IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

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Outline

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- Results
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Executive Summary

In the capstone project that we undertook, we wanted to predict the success rate of the SpaceX Falcon 9 first stage landing. Once we determine the success rate we can plan for the cost of a launch more accurately. We will use different machine learning algorithms to achieve similar result.

The methods used shall include data collection, data wrangling, pre-processing, exploratory analysis, visualization and then evaluating machine learning algorithms for the prediction.

There are some features of rocket launches which have direct correlation with the success rate of the landings. We shall describe them in detail.

The conclusion that is forming after following the methodology is that the algorithm Decision tree seems to be better at predicting the landing success.

Introduction

- The main goal of this capstone project is to predict whether the Falcon 9 first stage will land successfully. SpaceX prides itself in being able to reuse the first stage of a rocket launch so much so that they avertise on their website that their rocket launches cost 62 million while other provides cost upward 165 million. Much of these savings are down to the first stage's reusability. If we can determine if the first stage will land, we can determine the cost of a launch. This information can be used if an alternate company wants to bid against SpaceX for a rocket launch.

- This brings us to our main question that we are trying to answer : For a given set of features about a Falcon 9 rocket launch, will the first stage of the rocket land successfully?

Section 1

Methodology

Methodology

Executive Summary

- Data collection methodology:
  - Web scraping from Wikipedia page
  - Requesting data from SpaceX API
- Perform data wrangling
  - Transformation and cleaning using Pandas library
- Perform exploratory data analysis (EDA) using matplotlib and seaborn libraries
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Four different ML classification models are engaged. Logistic regression, Support vector machine, K-nearest neighbor, and decision tree classifier. Each model was trained, tuned and deployed to find the best one.

Data Collection – Scraping using SpaceX API

- Using GET request, extract the spacex launch data
- Normalize JSON output to a Data Frame (DF)
- Extract useful columns using functions
- Create new pandas DF from dictionary
- Filter DF to include only Falcon 9 Launches
- Handle missing values
- Export to CSV file

GIT HUB URL:

Data Collection API

Data Wrangling

- Calculate the number of launches per site
- Calculate the number and occurrence of each orbit
- Calculate the mission outcomes per orbit type
- Create a landing outcome label from outcome column using one-hot encoding
- Export to CSV

GITHUB URL:
EDA Data Wrangling

EDA with Data Visualization

- Scatter plots to represent the relationship between two variables, namely, Flight number versus launch site, Payload versus launch site, Flight number versus Orbit type, Payload versus Orbit type.
- Bar charts were used to compare multiple groups with category on X axis and discrete values on Y axis. Success rates were compared for orbit types.
- Line charts were used to show trends, rate of success over a number of years

GITHUB URL:
EDA with Data Visualization

EDA with SQL

- Using bullet point format, summarize the SQL queries you performed
- 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

- Objects were created and added to a Folium map. Marker objects were used to show all launch sites on a map as well as the successful/failed launches for each site on the map. Line objects were used to calculate the distances between a launch site to its proximities
- By adding these objects, following geographical patterns about launch sites are found:
  - Are launch sites in close proximity to railways? Yes
  - Are launch sites in close proximity to highways? Yes
  - Are launch sites in close proximity to coastline? Yes
  - Do launch sites keep certain distance away from cities? Yes

GITHUB URL:
IVA with Folium

Build a Dashboard with Plotly Dash

- The dashboard application contains two charts:
- A pie chart that shows the successful launch by each site. This chart is useful as you can visualize the distribution of landing outcomes across all launch sites or show the success rate of launches on individual sites.
- A scatter chart that shows the relationship between landing outcomes an the payload mass of different boosters. The dashboard takes two inputs, namely the site(s) and payload mass. This chart is useful as you can visualize how different variables affect the landing outcomes,

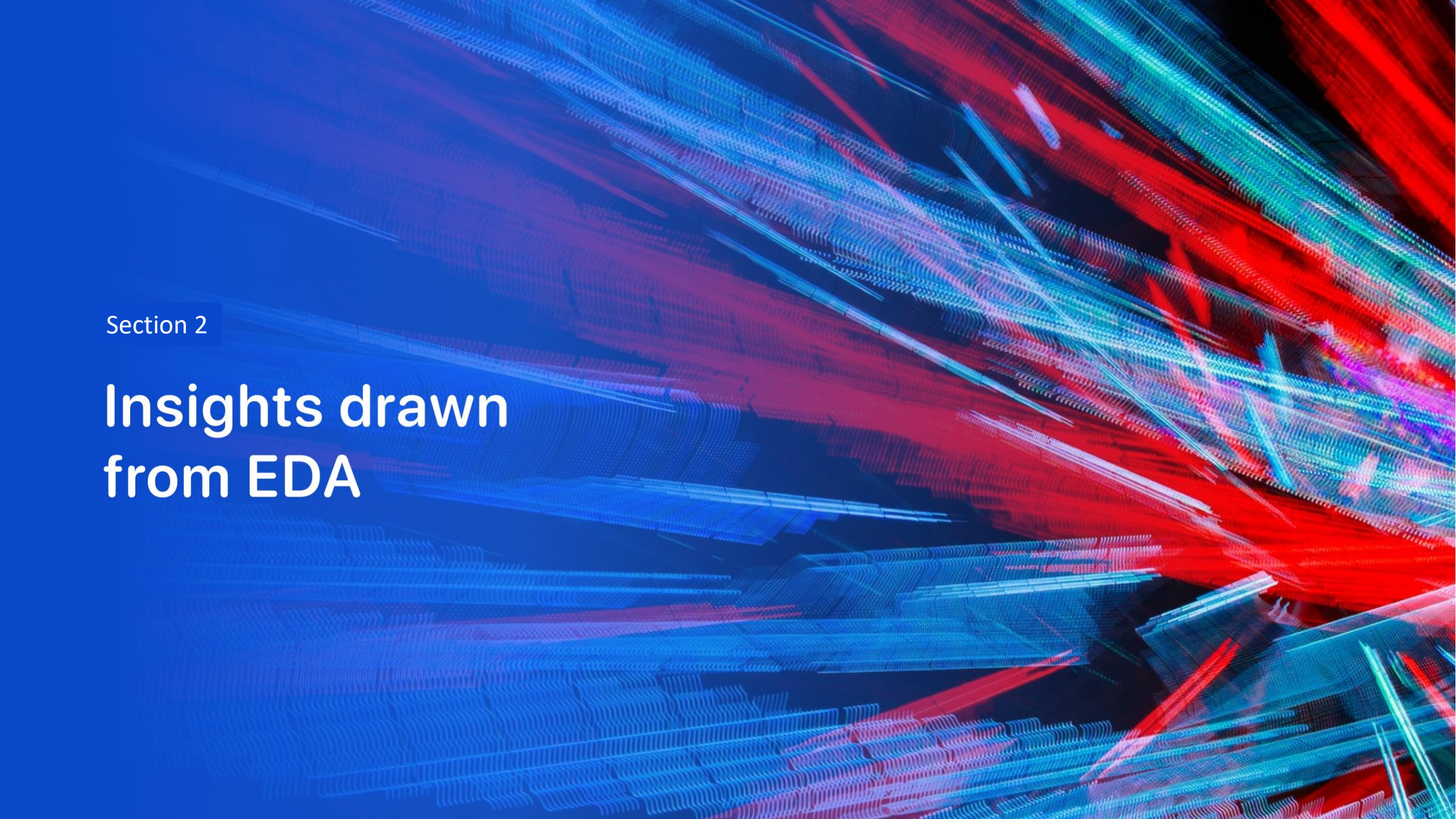
Predictive Analysis (Classification)

