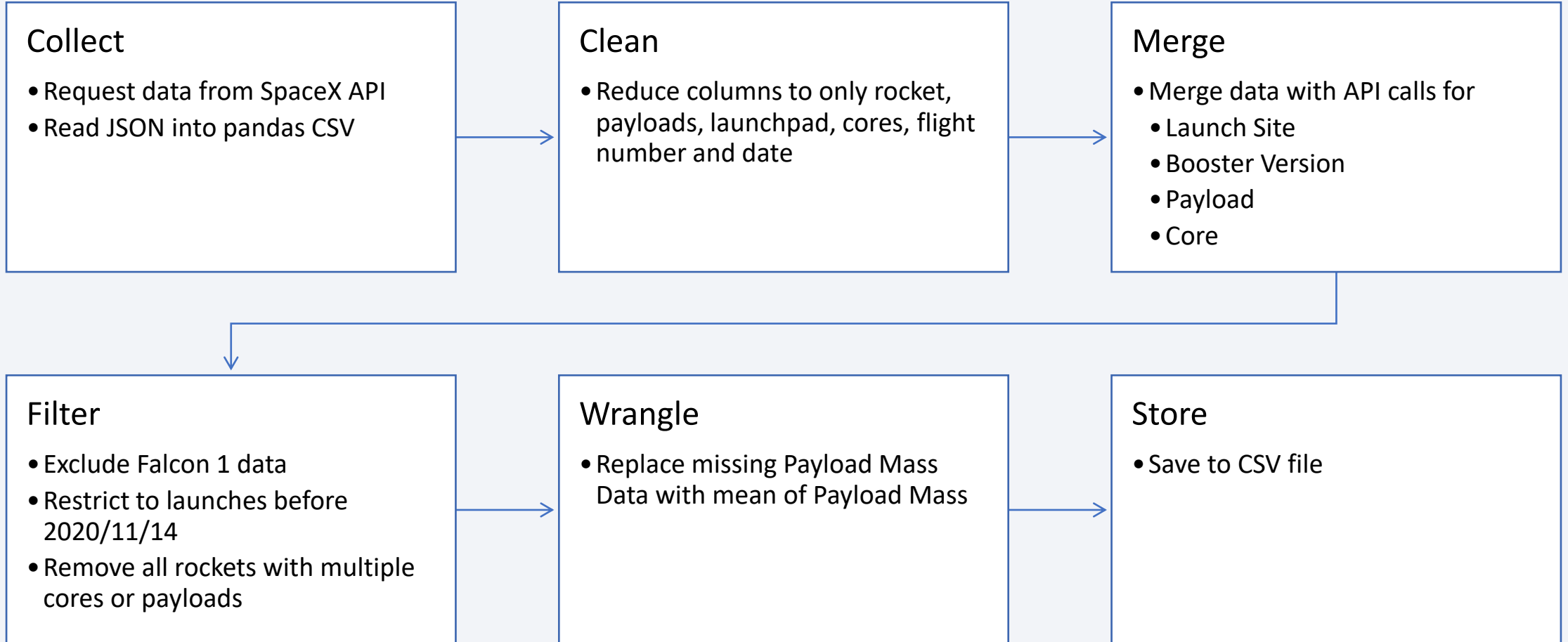
# Data Collection

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Data was collected via two different methods:

- Scraping the SpaceX Wikipedia website using BeautifulSoup
- Requesting SpaceX data directly from the SpaceX API and merging the multiple requests

