



IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

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Outline

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

Executive Summary

- Used the Data Science Life Cycle
 - Business Understanding
 - Data Mining
 - Data Cleaning
 - Data Exploration
 - Feature Engineering
 - Predictive Modeling
 - Data Visualizations
- Summary of all results
 - Used various machine learning models
 - Was able to accurately predict the outcome of most landings

Introduction

- SpaceX advertises Falcon 9 rocket launches cost of 62 million dollars
- Other providers cost upward of 165 million dollars each
- Much of the savings is because SpaceX can reuse the first stage
- 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
- Problems you want to find answers for
 - Success of the first stage landing
 - Cost of launch

Section 1

Methodology

Methodology

Executive Summary

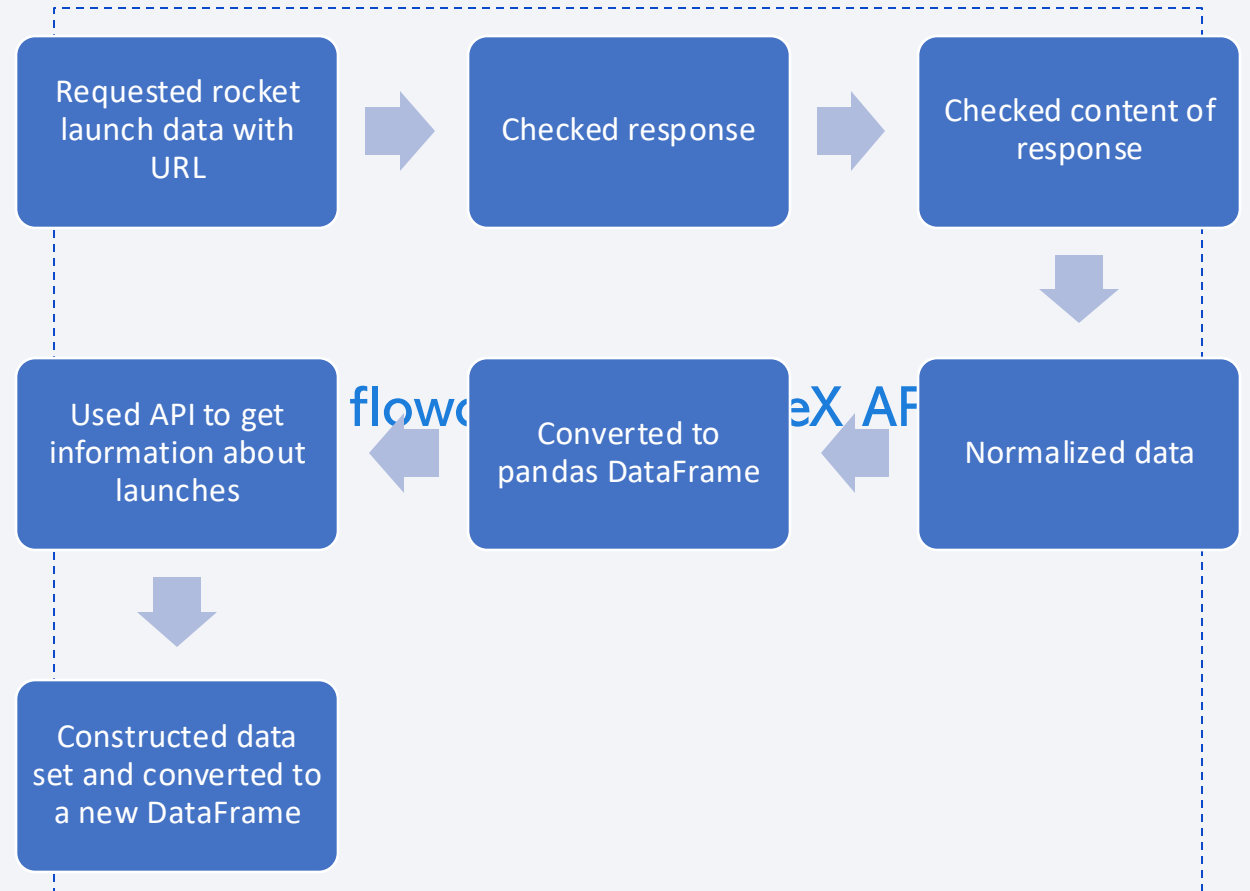
- Data collection methodology:
 - Data was collected utilizing the SpaceX API
- Perform data wrangling
 - Missing data was replaced with mean values
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - How to build, tune, evaluate classification models

Data Collection

- Describe how data sets were collected.
- Data was collected using SpaceX API
 - Collected via get request
- Abstracted the names from various columns
 - Booster Name
 - Rockets
 - Landing Sites
 - Cores

