

Winning Space Race with Data Science

<Yuki Nakayama> <1st,MAR,2024>

in process



Outline





Yuki Nakayama 1st,MAR,2024

Executive Summary



Summary of methodologies

- 1. Data Collection:
- 2. Data Preprocessing:
- 3. Exploratory Data Analysis (EDA):
- 4. Feature Engineering:
- 5. Model Selection:
- 6. Model Training:
- 7. Model Evaluation:

Summary of all results

- •Developed predictive models to forecast the success or failure of future SpaceX launches based on historical data, achieving a high level of accuracy and reliability. Utilized machine learning
- algorithms such as logistic regression and random forests to predict launch outcomes and assess the likelihood of mission success.

Introduction

Project background and context

Space X advertises Falcon 9 rocket launches on its website with a cost of 62 million dollars; other providers cost upward of 165 million dollars each, much of the savings is because Space X can reuse the first stage. Therefore 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 space X for a rocket launch.

Problems you want to find answers

• What factors contribute to the success of landing?

SPACEX







Methodology

Executive Summary

- Data collection methodology:
 - Accessing SpaceX data using the SpaceX API.
 - Retrieving information about launches, rockets, payloads, and other relevant data.
- Perform data wrangling
 - Cleaning and formatting the retrieved data. Handling missing values, duplicates, and outliers. Transforming data into a suitable format for analysis.
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Create classification models with algorithms like logistic regression, decision trees, or random forests, fine-tune parameters, and evaluate accuracy using metrics such as accuracy and F1-score.

Data Collection

Access Data Sources

Data Cleaning and Preprocessing Data Integration

Data Validation

Utilize APIs to access real-time data from SpaceX, ensuring up-to-date information on launches, rockets, and payloads.

Perform data cleaning to remove duplicates, handle missing values, and address inconsistencies in the dataset. Merge and integrate multiple datasets obtained from different sources to create a unified dataset for analysis. Validate the integrity and accuracy of the collected data by cross-referencing with trusted sources and performing data quality checks.

Data Collection – SpaceX API

Now let's start requesting rocket launch data from SpaceX API with the following URL:

[6]: spacex_url="https://api.spacexdata.com/v4/launches/past"

[7]: response = requests.get(spacex_url)

https://github.com/YukiG16/IBM-Data-Science-Certificate/blob/main/Course10_applied-data-science-capstone/Data-science-using-SpaceX-API/week1-1 jupyter-labs-spacex-data-collection-api.jpynb

Request to the SpaceX API

Response content as a Json using <a

Turn it into a Pandas dataframe using .json_normalize()

Filter the dataframe to only include `Falcon 9` launches

Data Wrangling

Dealing with Missing Values

Export it to a CSV

Data Collection - Scraping



https://github.com/YukiG16/IBM-Data-Science-Cer tificate/blob/main/Course10_applied-data-science-capstone/Data-science-using-SpaceX-API/week1-2_jupyter-labs-webscraping.ipynb

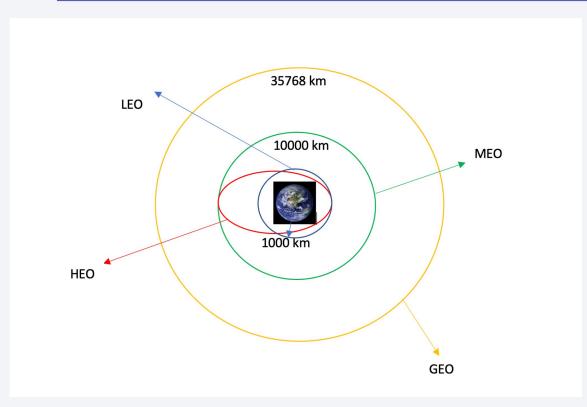
Request the Falcon9 Launch Wiki page from its URL