- A Class column is created
- Data is standardized
- Data is split into training and test sets
- Run the algorithms and find the best hyper parameter
- Determine accuracy scores
- Evaluate models using accuracy scores and confusion matrix

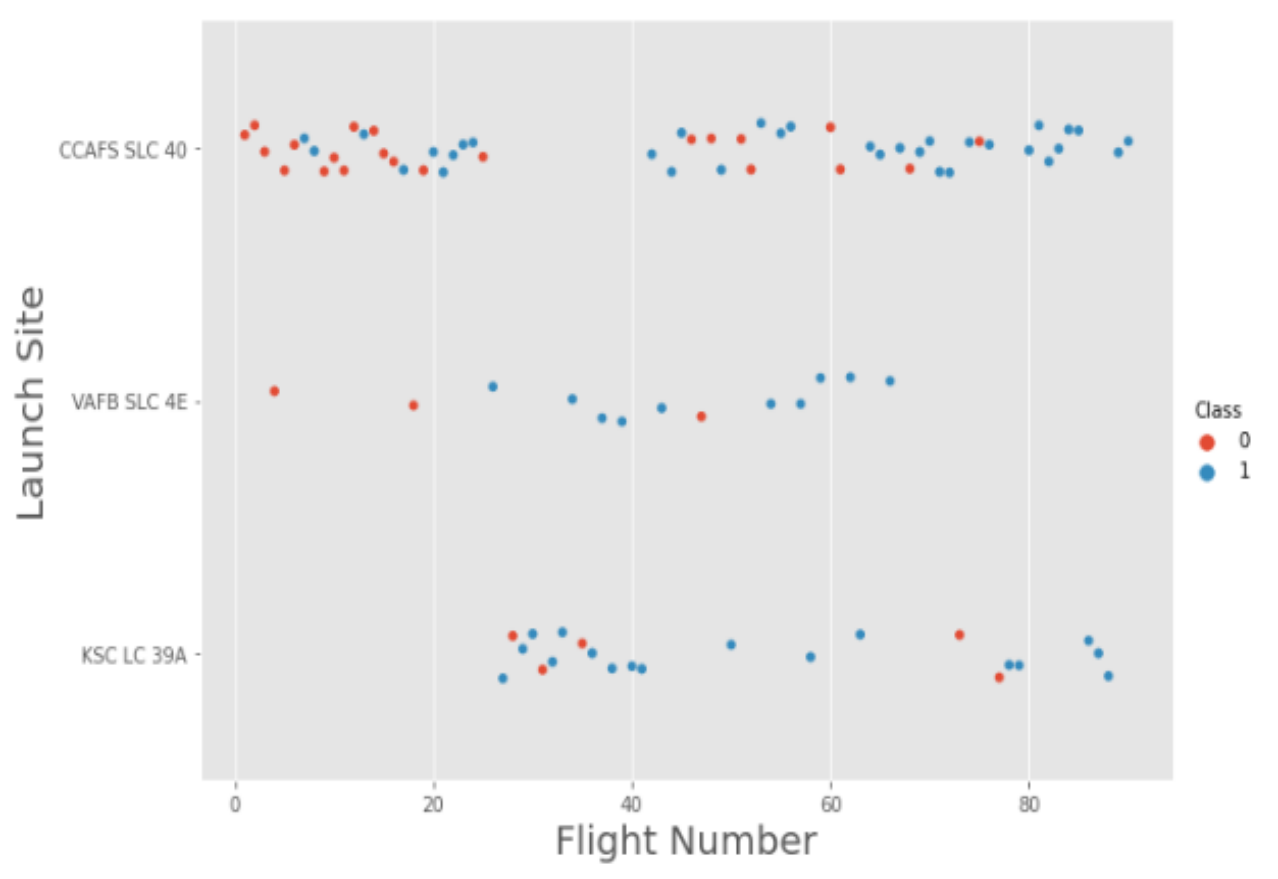
GITHUB URL:
ML Predictions and results

Results

- The results of the exploratory data analysis revealed that the success rate of the Falcon 9 landings was 66.66%
- The predictive analysis results showed that the Decision Tree algorithm was the best classification method with an accuracy of 94%