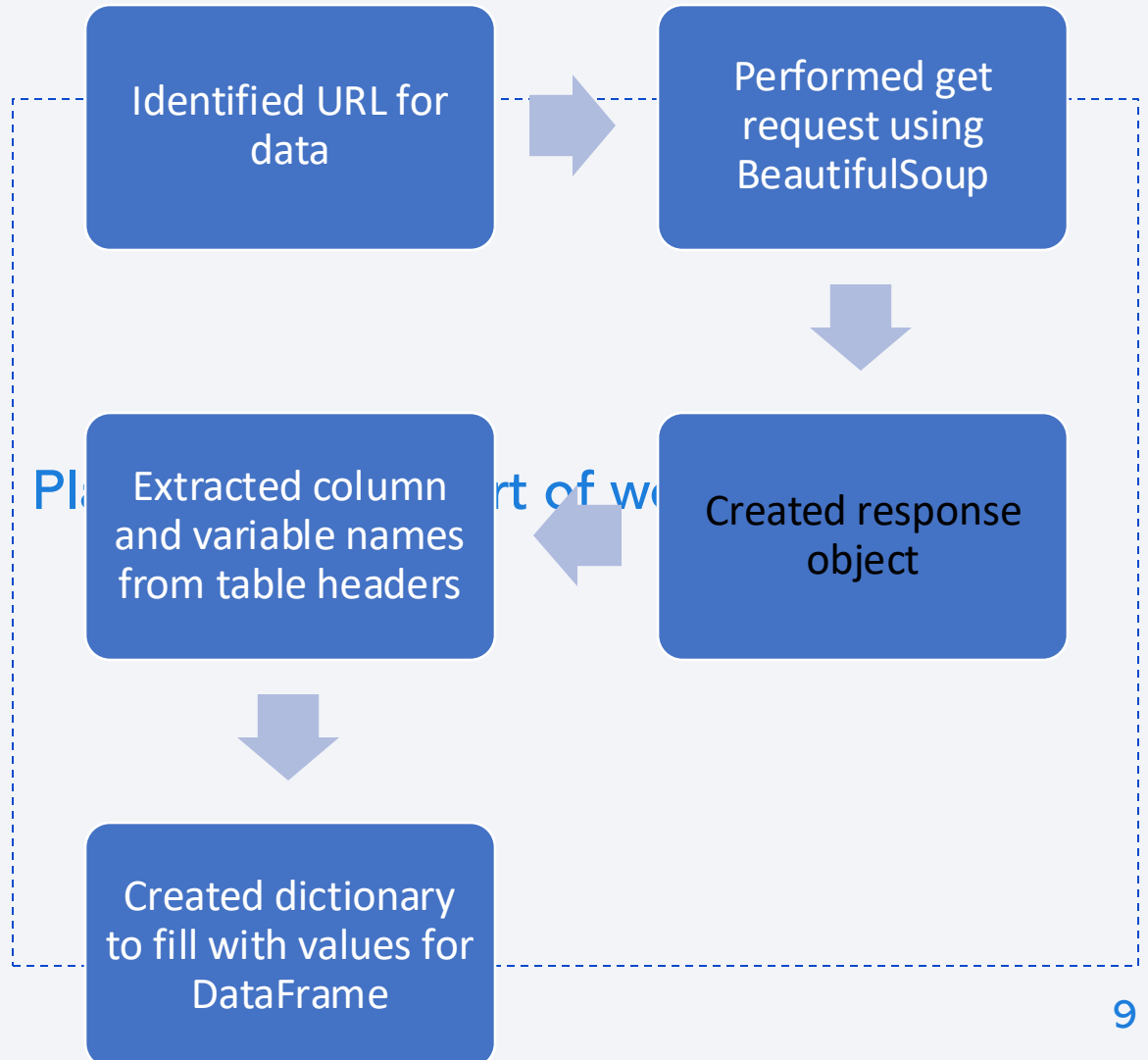
Data Collection – SpaceX API

- URL of Jupyter notebook
 - <https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/80d0a12af259758356d932cb10c8b7c4572f55c5/jupyter-labs-spacex-data-collection-api.ipynb>



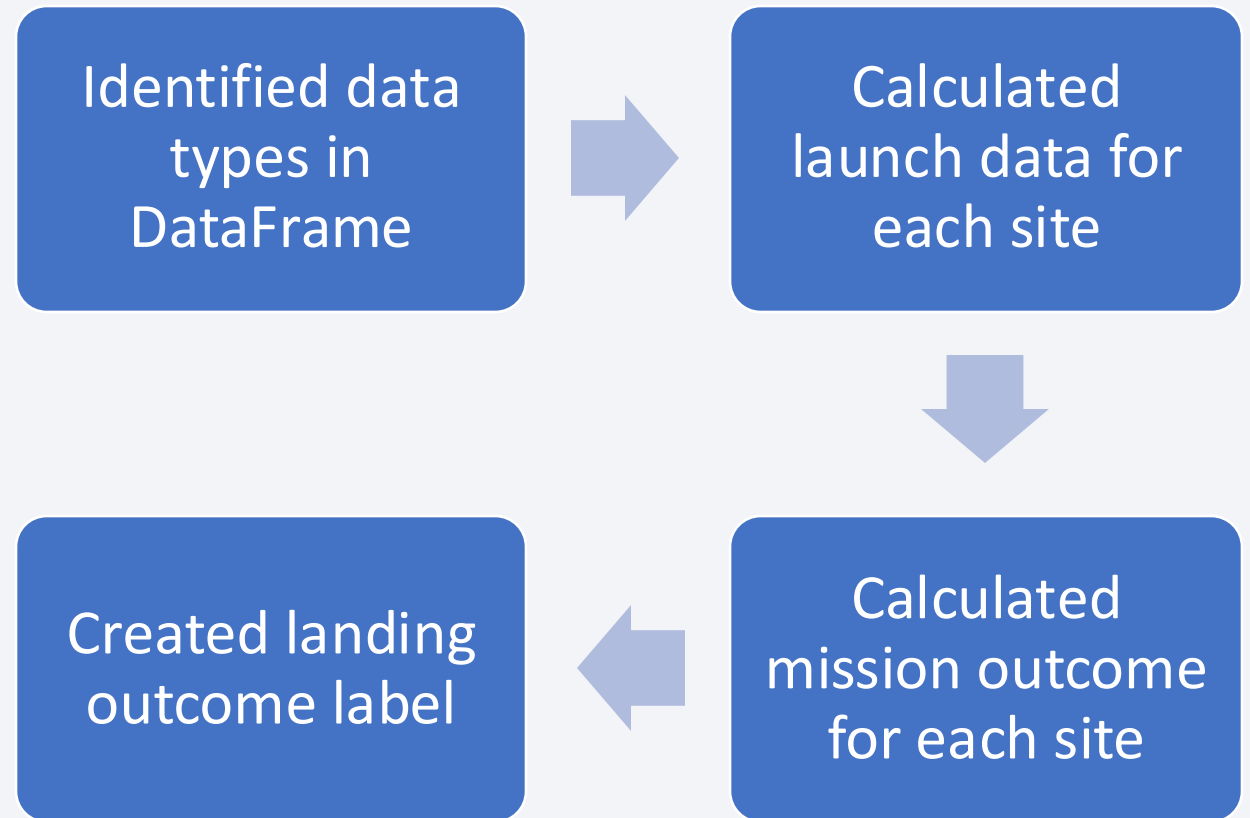
Data Collection - Scraping

- URL of Web Scraping Jupyter notebook
 - <https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/80d0a12af259758356d932cb10c8b7c4572f55c5/jupyter-labs-webscraping.ipynb>