Create a `BeautifulSoup` object from the HTML `response`

Extract all column/variable names from the HTML table heade

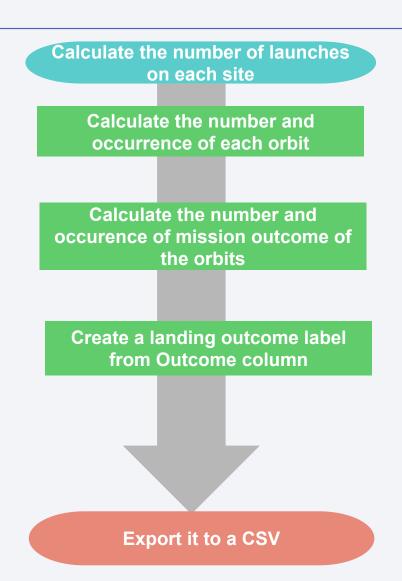
Export it to a CSV

Data Wrangling



Each launch aims to an dedicated orbit

https://github.com/YukiG16/IBM-Data-Science-Certificate/blob/main/Course10_applied-data-science-capstone/Data-science-using-SpaceX-AP l/week1-3_labs-jupyter-spacex-Data%20wrangling.ipynb



10

EDA with Data Visualization

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 dataset includes a record for each payload carried during a SpaceX mission into outer space.

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

- Summarize what map objects such as markers, circles, lines, etc. you created and added to a folium map
- Explain why you added those objects
- Add the GitHub URL of your completed interactive map with Folium map, as an external reference and pee evic or a complete section (e.e.s.)

Build a Dashboard with Plotly Dash

- Summarize what plots/graphs and interactions you have added to a dashboard
- Explain why you added those plots and interactions
- Add the GitHub URL of your completed Plotly Dash lab, as an external reference and peer-review property of the Cess

Predictive Analysis (Classification)

- 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 projects as lysis lab, as an external reference and peer-review purpose

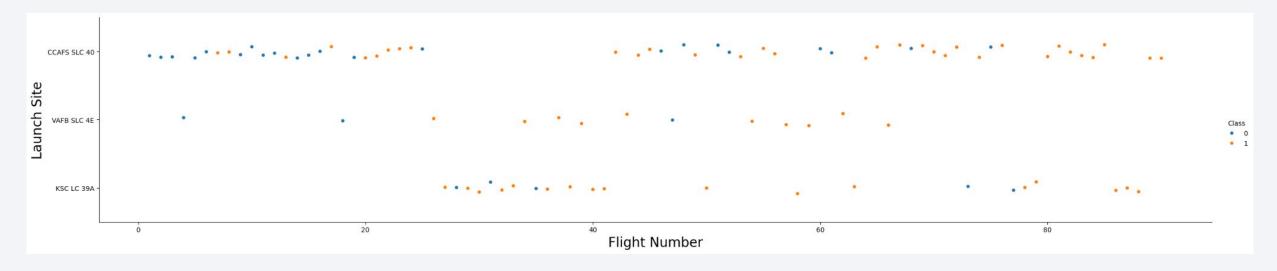
Results

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

in process



Flight Number vs. Launch Site

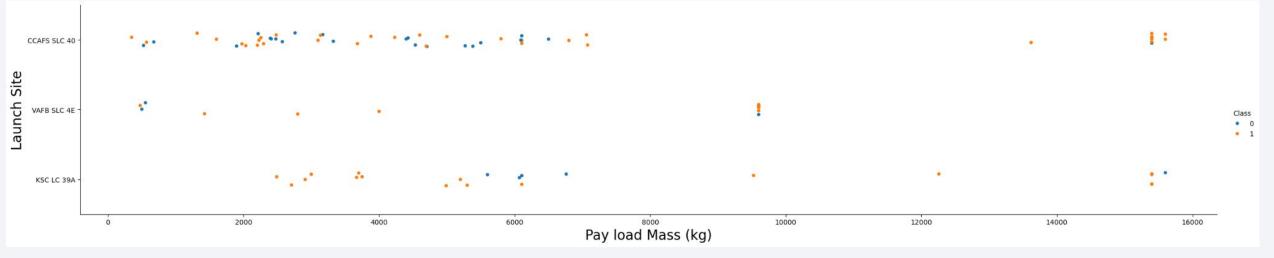


- •:Success
- •:failed

As the flight number increases, the first stage is more likely to land successfully.

This trend seems to be present across all launch sites.