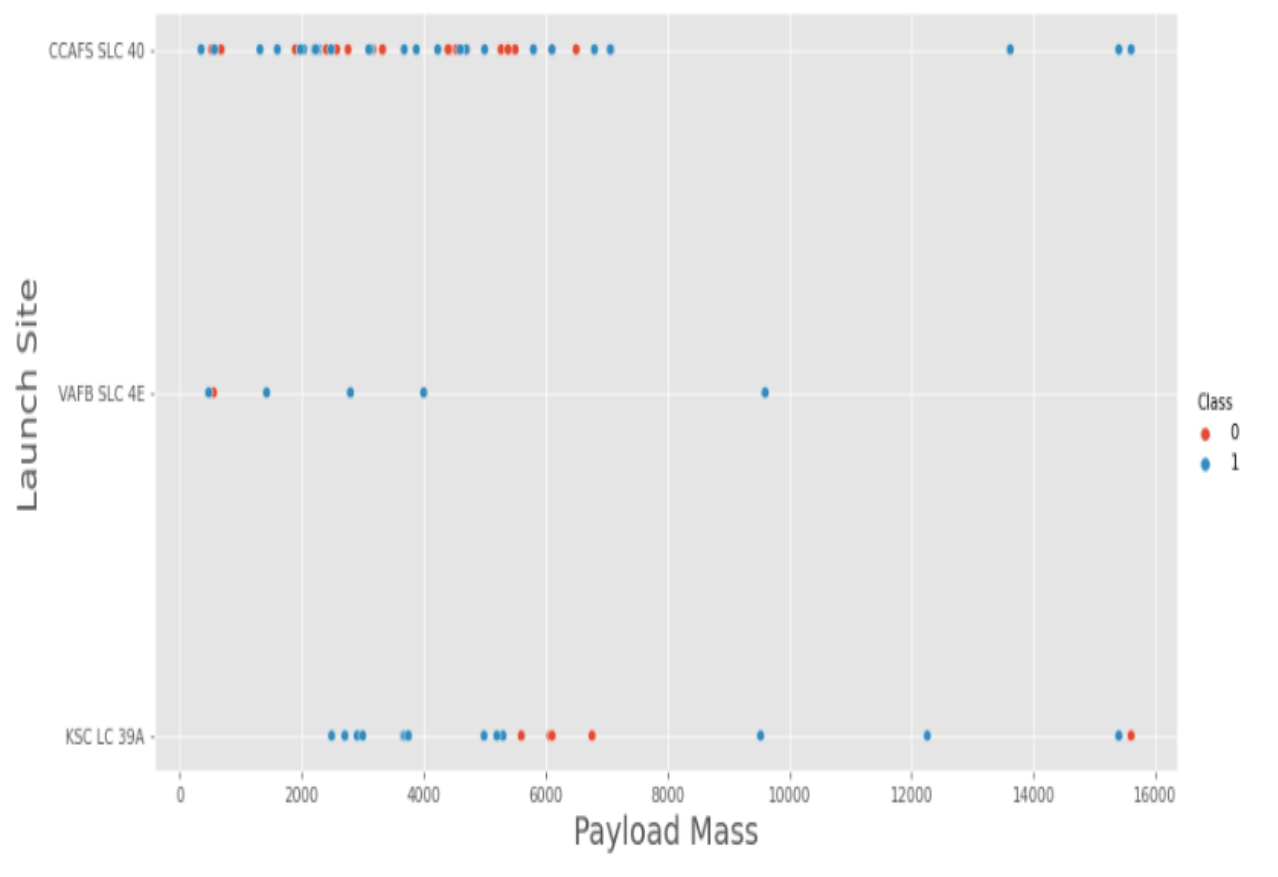

Section 2
Insights drawn from EDA

Flight Number vs. Launch Site



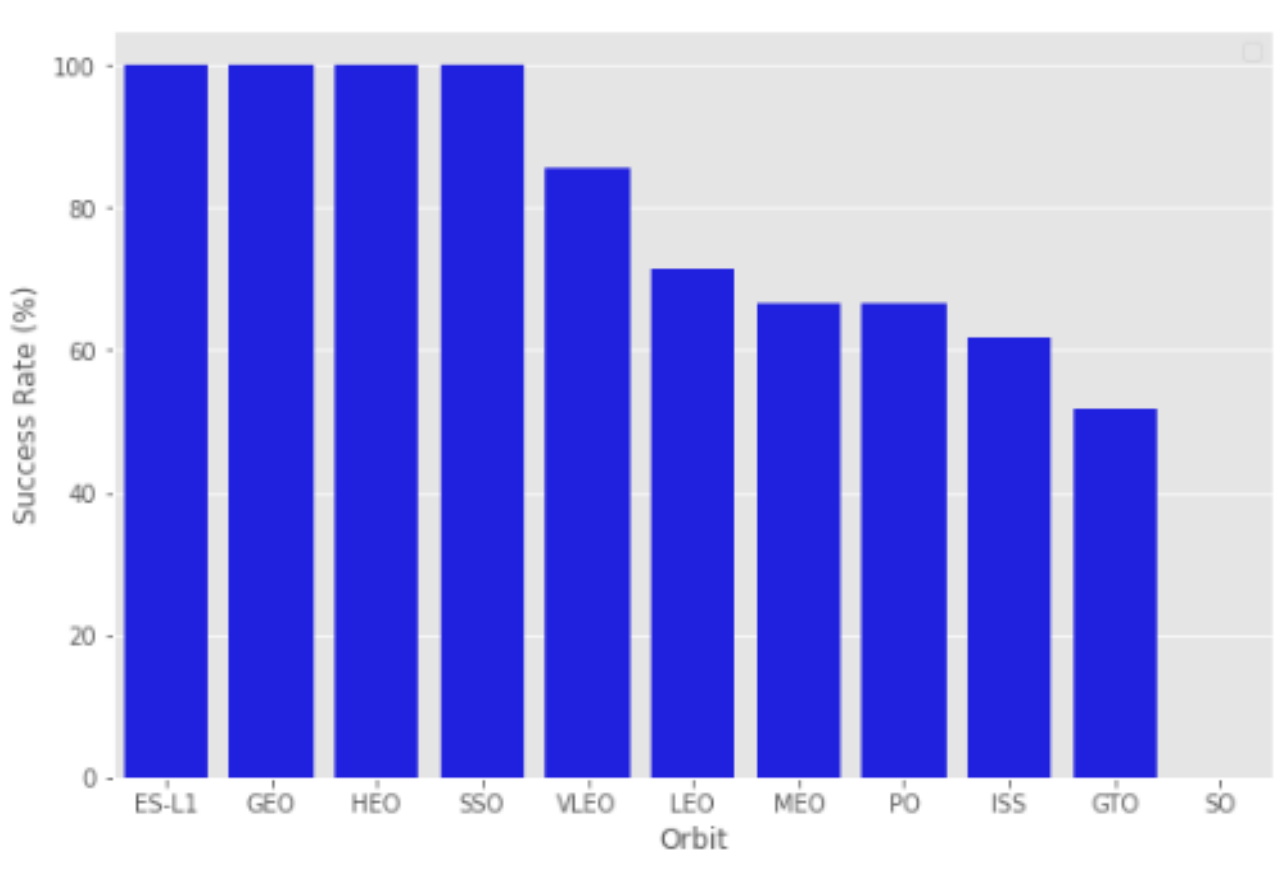
- This figure shows that the success rate increased as the number of flights increased.
- The blue dots represent the successful launches while the red dot represent unsuccessful launches.
- An increase in successful flights after the 40th launch is noticed.

Payload vs. Launch Site



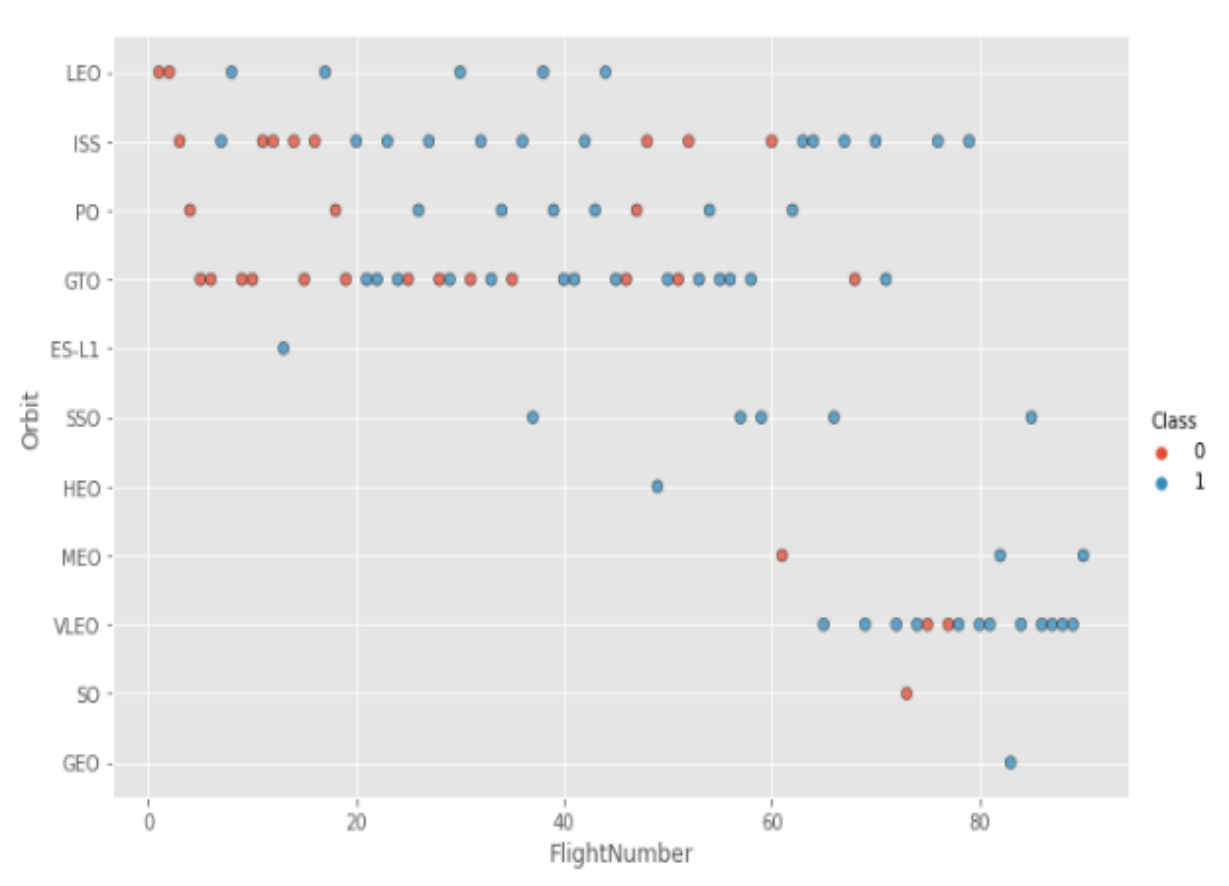
- The blue dots represent the successful launches while the red dots represent unsuccessful launches.
- For the VAFB-SLC launch site there are no rockets launched for heavy payload mass
- There seems to be a weak correlation between Payload and Launch Site and therefore decisions cannot be made using this metric.