Data Wrangling

- URL of Data Wrangling Jupyter notebook
 - <https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/80d0a12af259758356d932cb10c8b7c4572f55c5/labs-jupyter-spacex-Data%20wrangling.ipynb>



EDA with Data Visualization

- Utilized scatter point charts to compare relationships between features
 - Flight Number vs Launch Site
 - Payload mass vs Launch Site
 - Flight Number vs Orbit Type
 - Payload mass vs Orbit Type
- Plotted Bar chart to compare success rate of each orbit type
- Visualized yearly launch trends with line graph
- Feature engineered relative data and used one hot encoding for categorical values
- Notebook URL - <https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/80d0a12af259758356d932cb10c8b7c4572f55c5/edadataviz.ipynb>

EDA with SQL

- Some SQL queries performed:
 - Display the names of the unique launch sites
 - Display 5 records where launch sites begin with the string 'CCA'
 - Display the total payload mass carried by boosters
 - Display average payload mass
 - List the date when the first successful landing outcome
 - List the total number of successful and failure mission outcomes
- GitHub URL - https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/80d0a12af259758356d932cb10c8b7c4572f55c5/jupyter-labs-eda-sql-coursera_sqlite.ipynb

Build an Interactive Map with Folium

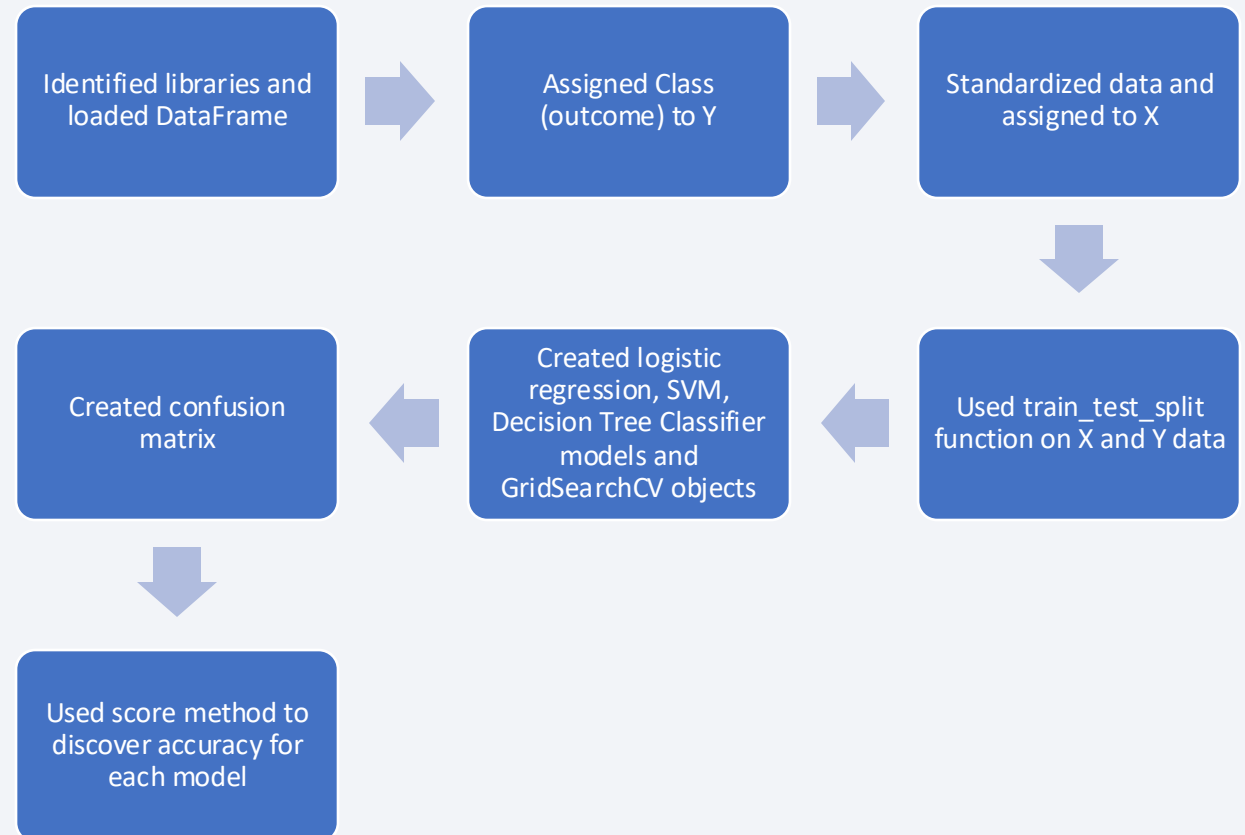
- Used circle to create zoom location
- Added circles for each launch site to see each location
- Added markers to identify names of launch sites
- Added lines to identify distances from landmarks
 - Highways
 - Cities
 - Coastline
- GitHub URL- https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/80d0a12af259758356d932cb10c8b7c4572f55c5/lab_jupyter_launch_site_location.ipynb