Payload vs. Launch Site

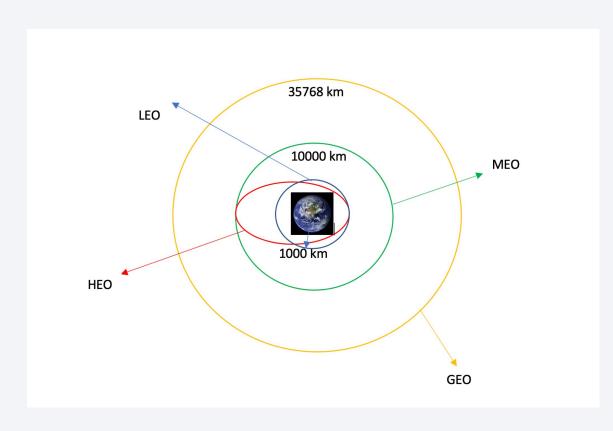


•:Success

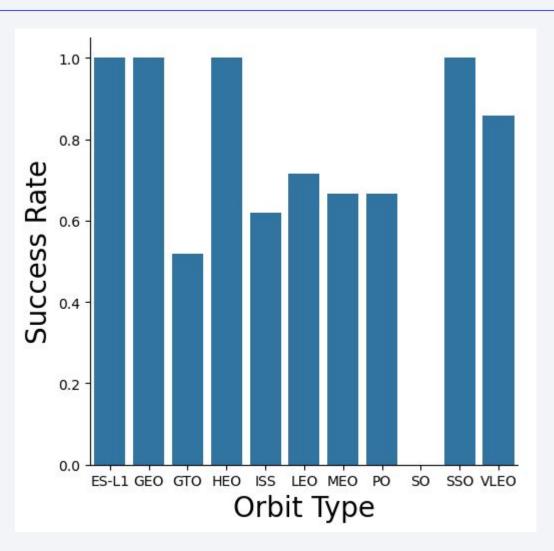
It appears that as the payload increases, so does the flight number.

•:failed

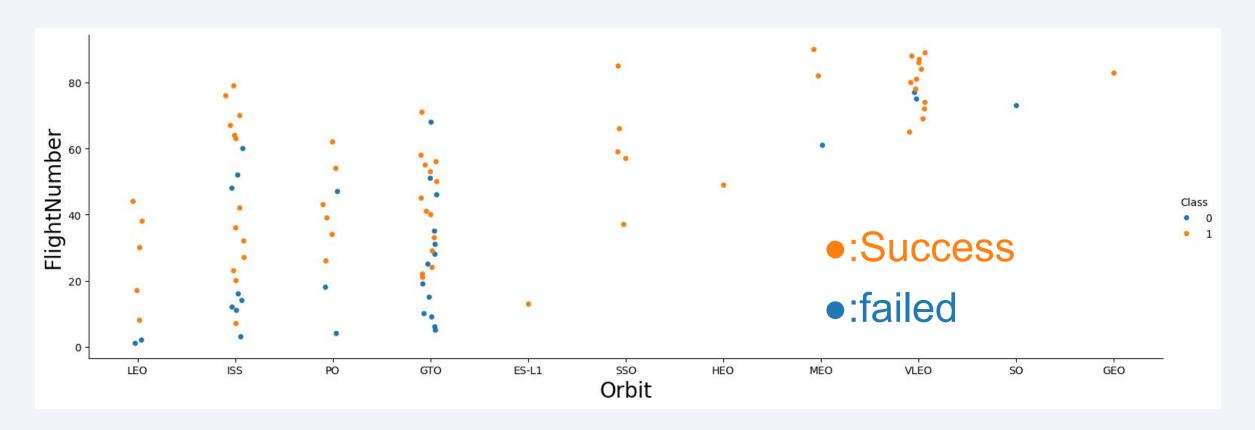
Success Rate vs. Orbit Type



ES-L1, GEO, HEO, and SSO have a high success rate.

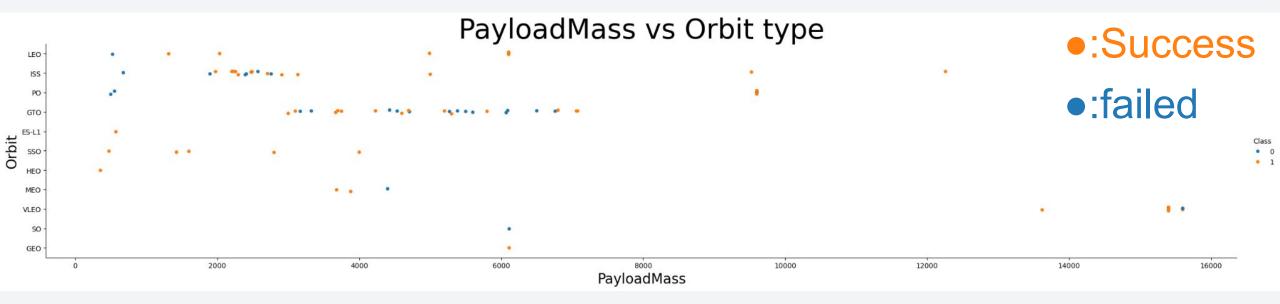


Flight Number vs. Orbit Type



LEO orbit the Success appears related to the number of flights; on the other hand, there seems to be no relationship between flight number when in GTO orbit.