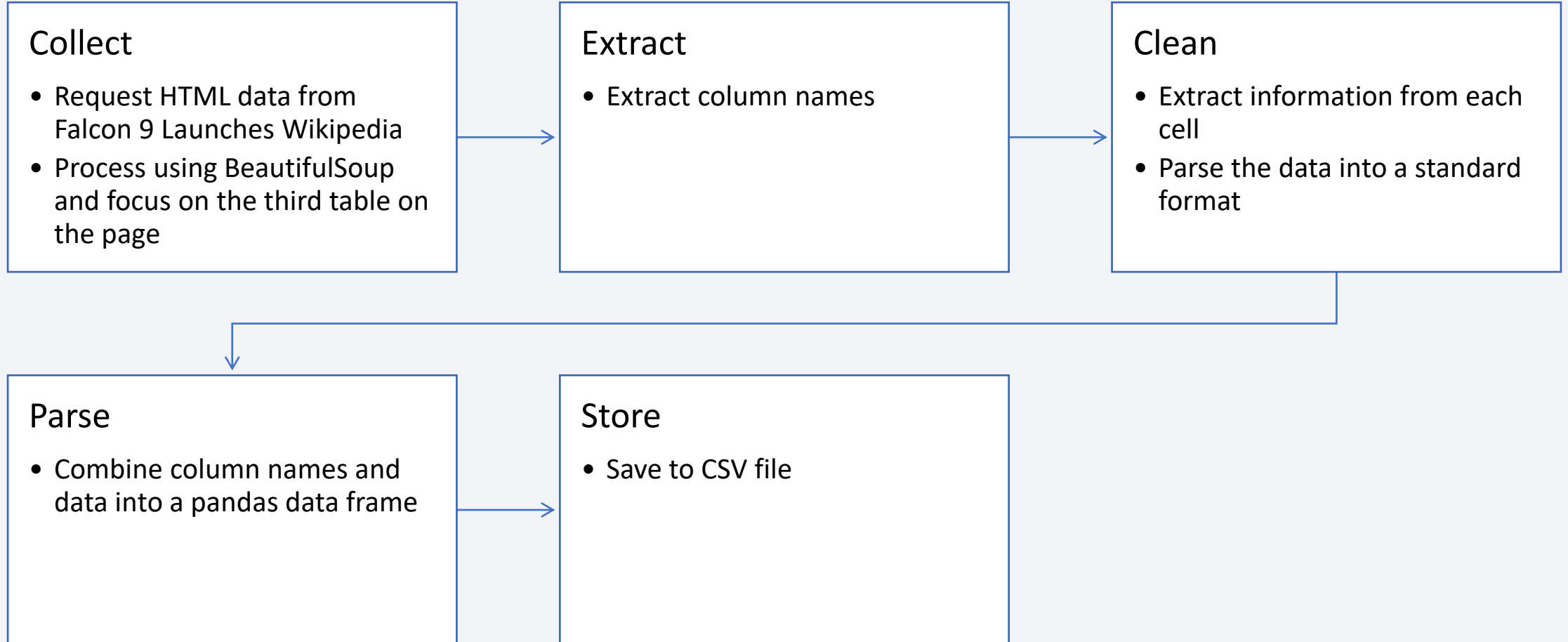
# Data Collection – SpaceX API





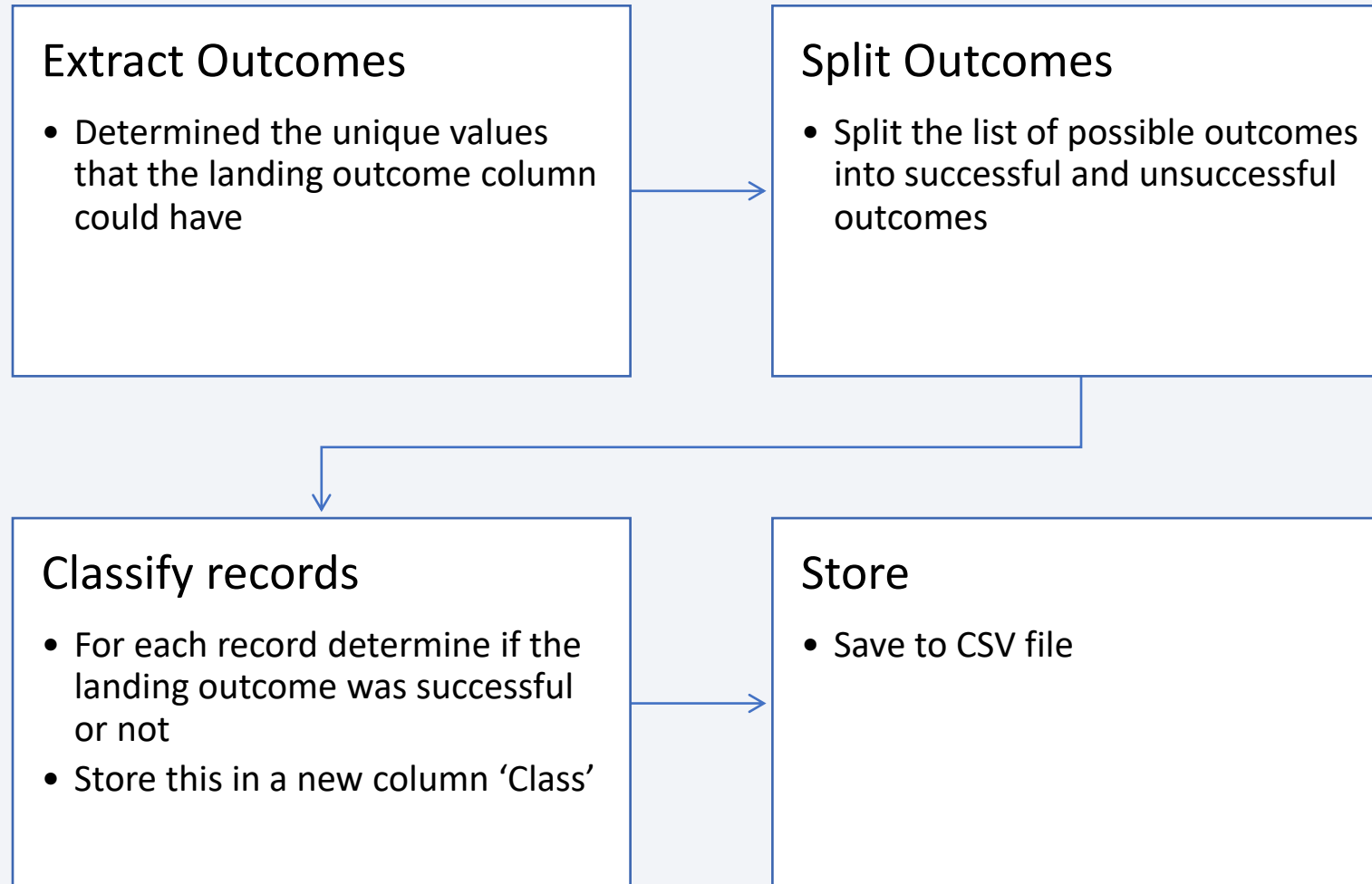
# Data Collection - Scraping

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# Data Wrangling

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# EDA with Data Visualization

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Charts were plotted to identify trends and correlations between available features.

Charts plotted:

- Flight number vs launch site
- Payload vs launch site
- Orbit type vs success rate
- Flight number vs orbit type
- Payload vs orbit type
- Success rate over time

# EDA with SQL

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- Names of unique launch sites
- Launch sites which begin with CCA
- Mass of payload launched by NASA
- Average payload carried by F9 v1.1 booster
- First successful ground pad landing date
- Successful boosters used in drone ship landings with payload mass between 4000 and 6000 kg
- Count of launches aggregated by outcome
- Booster versions which have carried maximum payload mass
- Failed drone ship landings in 2015 with launch site and booster version details
- Landing outcomes between 2010-06-04 and 2017-03-20, ranked in descending order