Build a Dashboard with Plotly Dash

- Attempted to add dropdown menus, pie chart, etc
- I was unable to. None of the visualizations ever showed (as with the original lab).
- GitHub URL - https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/80d0a12af259758356d932cb10c8b7c4572f55c5/spacex_dash_app.py

Predictive Analysis (Classification)

- Evaluated several models
 - Logistic Regression
 - SVM
 - Decision Tree Classifier
- Fit using GridSearchCV
- SVM had highest accuracy
- GitHub URL - https://github.com/Jhughey24/IBM-Data-Science-Capstone-Project/blob/2eb46b7d7fac1f55e6e83692c67974945dde6b04/SpaceX_Machine%20Learning%20Prediction_Part_5.ipynb



Results

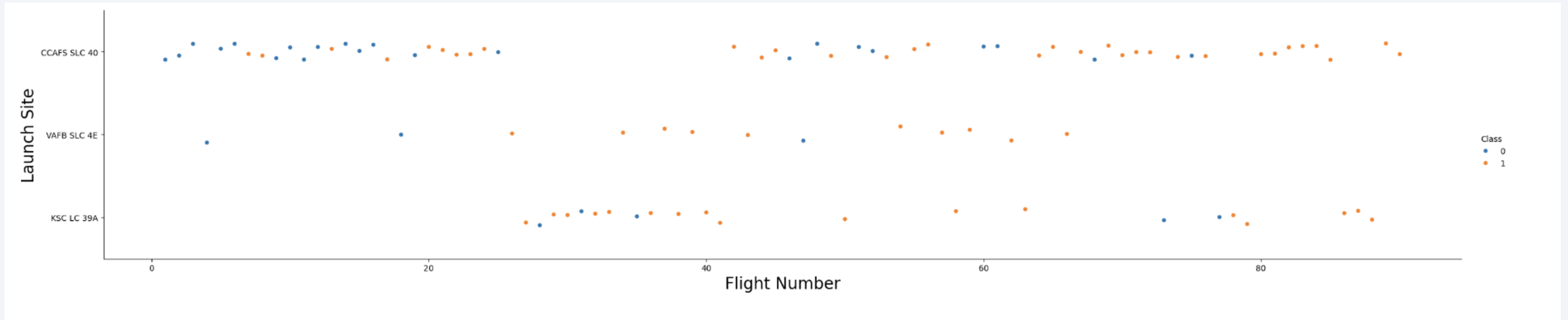
- Exploratory data analysis results
 - Created outcomes for each type of launch
 - Listed outcome with each launch site
- Dash did not work. Do not have the screenshots
- Predictive analysis results show that SVM had the highest accuracy and would allow us to predict the landing outcomes the best.