Payload vs. Orbit Type

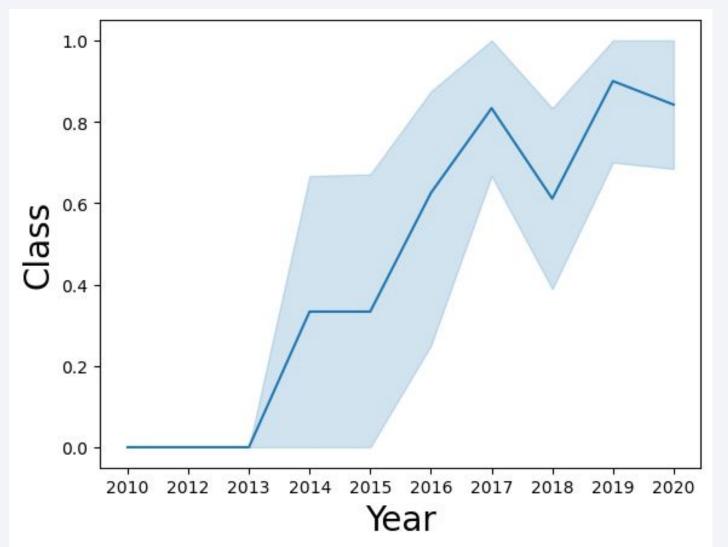


For heavy payloads, it appears that the successful landing rate or positive landing rate is higher for Polar, LEO, and ISS.

However for GTO we cannot distinguish this well as both positive landing rate and negative landing (unsuccessful mission) are both there here.

Launch Success Yearly Trend

It can be noted that the success rate has shown a consistent increase from 2013 to 2020.



All Launch Site Names

Display the names of the unique launch sites in the space mission

%sql select distinct(LAUNCH_SITE) from SPACEXTBL



Launch Site Names Begin with 'CCA'

Display 5 records where launch sites begin with the string 'CCA'										
%sql	select	* from	SPACEXTBL	where	LAUNCH_	SITE	like	"CCA%"	limit	5

Date	Time (UTC)	Booster_Versi on	Launch_Sit e	Payload	PAYLOAD_MAS	S KG_	Orbit	Customer	Mission_Outco me	Landing_Outco me
2010-0 6-04	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0		LEO	SpaceX	Success	Failure (parachute)
2010-1 2-08	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)
2012-0 5-22	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525		LEO (ISS)	NASA (COTS)	Success	No attempt
2012-1 0-08	0:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500		LEO (ISS)	NASA (CRS)	Success	No attempt
2013-0 3-01	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677		LEO (ISS)	NASA (CRS)	Success	No attempt

Total Payload Mass

Display the total payload mass carried by boosters launched by NASA (CRS)

%sql select sum(PAYLOAD_MASS__KG_) from SPACEXTBL where CUSTOMER = 'NASA (CRS)'

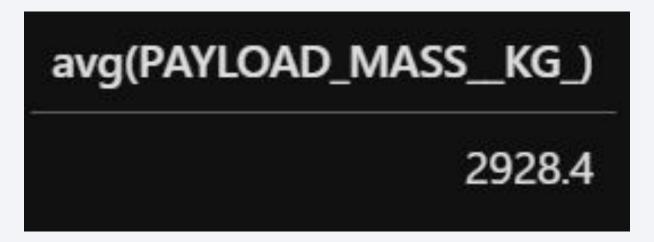
SQL QUERY WITH A SHORT EXPLANATION

sum(PAYLOAD_MASS_KG_)
45596

Average Payload Mass by F9 v1.1

Display average payload mass carried by booster version F9 v1.1

%sql select avg(PAYLOAD_MASS__KG_) from SPACEXTBL where BOOSTER_VERSION = 'F9 v1.1'



First Successful Ground Landing Date

List the date when the first succesful landing outcome in ground pad was acheived.