# Build an Interactive Map with Folium

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- Circle added to each launch site labelled with the site name
- Cluster of markers added to each site, one per launch, colour coded according to if the launch was successful or not
- Added a line between a launch site and nearby points of interest

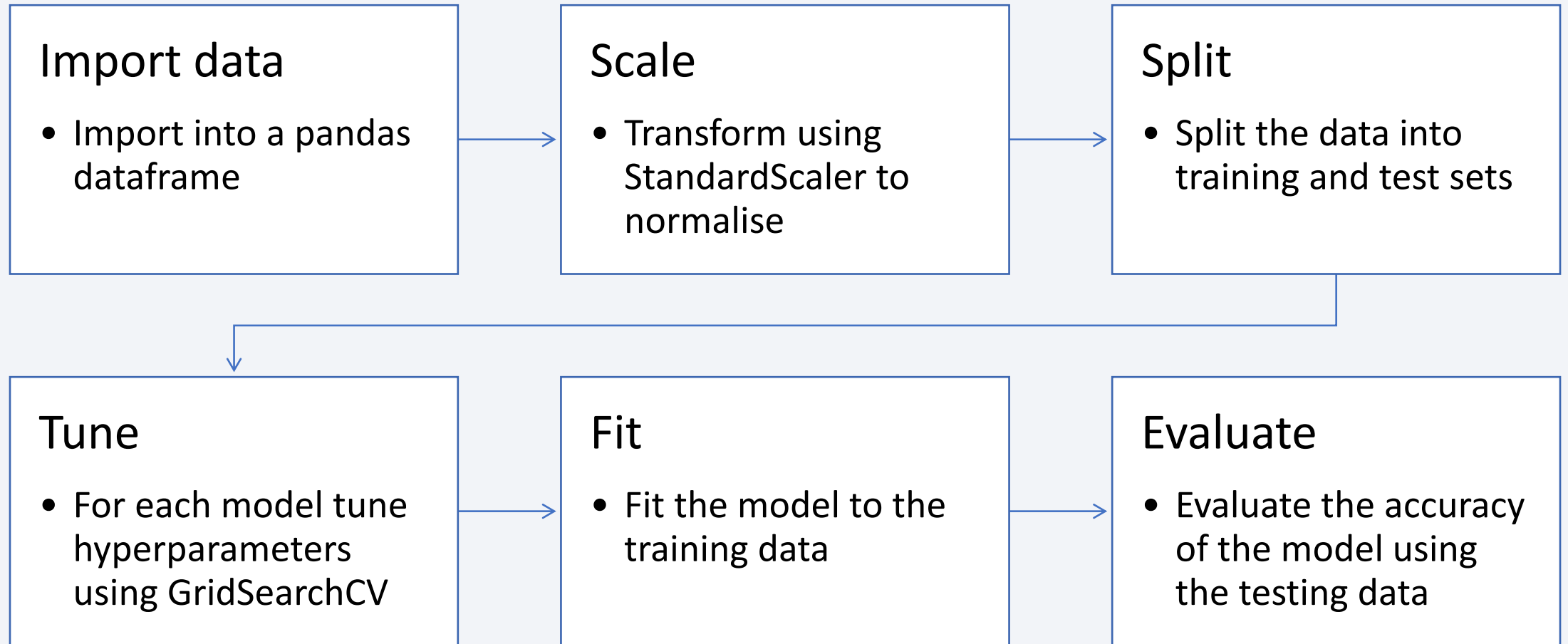


# Build a Dashboard with Plotly Dash

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- Pie chart which can show successes for each site or successes vs failures for one site (dropdown to select site)
- Scatter chart showing successes vs payload mass, each point was colour coded according to its launch site. The launch site is also affected by the site dropdown. There is a slider filter which allows restriction of the range of payload masses which are included

# Predictive Analysis (Classification)



# Results

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



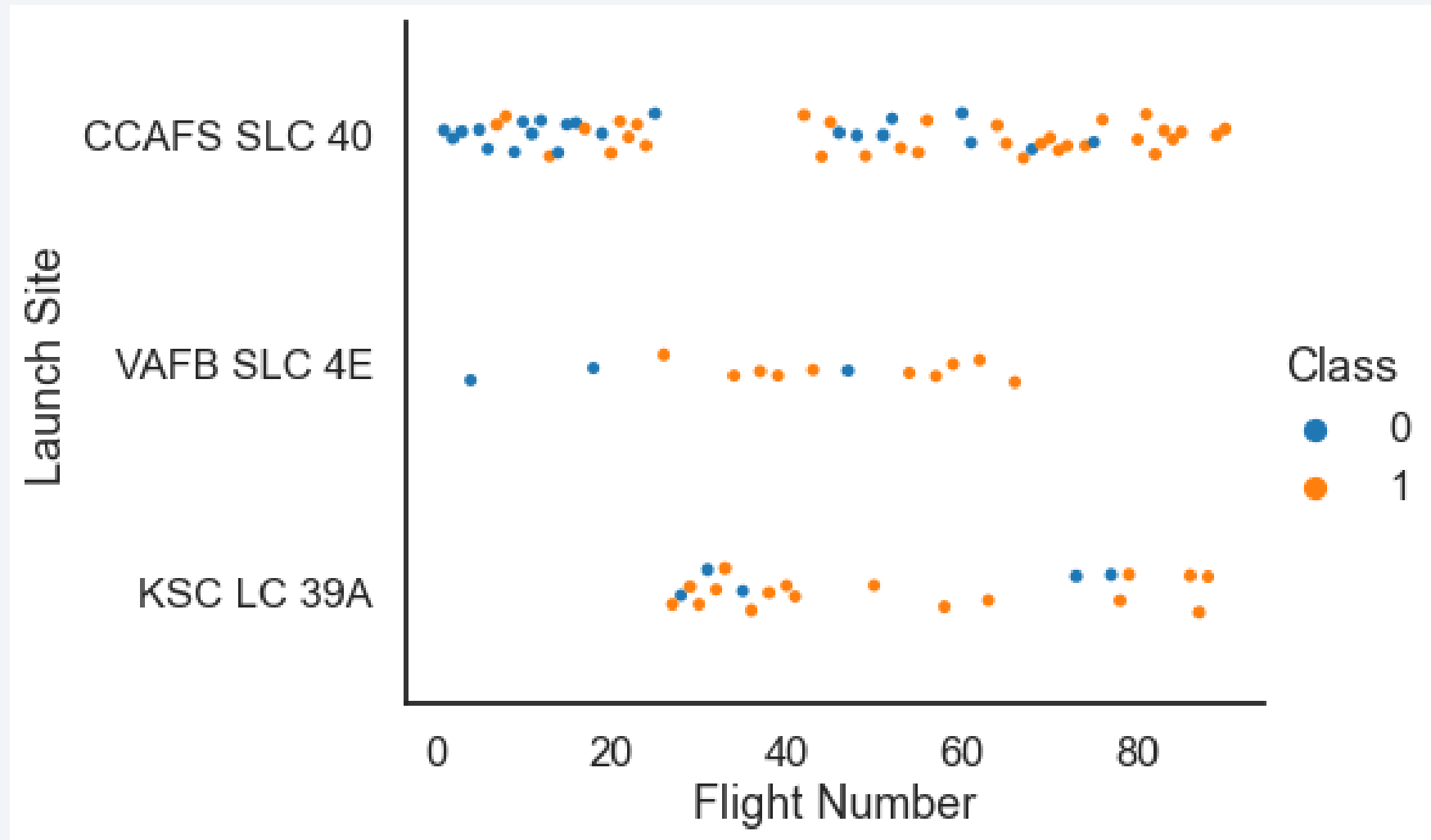
The background of the slide is an abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks in shades of red and cyan. A faint, light blue grid pattern is also visible, particularly in the lower half of the image. The overall effect is dynamic and technological.