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

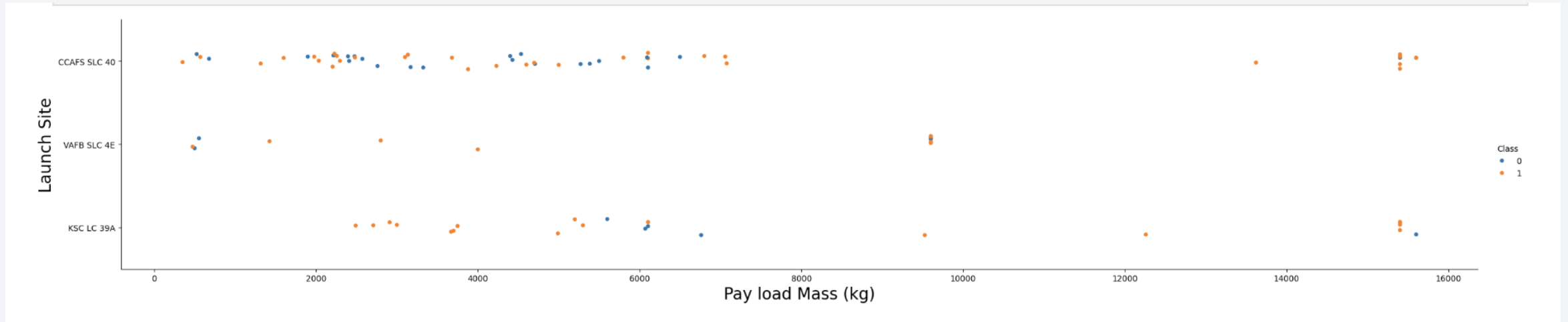
Insights drawn from EDA

Flight Number vs. Launch Site



- Launches became more successful with more flights

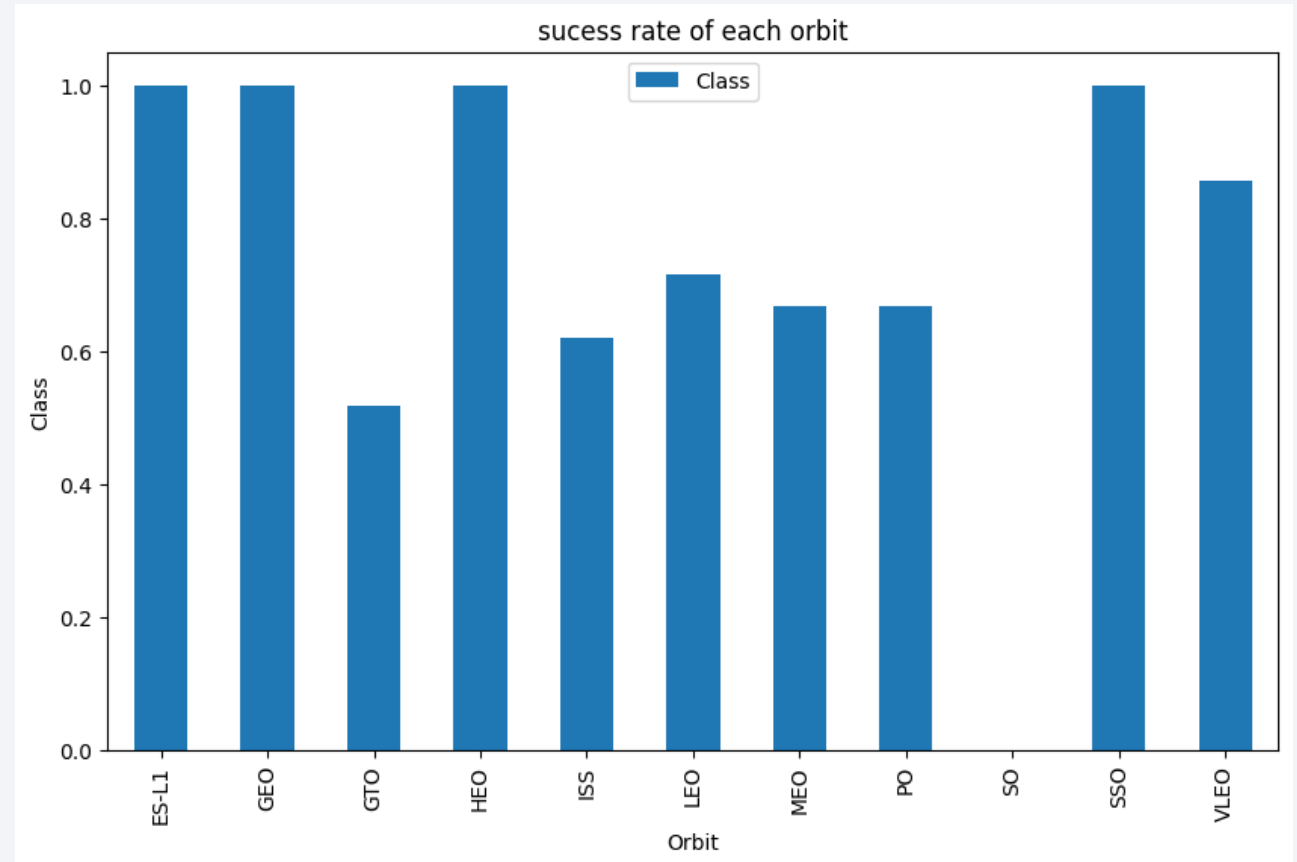
Payload vs. Launch Site



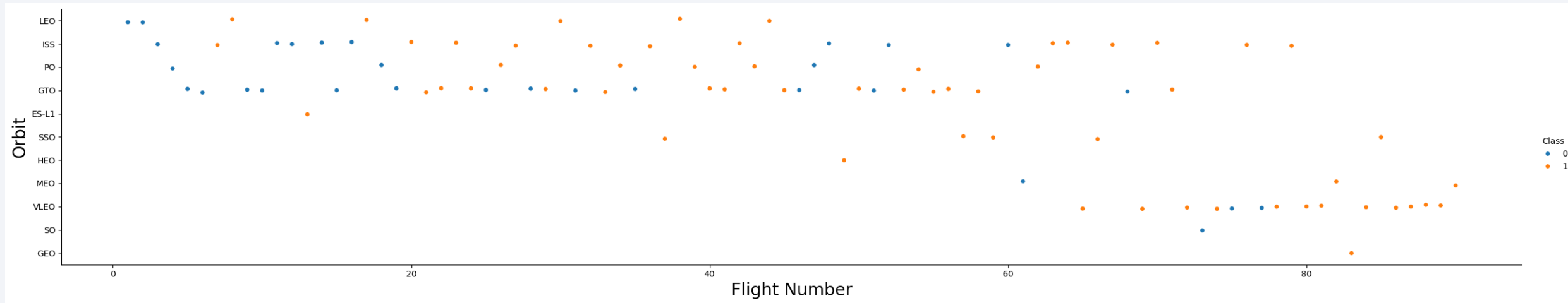
- VAFB-SLC launchsite - no rockets launched for heavypayload mass (greater than 10000)
- KSC LC 39A did better with smalled payloads
- Mixe

Success Rate vs. Orbit Type

- Half orbits had moderate success rates
- The other half was generally successful

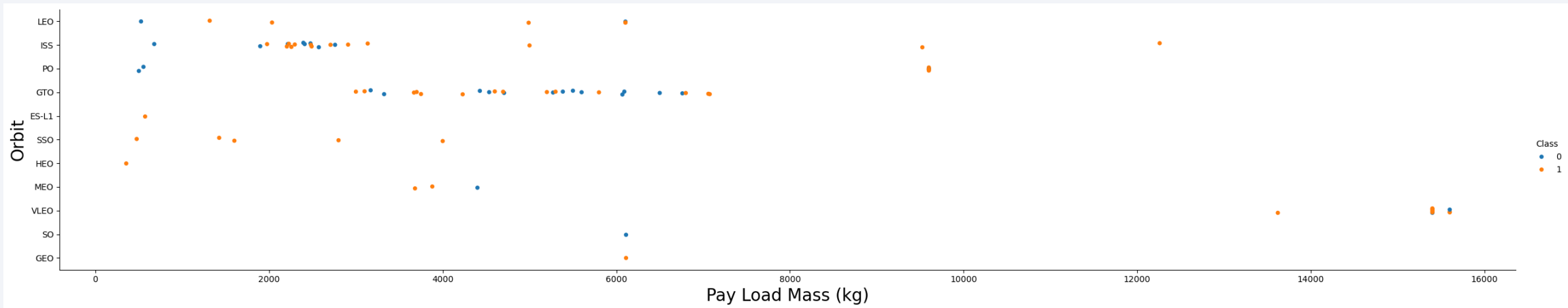


Flight Number vs. Orbit Type



- In the LEO orbit, success seems to be related to the number of flights
- Conversely, in the GTO orbit, there appears to be no relationship between flight number and success.