Success Rate vs. Orbit Type



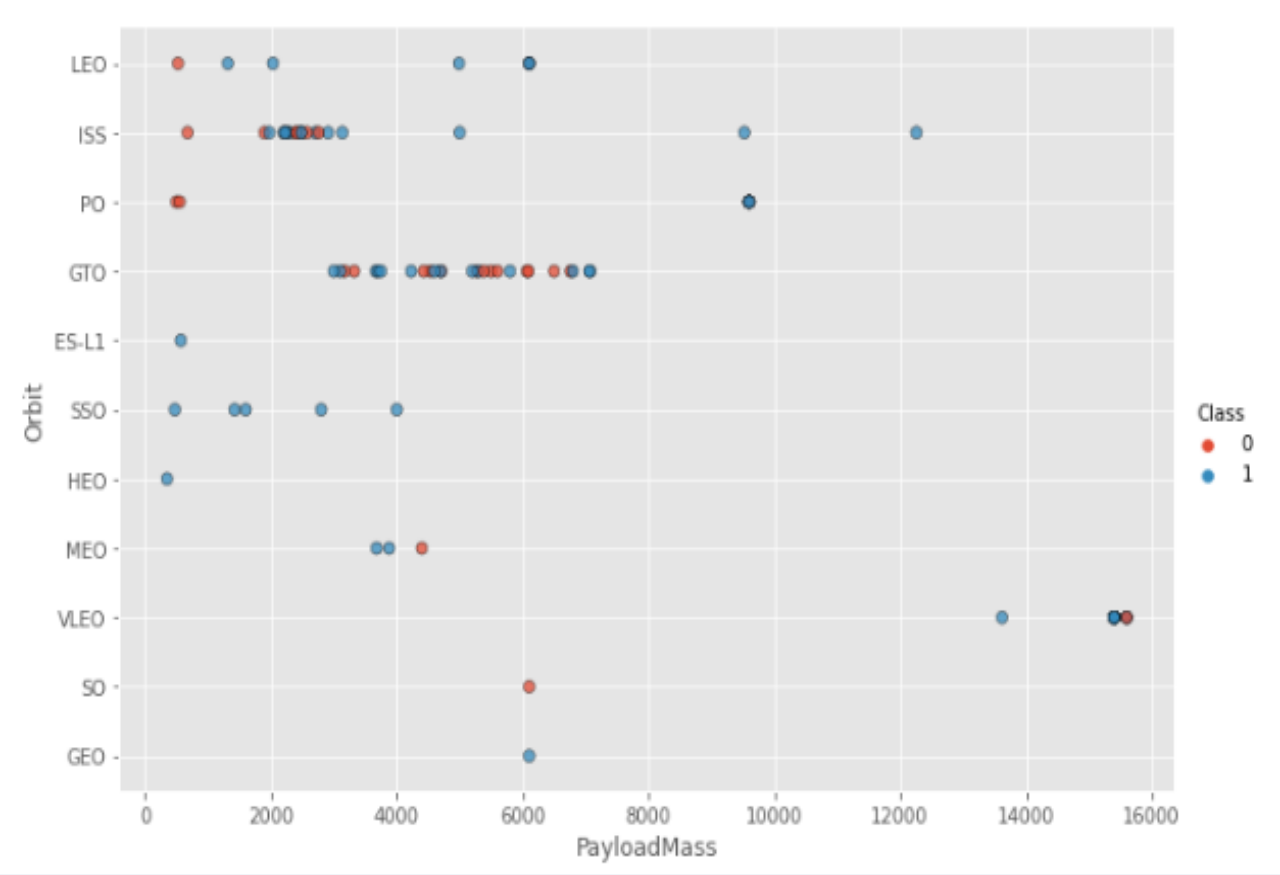
- Orbits SSO, HEO, GEO, and ES-L1 have 100% success rates.
- SO orbit did not have any successful launches with a 0% success rate.

Flight Number vs. Orbit Type



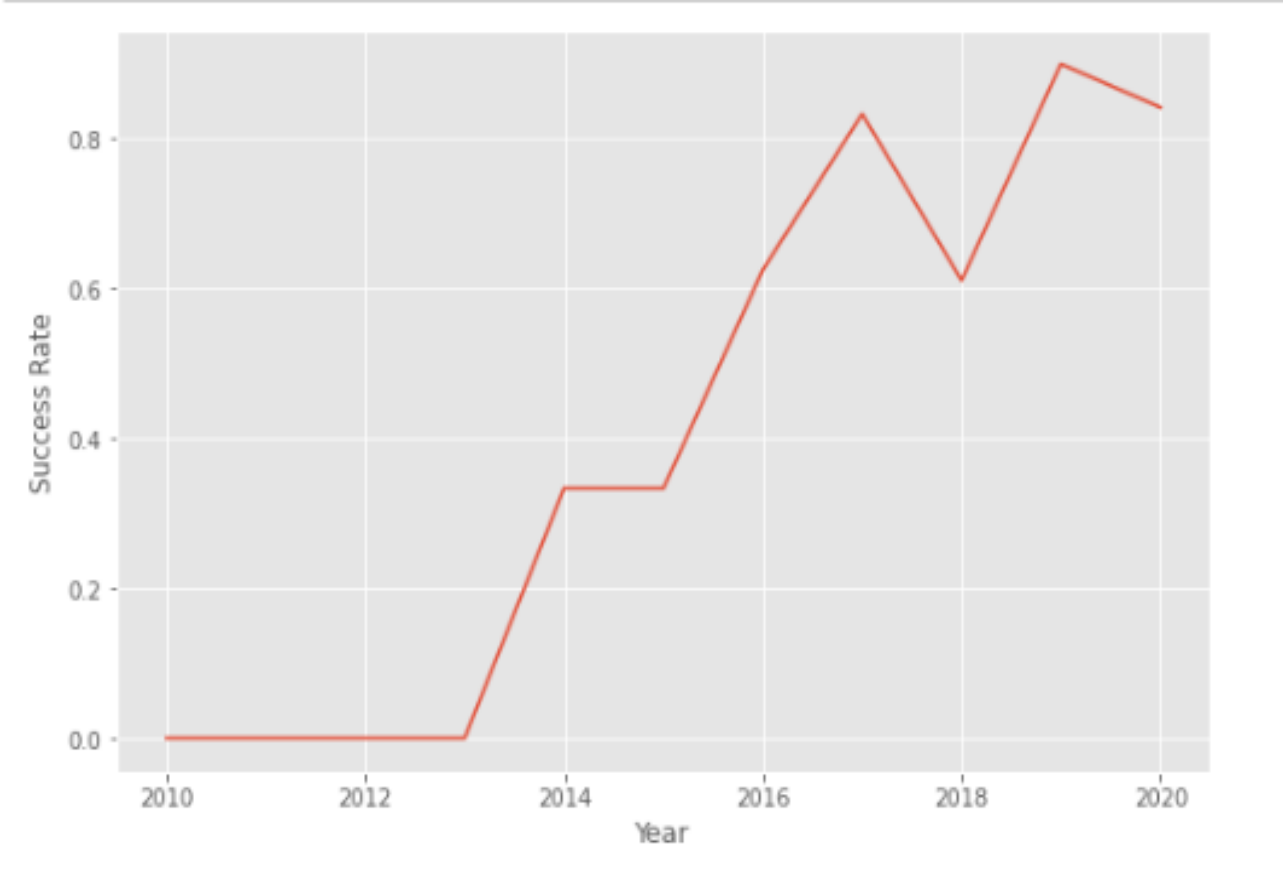
- In the LEO orbit, the success is positively correlated to the the number of flights.
- There seems to be no relationship between flight number in the GTO orbit.
- The SSO orbit has a 100% success rate however with fewer flights than the other orbits
- Flights numbers greater than 40 have a higher success rate than flight numbers between 0-40.