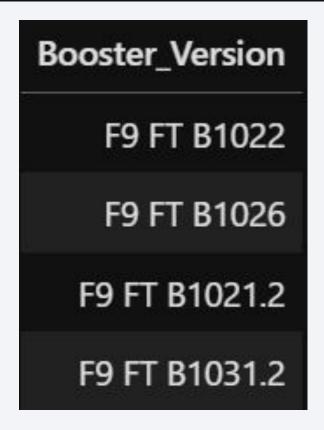
%sql select min(DATE) from SPACEXTBL where LANDING_OUTCOME = 'Success (ground pad)'



Successful Drone Ship Landing with Payload between 4000 and 6000

List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

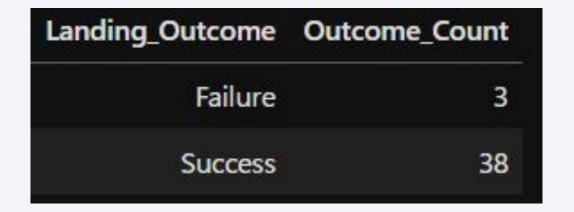
%sql select BOOSTER_VERSION from SPACEXTBL where LANDING_OUTCOME = 'Success (drone ship)' and PAYLOAD_MASS__KG_ between 4000 and 6000



Total Number of Successful and Failure Mission Outcomes

```
List the total number of successful and failure mission outcomes
%sql \
SELECT \
    Landing Outcome, \
    COUNT(*) AS Outcome_Count \
FROM \
    SPACEXTBL \
WHERE \
    Landing Outcome IN ('Success', 'Failure') \
GROUP BY \
    Landing Outcome;
```

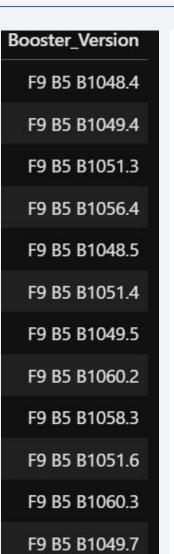
SQL QUERY WITH A SHORT EXPLANATION



Boosters Carried Maximum Payload

List the names of the booster_versions which have carried the maximum payload mass. Use a subquery

%sql SELECT DISTINCT Booster_Version \
FROM SPACEXTBL \
WHERE PAYLOAD_MASS__KG_ = (SELECT MAX(PAYLOAD_MASS__KG_) FROM SPACEXTBL LIMIT 1);





2015 Launch Records

Month	Booster_Version	Launch_Site	Landing_Outcome
01	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
04	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

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

Note: SQLLite does not support monthnames. So you need to use substr(Date, 6,2) as month to get the months and substr(Date, 0,5)='2015' for year.

```
%sql \
SELECT \
    substr(Date, 6, 2) AS Month, \
    Booster_Version, \
    Launch_Site, \
    Landing_Outcome \
FROM \
    SPACEXTBL \
WHERE \
    substr(Date, 0, 5) = '2015' \
    AND Landing_Outcome LIKE 'Failure (drone ship)%';
```