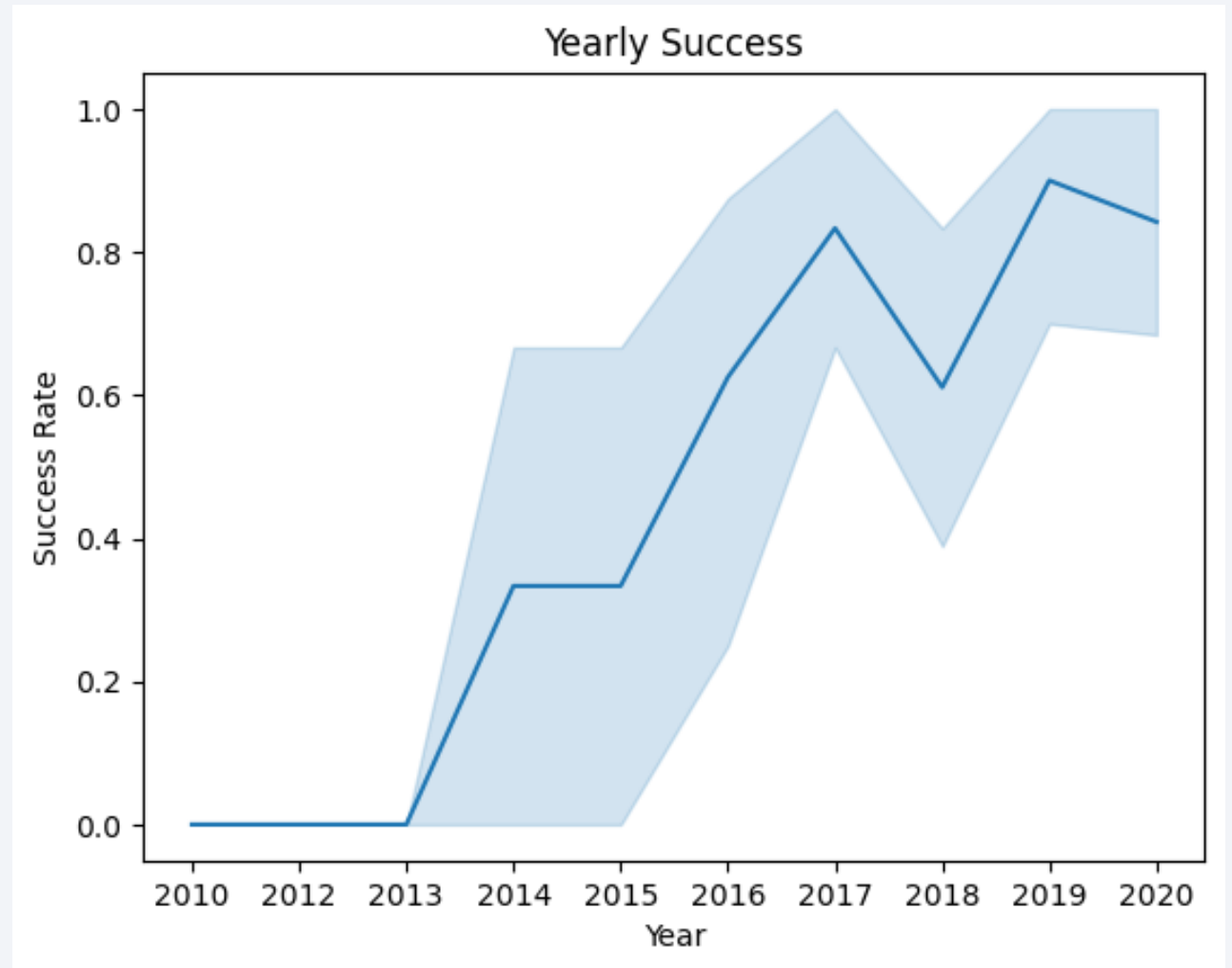
Payload vs. Orbit Type



- With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS
- However, for GTO, it's difficult to distinguish between successful and unsuccessful landings as both outcomes are present.

Launch Success Yearly Trend

- You can observe that the success rate since 2013 kept increasing till 2020



All Launch Site Names

- Names of the unique launch sites
- Listed with Latitude and Longitude coordinates used for plotting

	Launch Site	Lat	Long
0	CCAFS LC-40	28.562302	-80.577356
1	CCAFS SLC-40	28.563197	-80.576820
2	KSC LC-39A	28.573255	-80.646895
3	VAFB SLC-4E	34.632834	-120.610745

Launch Site Names Begin with 'CCA'

- Find 5 records where launch sites begin with `CCA`
- Query used
 - `%sql SELECT "Launch_Site" from SPACEXTBL where "Launch_Site" LIKE 'CCA%' LIMIT 5;`
- Present your query result with a short explanation here
 - I did not select the whole row and missed some the data
 - I only retrieved the launch site - CCAFS LC-40

Total Payload Mass

- Calculate the total payload carried by boosters from NASA
- Query used: `%sql SELECT SUM("PAYLOAD_MASS__KG_") from SPACEXTBL where Customer = 'NASA (CRS)'`
- Result: 45596 KG

Average Payload Mass by F9 v1.1

- Calculate the average payload mass carried by booster version F9 v1.1
- Query used: `%sql SELECT AVG("PAYLOAD_MASS__KG_") from SPACEXTBL where "Booster_Version" = 'F9 v1.1'`
- Result: 2928.4 KG

First Successful Ground Landing Date

- Find the dates of the first successful landing outcome on ground pad
- Query used: `%sql SELECT min(Date) from SPACEXTBL where "Mission_Outcome" = 'Success';`
- Result: 2010-06-04

Successful Drone Ship Landing with Payload between 4000 and 6000

- List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000
- Query used: %sql SELECT "Booster_Version" from SPACEXTBL where "Mission_Outcome" = 'Success' & "PAYLOAD_MASS__KG_" BETWEEN 4000 AND 6000;
- Result:

Total Number of Successful and Failure Mission Outcomes

- Calculate the total number of successful and failure mission outcomes
- Query used: `%sql SELECT (SELECT COUNT("Mission_Outcome") from SPACEXTBL where "Mission_Outcome" like '%Success%') as Success, (SELECT COUNT("Mission_Outcome") from SPACEXTBL where "Mission_Outcome" like '%Failure%') as Failure;`
- Result: Success: 100 – Failure: 1

Boosters Carried Maximum Payload

- List the names of the booster which have carried the maximum payload mass
- Query used: %sql SELECT "Booster_Version" from SPACEXTBL where "PAYLOAD_MASS__KG_" = (SELECT max(PAYLOAD_MASS__KG_) from SPACEXTBL)

Booster_Version

F9 B5 B1048.4

F9 B5 B1049.4

F9 B5 B1051.3

F9 B5 B1056.4

F9 B5 B1048.5

F9 B5 B1051.4

F9 B5 B1049.5

F9 B5 B1060.2

F9 B5 B1058.3

F9 B5 B1051.6

F9 B5 B1060.3

F9 B5 B1049.7

2015 Launch Records

- List the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015

- Query:

%sql

```
SELECT Date, "Landing_Outcome","Booster_Version","Launch_Site" from SPACEXTBL as  
'Failure' where "Landing_Outcome" like '%drone ship%' and substr(Date, 6,2) and  
substr(Date,0,5)='2015'
```

Date	Landing_Outcome	Booster_Version	Launch_Site
2015-01-10	Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
2015-04-14	Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40
2015-06-28	Precluded (drone ship)	F9 v1.1 B1018	CCAFS LC-40

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

- Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order
- Query used: `%%sql`
- SELECT** Date,
"Landing_Outcome" **from**
SPACEXTBL **where** Date
between '2010-06-04'
AND '2017-03-20' **ORDER**
BY DATE **DESC**

Date	Landing_Outcome
2017-03-16	No attempt
2017-02-19	Success (ground pad)
2017-01-14	Success (drone ship)
2016-08-14	Success (drone ship)
2016-07-18	Success (ground pad)
2016-06-15	Failure (drone ship)
2016-05-27	Success (drone ship)
2016-05-06	Success (drone ship)
2016-04-08	Success (drone ship)
2016-03-04	Failure (drone ship)
2016-01-17	Failure (drone ship)
2015-12-22	Success (ground pad)
2015-06-28	Precluded (drone ship)
2015-04-27	No attempt
2015-04-14	Failure (drone ship)

2015-03-02	No attempt
2015-02-11	Controlled (ocean)
2015-01-10	Failure (drone ship)
2014-09-21	Uncontrolled (ocean)
2014-09-07	No attempt
2014-08-05	No attempt
2014-07-14	Controlled (ocean)
2014-04-18	Controlled (ocean)
2014-01-06	No attempt
2013-12-03	No attempt
2013-09-29	Uncontrolled (ocean)
2013-03-01	No attempt
2012-10-08	No attempt
2012-05-22	No attempt
2010-12-08	Failure (parachute)
2010-06-04	Failure (parachute)

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 Screenshots

- Due to an unknown error I was not able to make the visualizations work.
- I completed to code to the best of my ability but nothing showed up on the skeleton dashboard