Payload vs. Orbit Type



- As the payloads get heavier, the success rate increases in the PO, SSO, LEO and ISS orbits.
- There seems to be no direct correlation between obrit type and payload mass for GTO orbit as both successful and failed launches are eually present

Launch Success Yearly Trend



- The general trend of the chart shows an increase in landing success rate as the years pass. There is however a dip in 2018 as well as in 2020.

All Launch Site Names

- The DISTINCT clause was used to return only the unique rows from the launch_site column.
- The names of the launch sites are CCAFS LC-40, CCAFS SLC-40, KSC LC-39A, VAFB SLC-4E .

launch_site
CCAFS LC-40
CCAFS SLC-40
KSC LC-39A
VAFB SLC-4E

Launch Site Names Begin with 'CCA'

- The LIMIT and LIKE clauses were used to display only the top five results where the launch_site name starts with 'CCA'

DATE	time__utc_	booster_version	launch_site	payload	payload_mass__kg_	orbit	customer	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	07:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
2012-10-08	00:35:00	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

- The SUM() function was used to the calculate the total payload carried by boosters from NASA from the payload_mass__kg column.

total_payload_mass_kg
45596

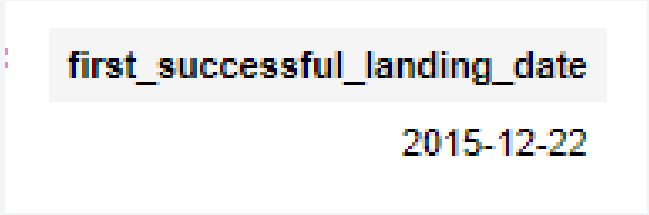
Average Payload Mass by F9 v1.1

- The AVG() function was used to the calculate the average payload the average payload mass carried by booster version F9 v1.1
- The WHERE clause was used to filter results so that the calculations were only performed on booster_versions only if they were named “F9 v1.1”

avg_payload_mass_kg
2928

First Successful Ground Landing Date

- The MIN(DATE) function was used to find the date of the first successful landing outcome on ground pad
- The WHERE clause ensured that the results were filtered to match only when the 'landing_outcome' column is 'Success (ground pad)'



first_successful_landing_date
2015-12-22

Successful Drone Ship Landing with Payload between 4000 and 6000

- The BETWEEN clause was used to retrieve only those results of payload mass greater than 4000 but less than 6000. The WHERE clause filtered the results to include only boosters which successfully landed on drone ship

booster_version
F9 FT B1022
F9 FT B1026
F9 FT B1021.2
F9 FT B1031.2

Total Number of Successful and Failure Mission Outcomes

- The COUNT() function is used to count the number of occurences of different mission outcomes with the help of the GROUPBY clause applied to the 'mission_outcome' column. A list of the total number of successful and failure mission outcomes os returned.
- There have been 99 successful mission outcomes out of 101 missions.

mission_outcome	total_number
Failure (in flight)	1
Success	99
Success (payload status unclear)	1

Boosters Carried Maximum Payload

- The MAX() function was used in a subquery to retrieve a list of boosters which have carried the maximum payload mass

booster_version	payload_mass__kg_
F9 B5 B1048.4	15600
F9 B5 B1048.5	15600
F9 B5 B1049.4	15600
F9 B5 B1049.5	15600
F9 B5 B1049.7	15600
F9 B5 B1051.3	15600
F9 B5 B1051.4	15600
F9 B5 B1051.6	15600
F9 B5 B1056.4	15600
F9 B5 B1058.3	15600
F9 B5 B1060.2	15600
F9 B5 B1060.3	15600

2015 Launch Records

- The SELECT statement was used to retrieve multiple columns from the table. The YEAR(DATE) function was used to retrieve only those rows with a 2015 launch date.

landing__outcome	booster_version	launch_site
Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40

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

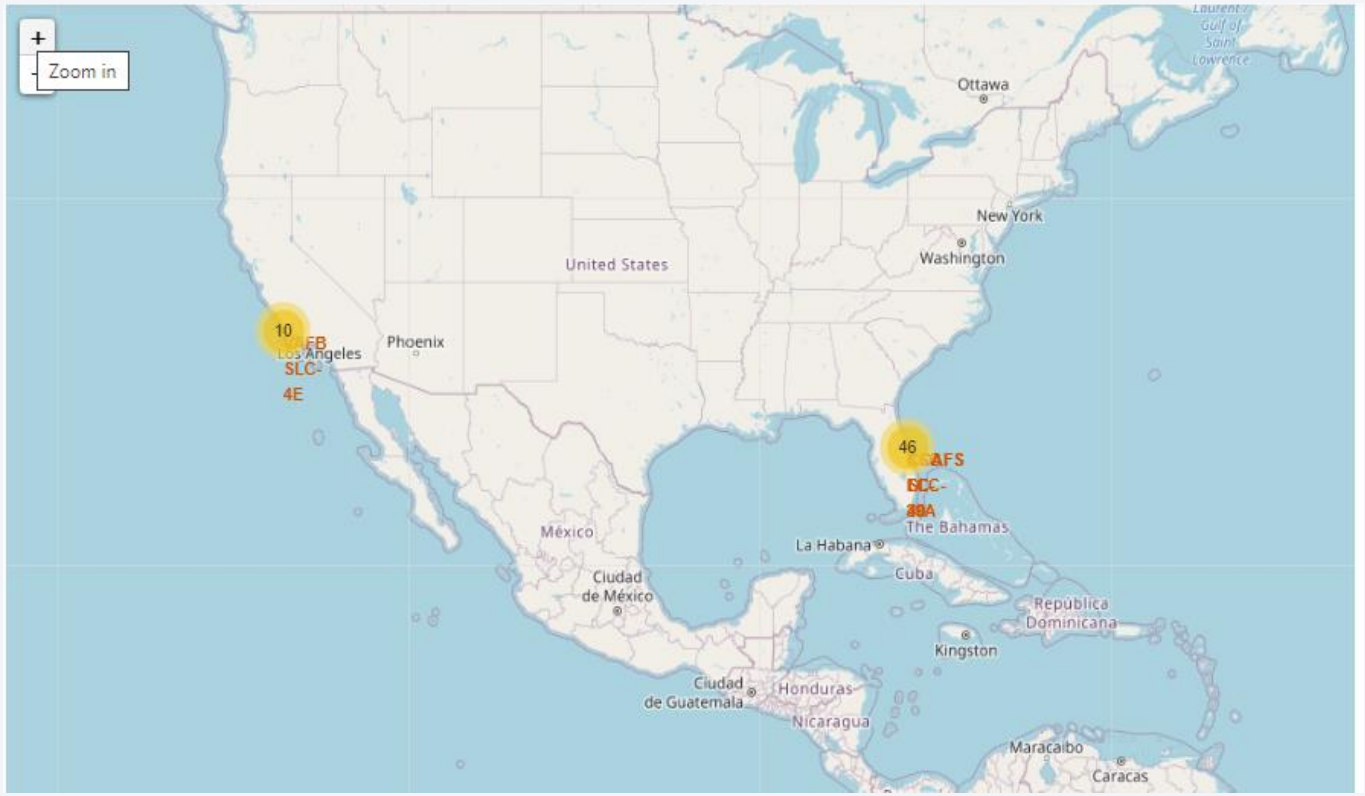
- COUNT() function was used to count the different landing outcomes.'The WHERE and BETWEEN clauses filtered the results to only include results between 2010-06-04 and 2017-03-20. The GROUPBY clause ensure that the counts were grouped by their outcome. The ORDERBY and DESC clauses were used to sort the results by descending order.