Section 2

# Insights drawn from EDA

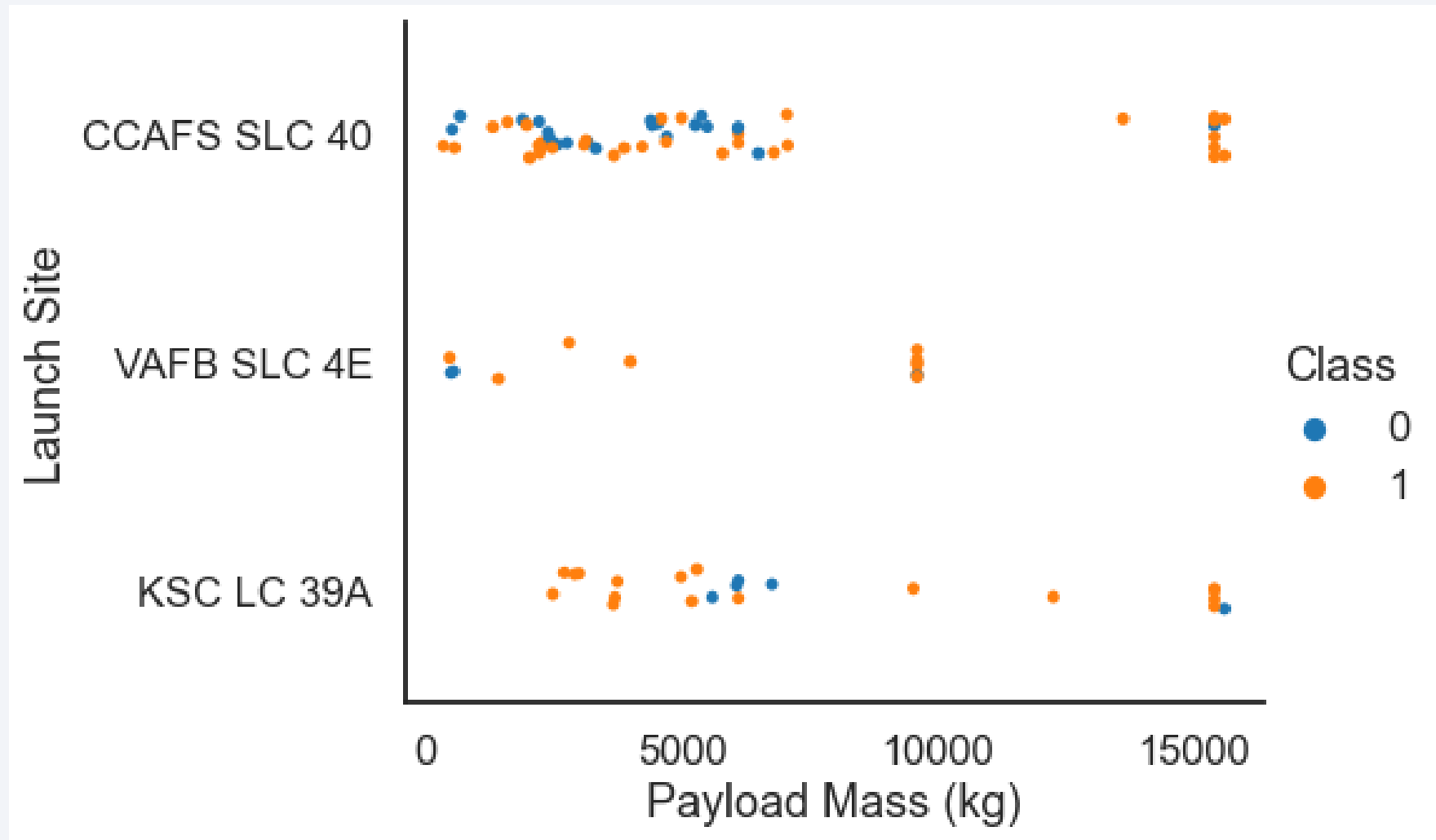


# Flight Number vs. Launch Site

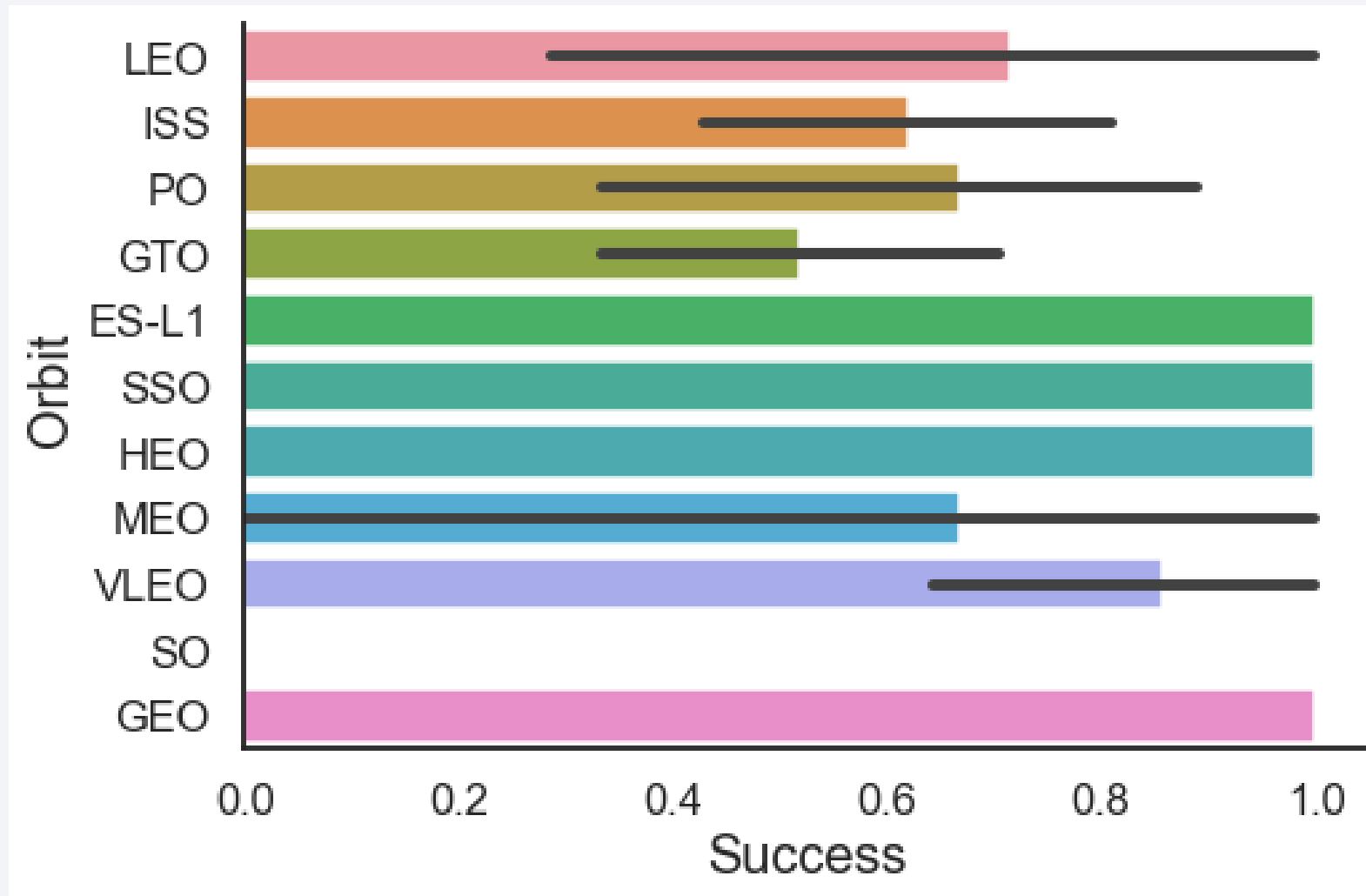




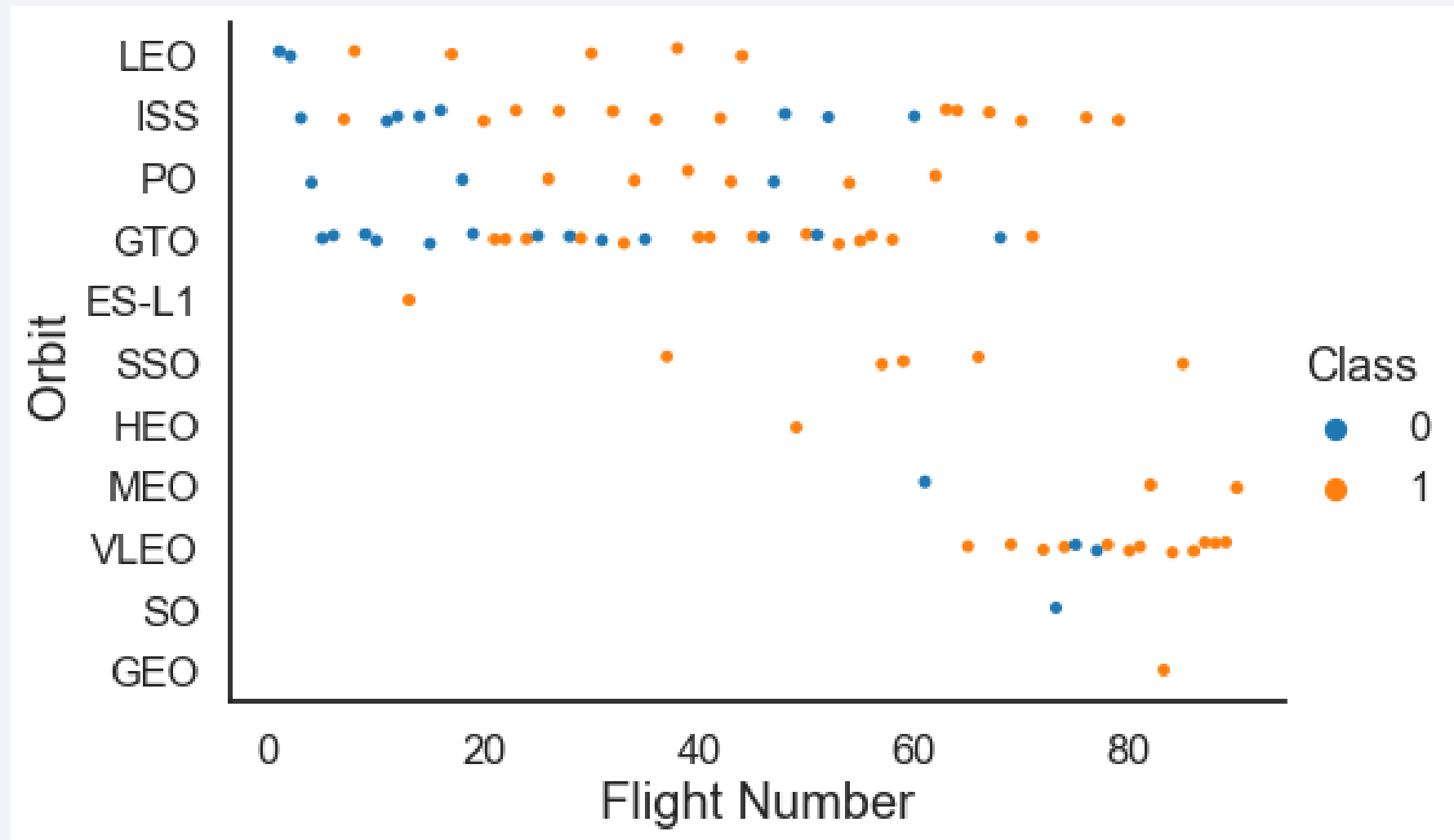
# Payload vs. Launch Site



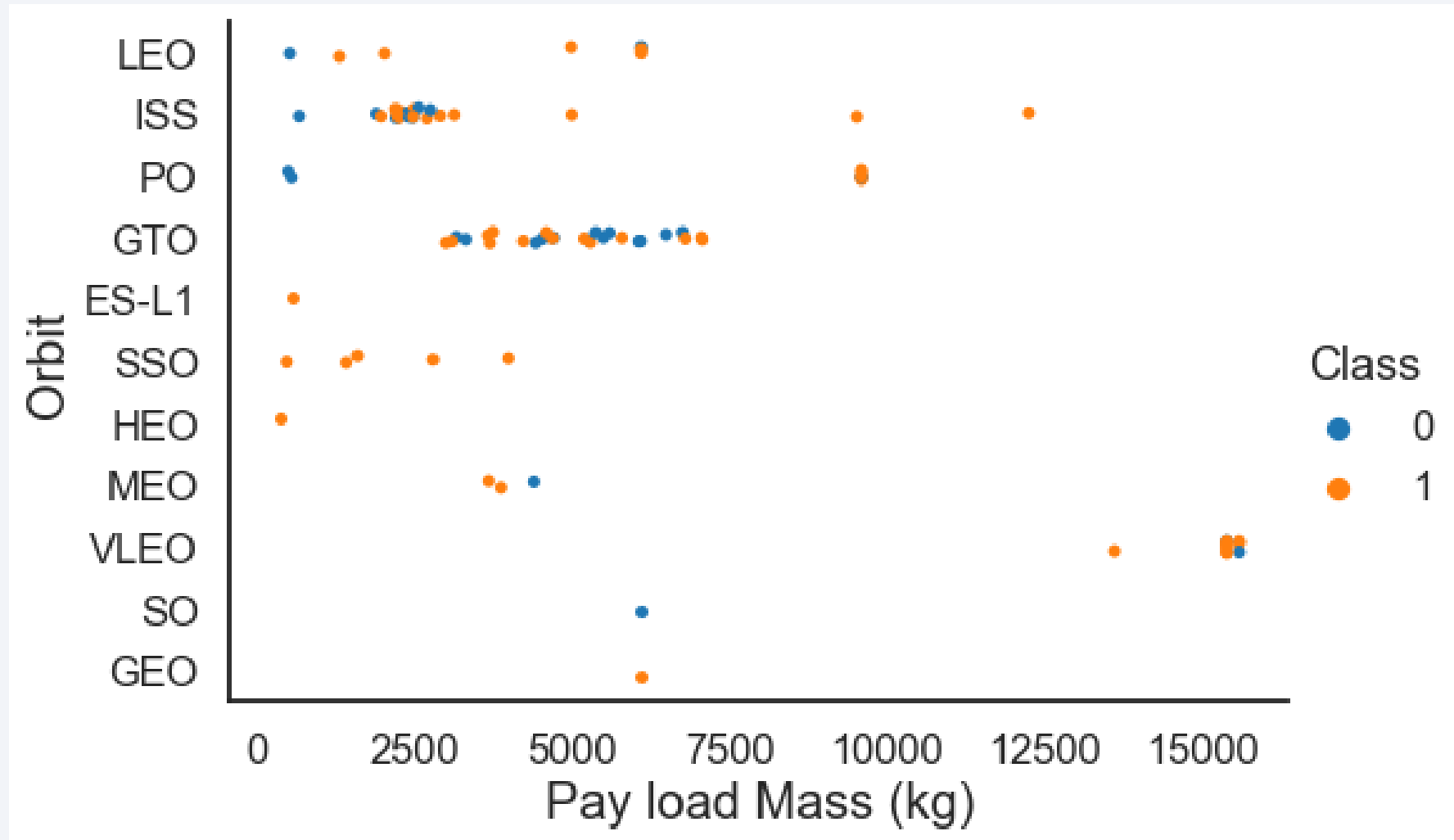
# Success Rate vs. Orbit Type



# Flight Number vs. Orbit Type

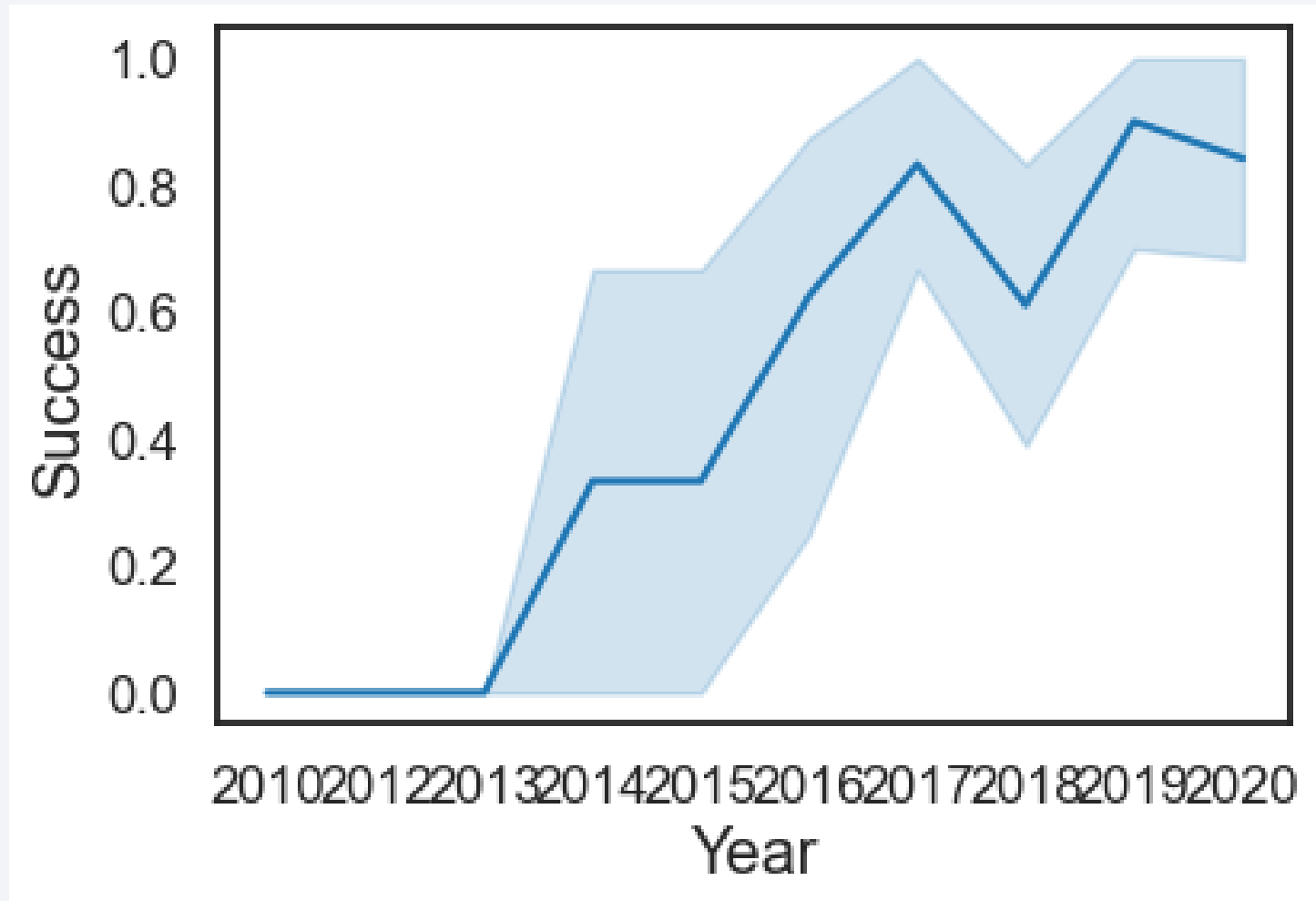


# Payload vs. Orbit Type



# Launch Success Yearly Trend

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# All Launch Site Names

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## Launch Site

0 CCAFS LC-40

1 VAFB SLC-4E

2 KSC LC-39A

3 CCAFS SLC-40

# Launch Site Names Begin with 'CCA'

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	<u>Date</u>	<u>Time (UTC)</u>	<u>Booster Version</u>	<u>Launch Site</u>	<u>Payload</u>	<u>Payload Mass KG</u>	<u>Orbit</u>	<u>Customer</u>	<u>Mission Outcome</u>	<u>Landing Outcome</u>
0	04-06-2010	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
1	08-12-2010	15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of...	0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
2	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
3	08-10-2012	00:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
4	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

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```
SELECT sum(Payload_Mass_KG) as Total FROM spacex WHERE Customer =  
'NASA (CRS)
```

Total

0 45596.0

# Average Payload Mass by F9 v1.1

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```
select avg(Payload_Mass_KG) as Average from spacex where Booster_Version  
like 'F9 v1.1%%'
```