Section 4

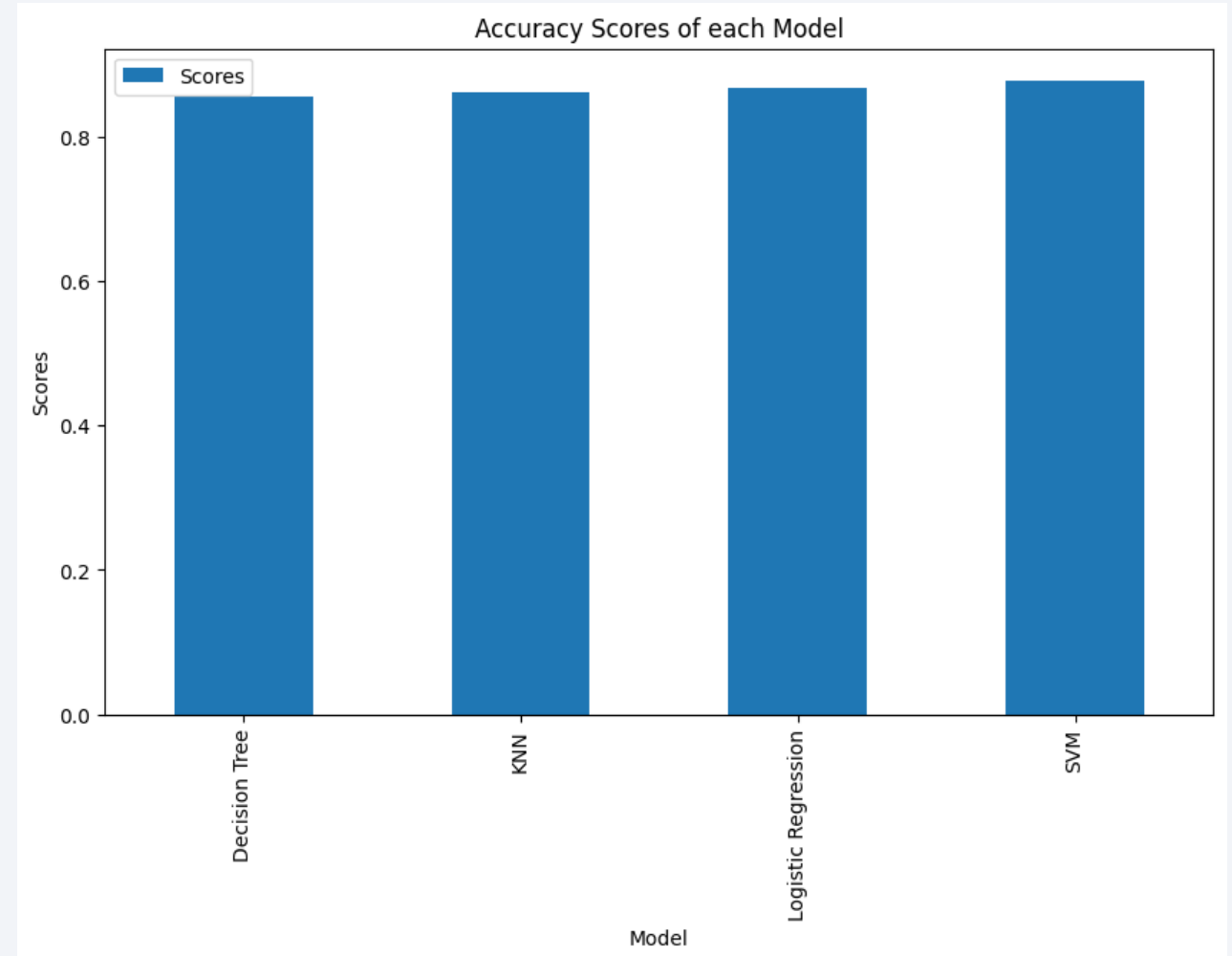
Build a Dashboard with Plotly Dash

Section 5

Predictive Analysis (Classification)

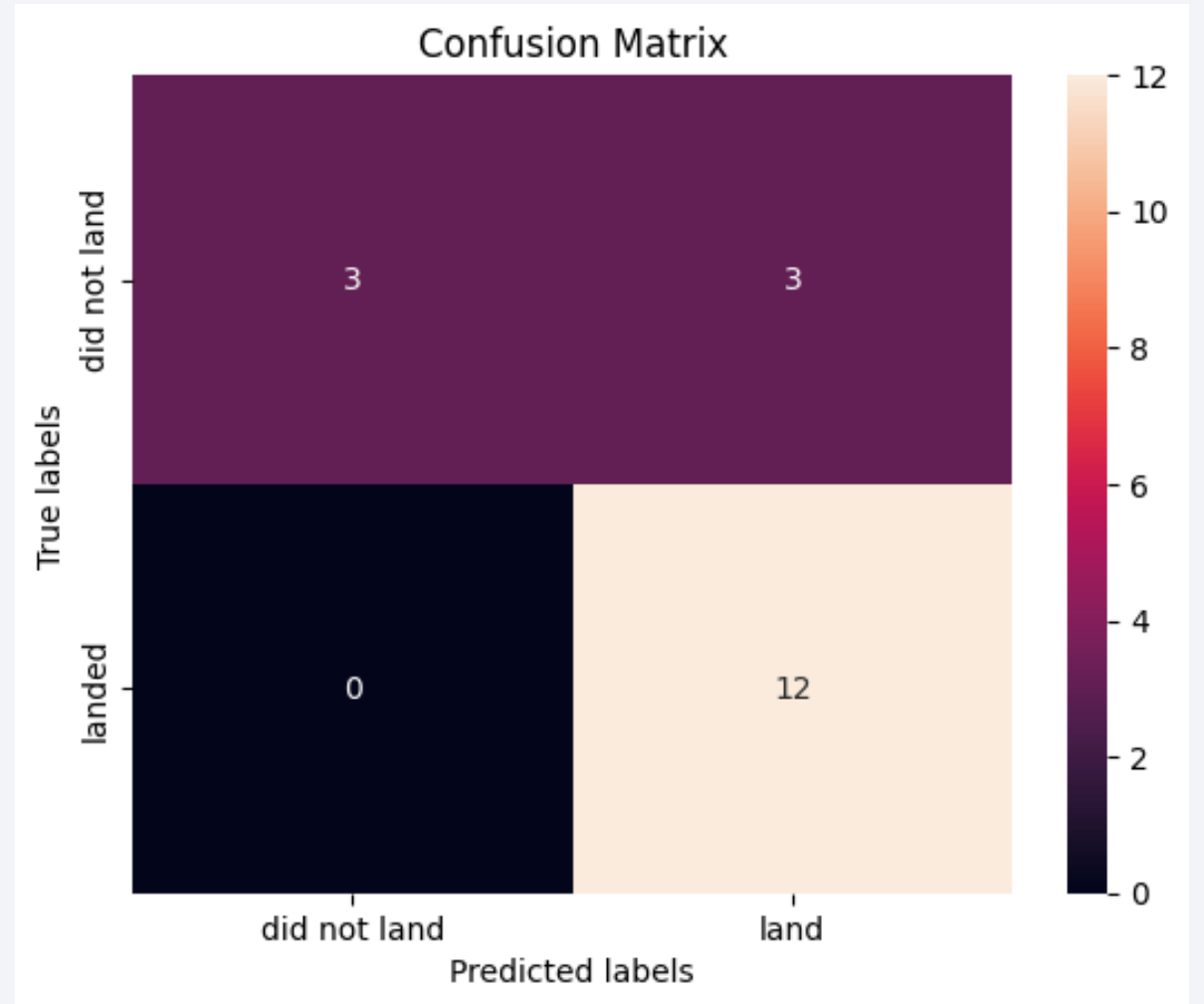
Classification Accuracy

- As observed, the models have fairly the same amount of accuracy
- The SVM model has a slightly higher accuracy score



Confusion Matrix

- Confusion Matrix
- Positively predicted 12 successful landings
- Incorrectly predicted 3 failed landings as successful
- Correctly predicted 3 failed landings
- Did not predict failed landings as successful landings



Conclusions

- Landings generally get better as more flights are launched
- Heavier payloads can lead to failed landings
- Some launch sites are more successful than others
- The best model to predict successful landings is with SVM machine

Appendix

- None

Thank you!