landing__outcome	total_number
No attempt	10
Failure (drone ship)	5
Success (drone ship)	5
Controlled (ocean)	3
Success (ground pad)	3
Failure (parachute)	2
Uncontrolled (ocean)	2
Precluded (drone ship)	1

Section 3

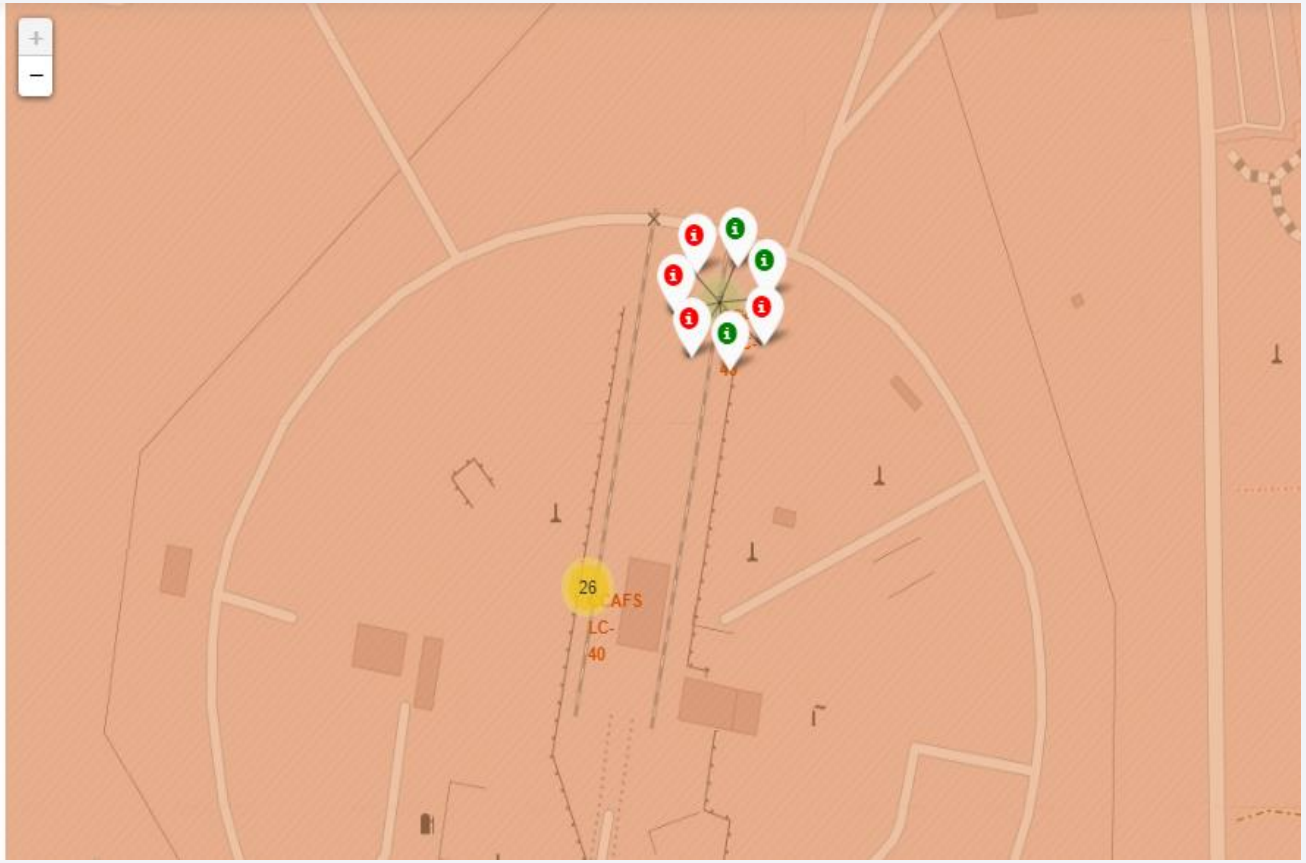
Launch Sites Proximities Analysis

Launch site locations



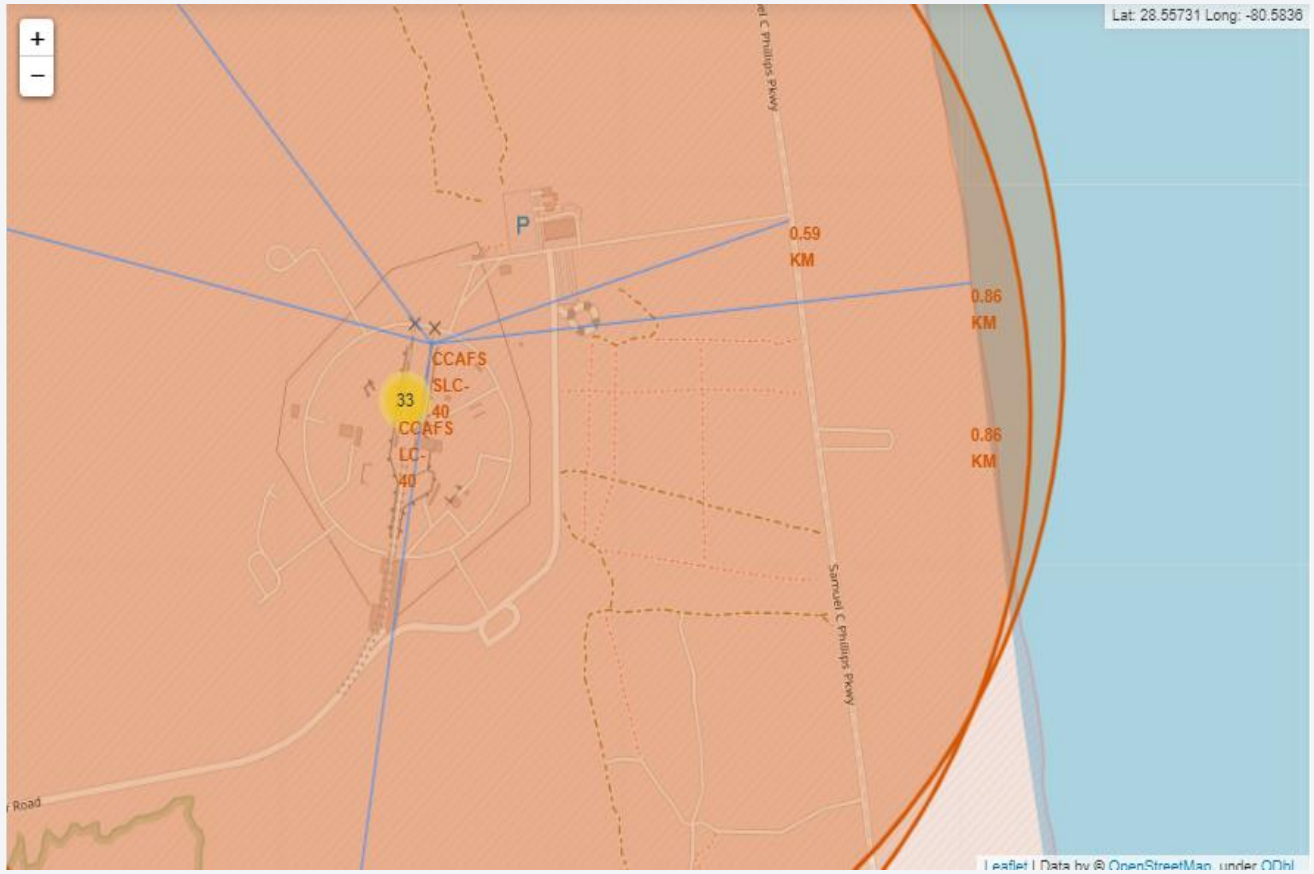
- The yellow markers are indicators of where the locations of all the SpaceX launch sites are situated in the US.
- The launch sites have been strategically placed near the coast