Average

0 2534.6667

# First Successful Ground Landing Date

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```
select min(Date) as First_Date from spacex where Landing_Outcome = 'Success  
(ground pad)'
```

First\_Date

0 01-05-2017



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

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Booster\_Version

0 F9 FT B1022

1 F9 FT B1026

2 F9 FT B1021.2

3 F9 FT B1031.2

# Total Number of Successful and Failure Mission Outcomes

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Outcome	Count
0 Success	98
1 Failure (in flight)	1
2 Success (payload status unclear)	1
3 Success	1

# Boosters Carried Maximum Payload

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## Booster Version

- 0 F9 B5 B1048.4
- 1 F9 B5 B1049.4
- 2 F9 B5 B1051.3
- 3 F9 B5 B1056.4
- 4 F9 B5 B1048.5
- 5 F9 B5 B1051.4
- 6 F9 B5 B1049.5
- 7 F9 B5 B1060.2
- 8 F9 B5 B1058.3
- 9 F9 B5 B1051.6
- 10 F9 B5 B1060.3
- 11 F9 B5 B1049.7

# 2015 Launch Records

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	Date	Booster_Version	Launch_Site
0	2015-01-10	F9 v1.1 B1012	CCAFS LC-40
1	2015-04-14	F9 v1.1 B1015	CCAFS LC-40

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

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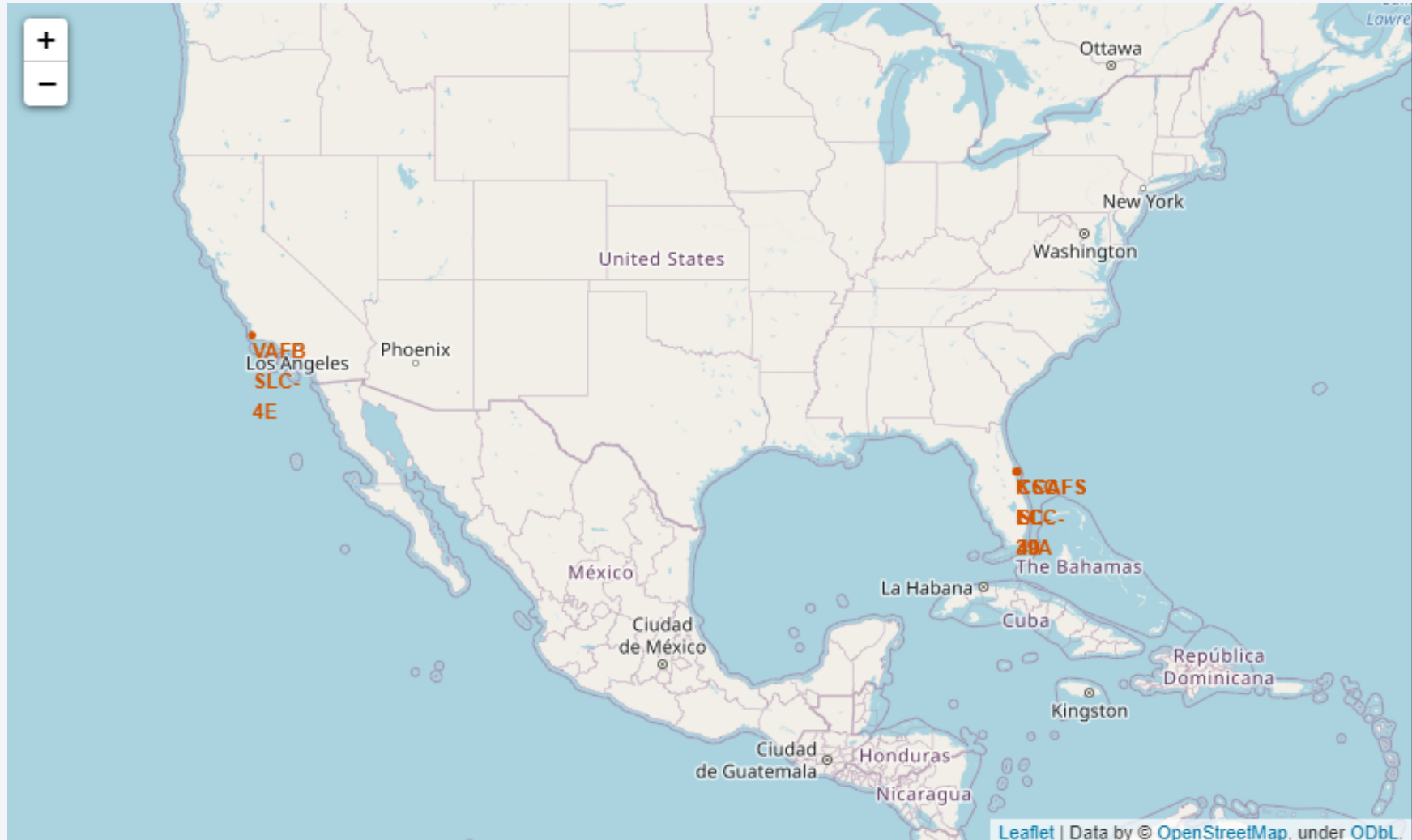
Outcome	Count
0 No attempt	10
1 Failure (drone ship)	5
2 Success (drone ship)	5
3 Controlled (ocean)	3
4 Success (ground pad)	3
5 Failure (parachute)	2
6 Uncontrolled (ocean)	2
7 Precluded (drone ship)	1

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.

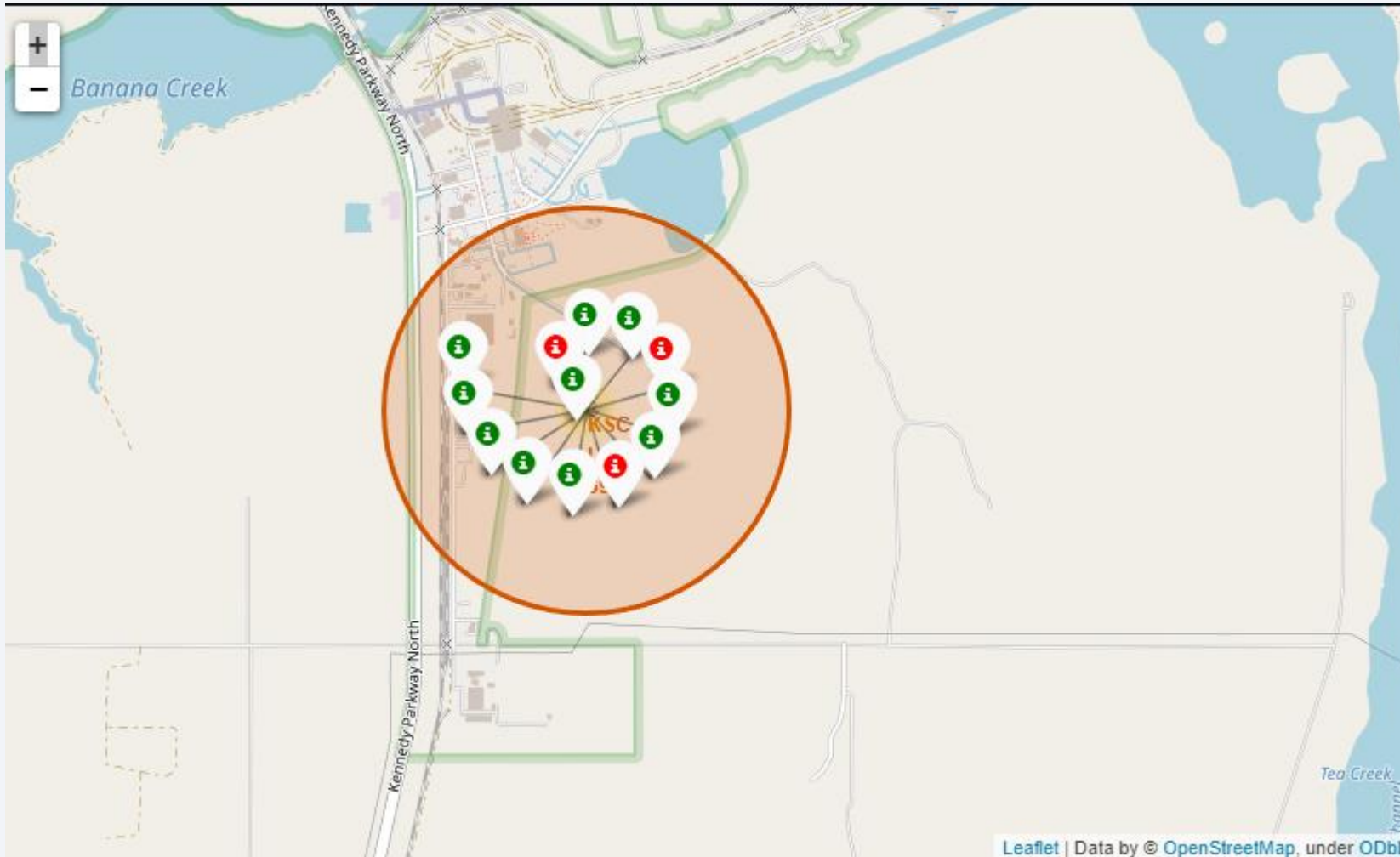
Section 3

# Launch Sites Proximities Analysis

# Launch Sites

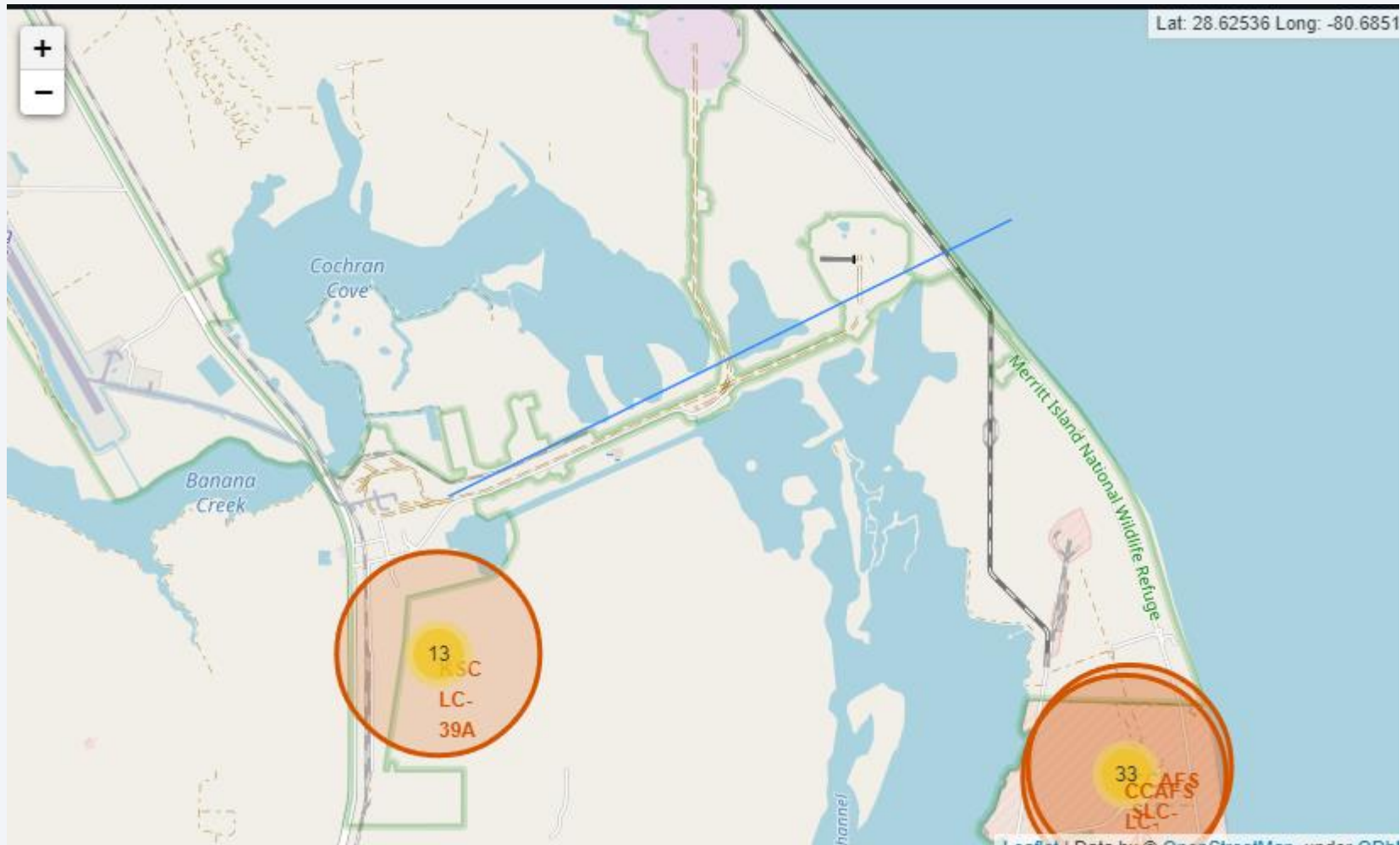


# Successes/Failures per launch site





# Distance of launch site to coast



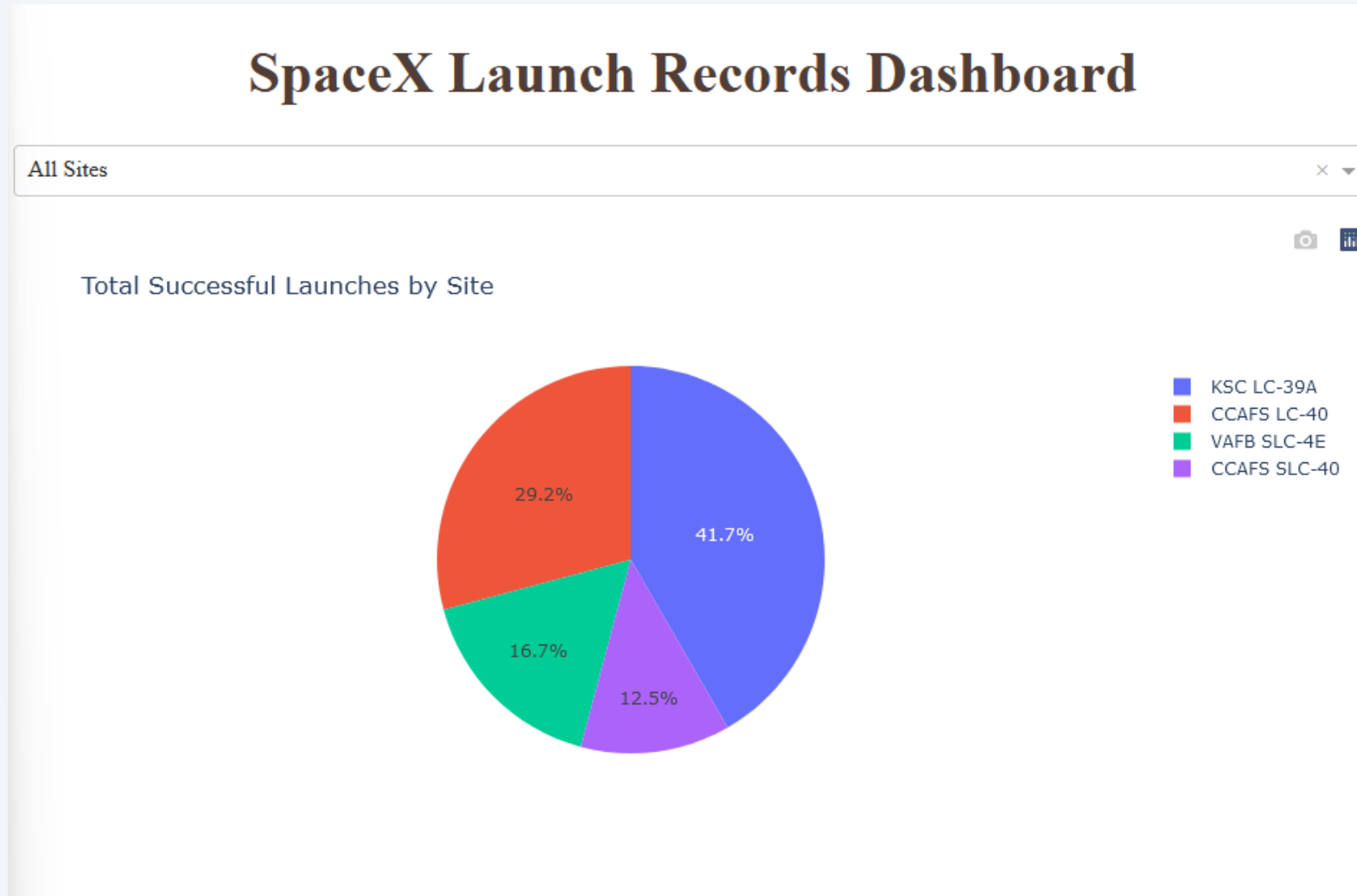


Section 4

# Build a Dashboard with Plotly Dash

# Total Successful Launches by Site

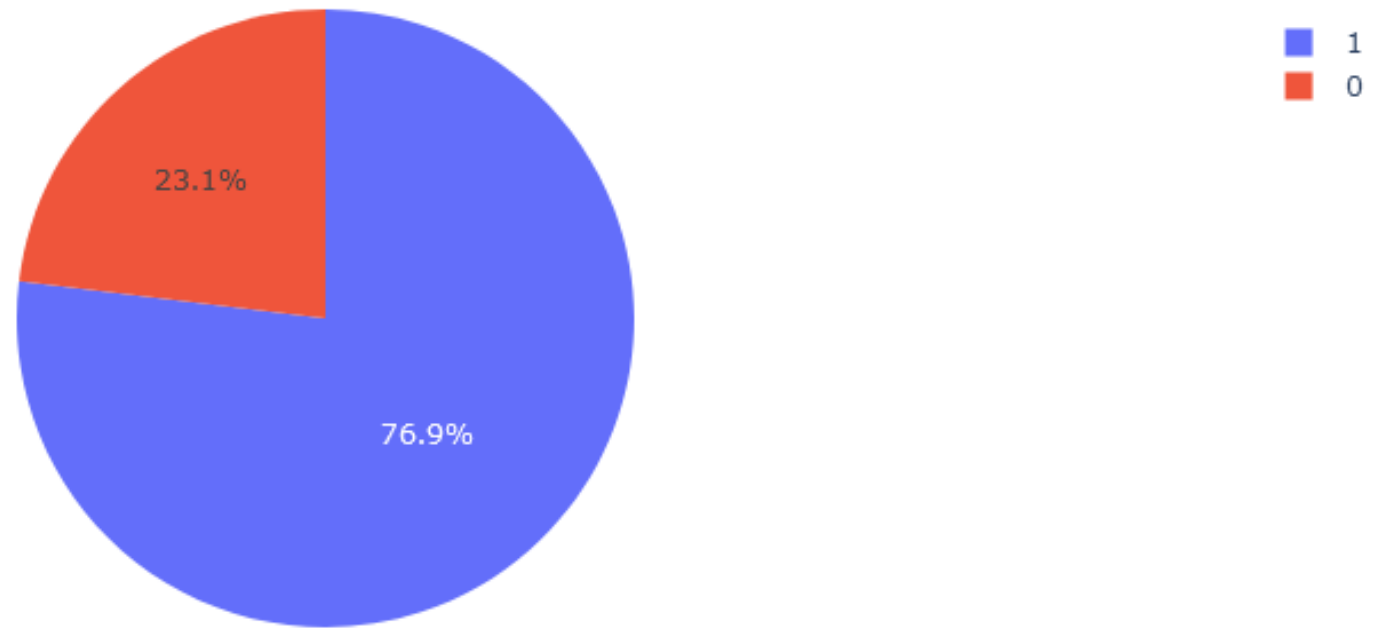
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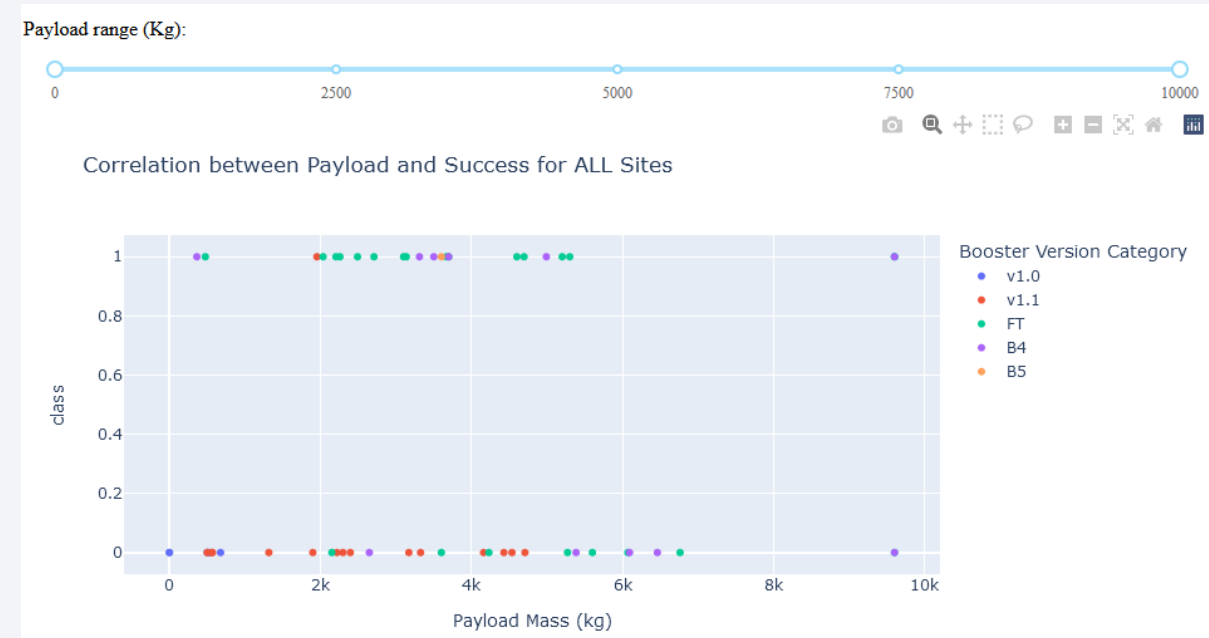
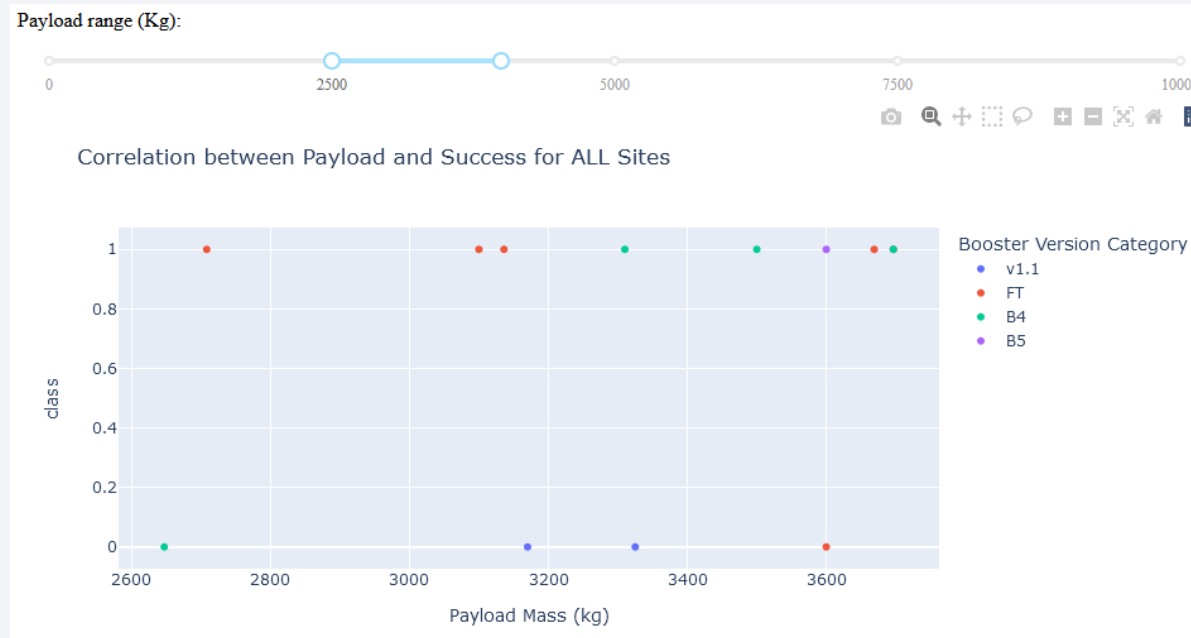