Success or failure clusters

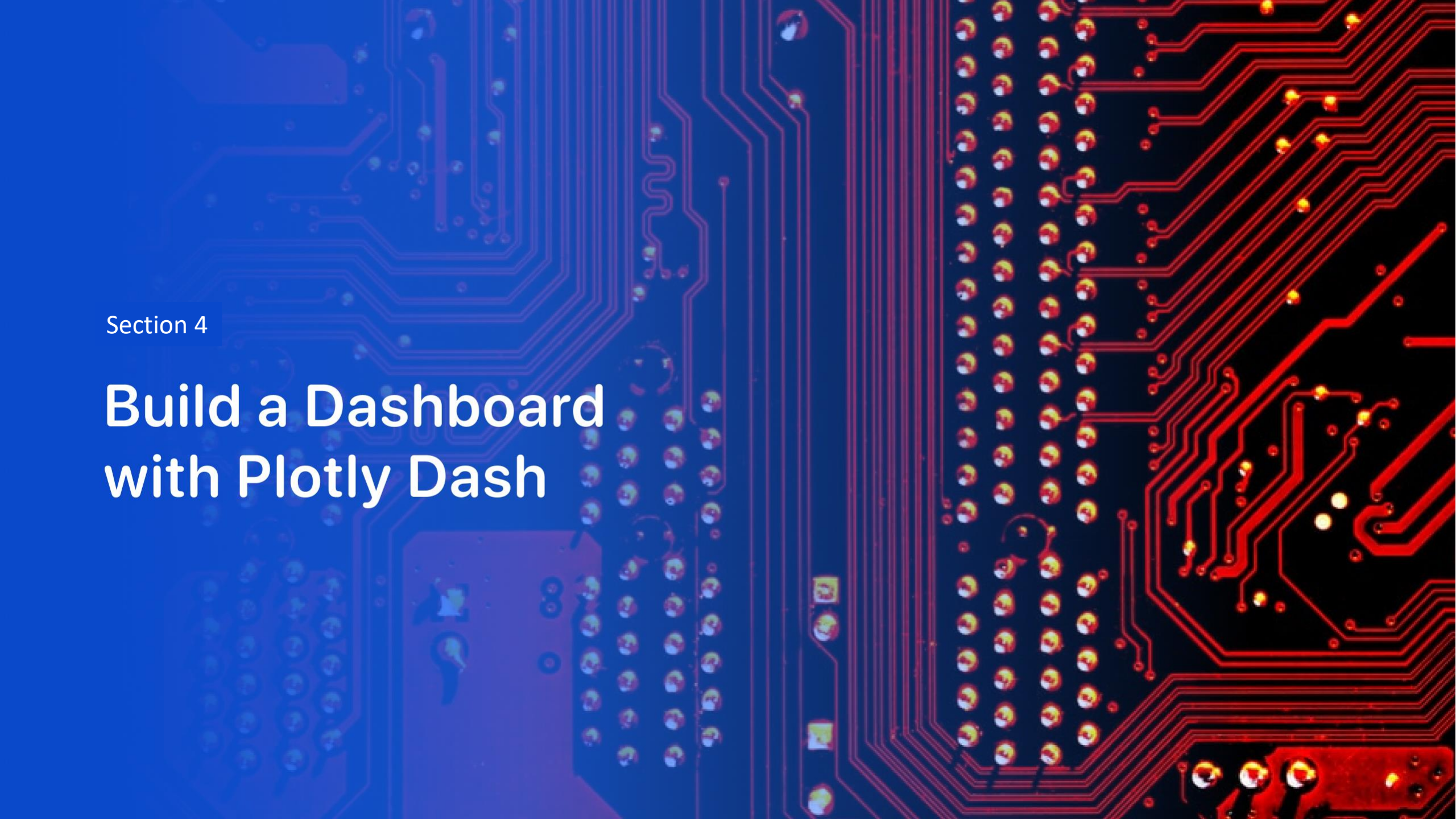


- When we zoom in on a launch site, we can click on the launch site which will display marker clusters of successful landings (green) or failed landing (red).

Site proximities

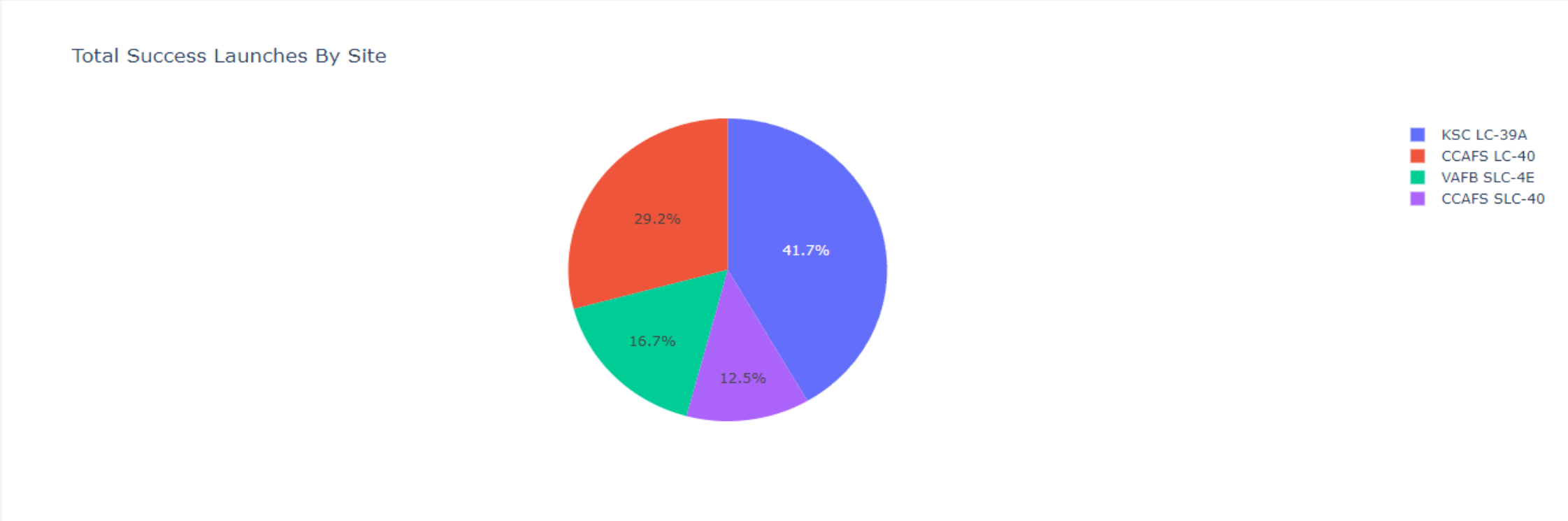


- The generated map shows that the selected launch site is close to a highway for transportation of personnel and equipment. The launch site is also close to the coastlines for launch failure testing.
- The launch sites also maintain a certain distance from the cities. (Can be viewed in notebook).