# Ratio of successful launches for KSC LC-39A

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Total Success Launches for KSC LC-39A



# Scatter chart of Payload Mass Vs Success



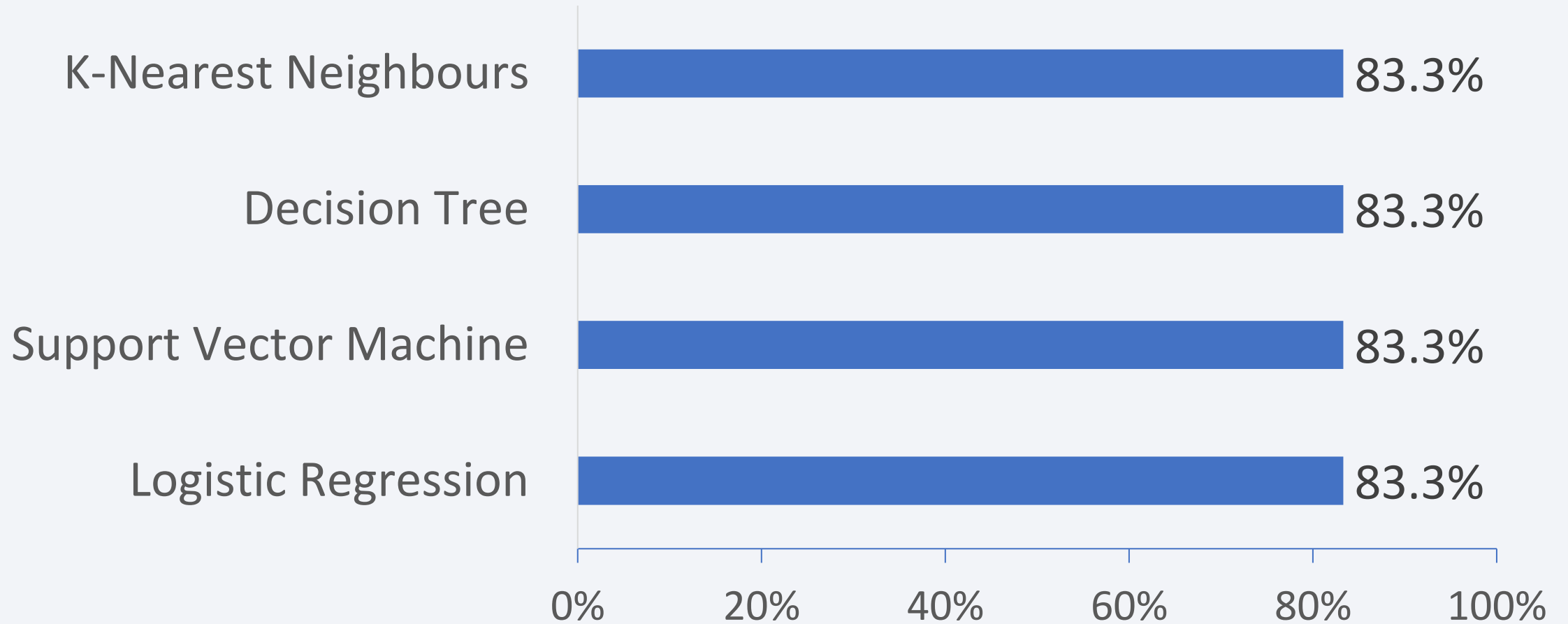


Section 5

# Predictive Analysis (Classification)

# Classification Accuracy

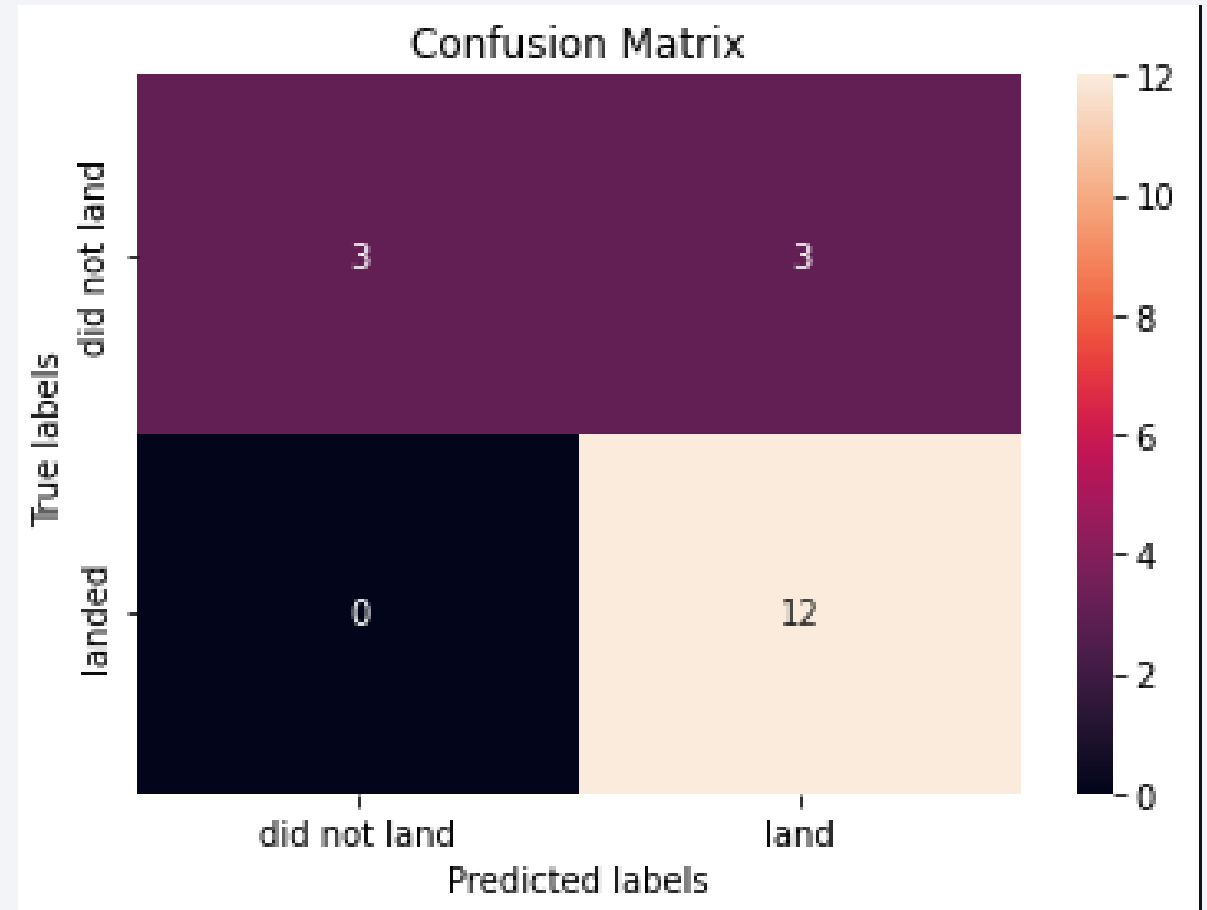
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# Confusion Matrix

- Our model correctly predicted 12 successful landings and 3 unsuccessful landings
- 3 unsuccessful landings were incorrectly predicted to be successful by our model (false positives)
- No landings that were successful were incorrectly predicted to be unsuccessful (false negatives)



# Conclusions

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- Data about the SpaceX launches can be scraped from Wikipedia or their API
- All models created from this data had the same overall accuracy
- Our models have false positives but not false negatives, therefore it can be used to predict if a stage will fail land
- The lack of overall data points limit the possibilities for testing the model

Thank you!