Section 4
Build a Dashboard with Plotly Dash

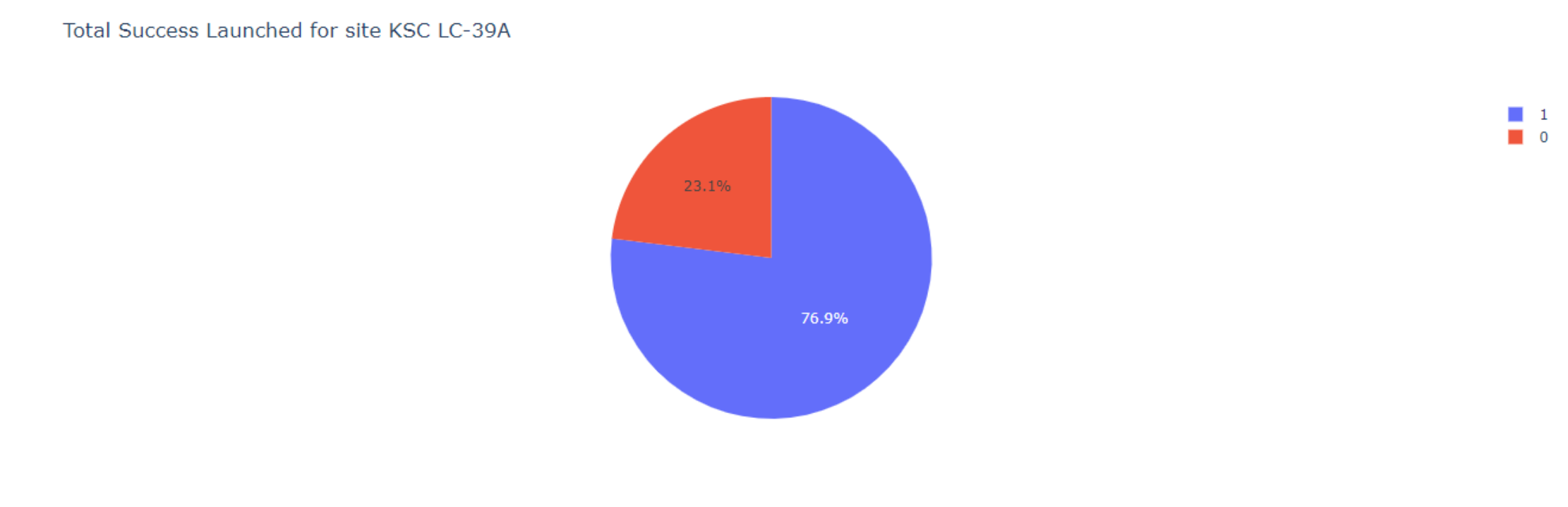
Successful launches by site

- The KSC LC-39A Launch site has the most successful launches with 10 in total.



Launch site success ratio

- The KSLC-39A has the highest success rate with 76.9%.



Payloads vs. launch outcome

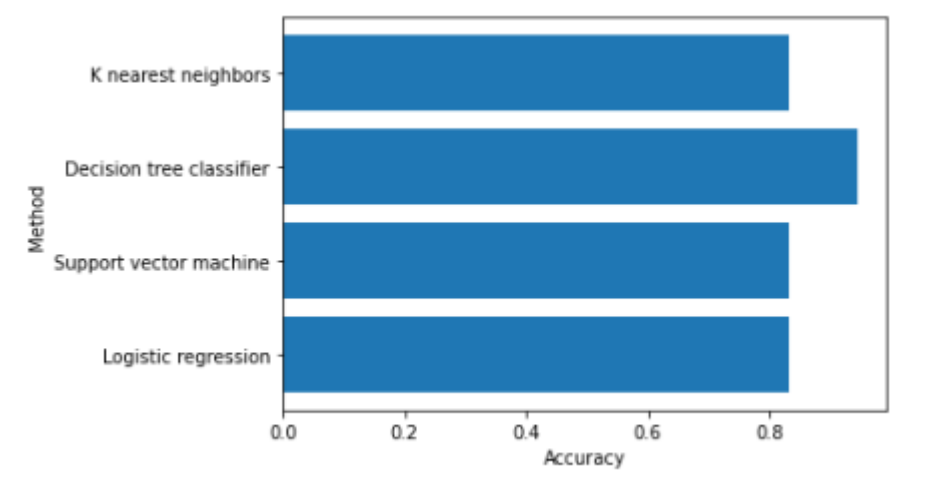
- The launch success rate for payloads 0-2500 kg is slightly lower than that of payloads 2500-5000 kg. There is in fact not much difference between the two.
- The booster version that has the largest success rate, in both weight ranges is the v1.1.

Section 5

Predictive Analysis (Classification)

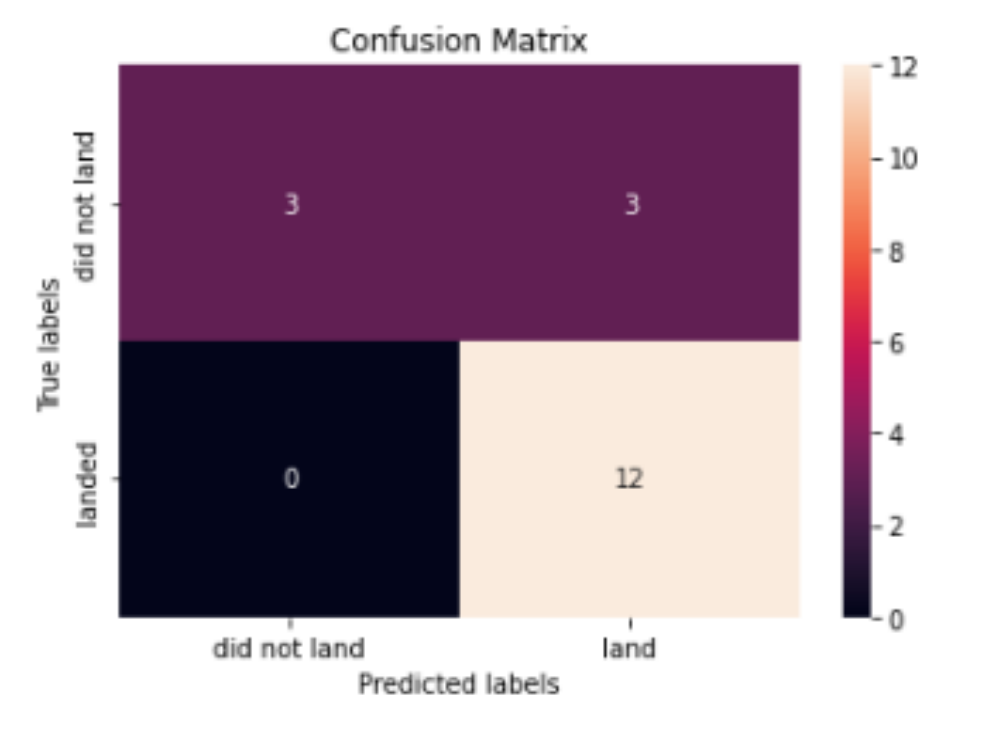
Classification Accuracy

- The Decision Tree classifier had the best accuracy at 94%.



	method	accuracy
0	Logistic regression	0.833333
1	Support vector machine	0.833333
2	Decision tree classifier	0.944444
3	K nearest neighbors	0.833333

Confusion Matrix



- The model predicted 12 successful landings when the True label was successful (True Positive) and 3 unsuccessful landings when the True label was failure (True Negative).
- The model also predicted 3 successful landings when the True label was unsuccessful landing (False Positive).
- The model generally predicted successful landings.

Conclusions

- The analysis showed that there is a positive correlation between number of flights and success rate as the success rate has improved over the years.
- There are certain orbits like SSO, HEO, GEO, and ES-L1 where launches were the most successful.
- Success rate can be linked to payload mass as the lighter payloads generally proved to be more successful than the heavier payloads.
- The launch sites are strategically located near highways and railways for transportation of personnel and cargo, but also far away from cities for safety.
- The best predictive model to use for this dataset is the Decision Tree Classifier as it had the highest accuracy with 94%.

Appendix

- GITHUB with all code and results:
  - https://github.com/MuraliManoharAkula/IBMDS-Capstone

Thank you!