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.
%sql \
SELECT \
    Landing Outcome, \
    COUNT(*) AS Outcome Count \
FROM \
    SPACEXTBL \
WHERE \
    Date BETWEEN '2010-06-04' AND '2017-03-20' \
GROUP BY \
    Landing Outcome \
ORDER BY \
    Outcome Count DESC;
```

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



<Folium Map Screenshot 1>

Replace <Folium map screenshot 1> title with an appropriate title

 Explore the generated folium map and make a proper screenshot to include all launch sites' location markers on a global map

<Folium Map Screenshot 2>

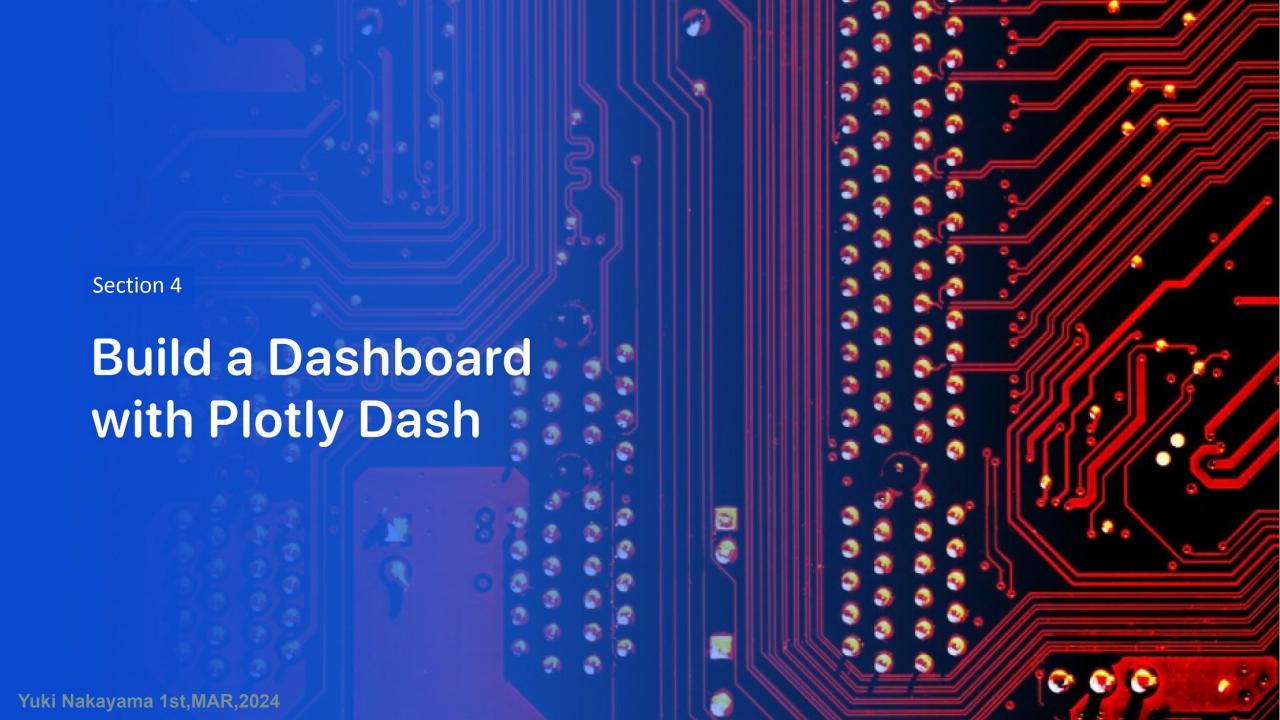
Replace <Folium map screenshot 2> title with an appropriate title

 Explore the folium map and make a proper screenshot to show the color-labeled launch outcomes on the map

<Folium Map Screenshot 3>

Replace <Folium map screenshot 3> title with an appropriate title

 Explore the generated folium map and show the screenshot of a selected launch site to its proximities such as railway, highway, coastline, with distance calculated and displayed



<Dashboard Screenshot 1>

Replace < Dashboard screenshot 1> title with an appropriate title

Show the screenshot of launch success count for all sites, in a piechart

< Dashboard Screenshot 2>

Replace < Dashboard screenshot 2> title with an appropriate title

 Show the screenshot of the piechart for the launch site with highest launch success ratio

<Dashboard Screenshot 3>

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.

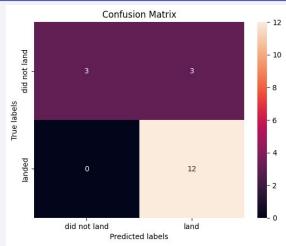


Classification Accuracy

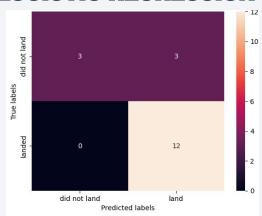
 Visualize the built model accuracy for all built classification models, in a bar chart

 Find which model has the highest classification accuracy

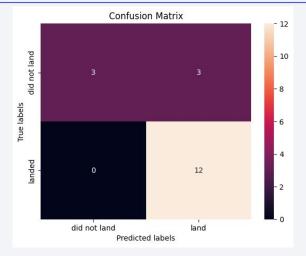
Confusion Matrix



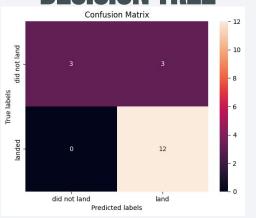
LOGISTIC REGRESSION



SUPPORT VECTOR MACHINE



DECISION TREE



K NEAREST NEIGHBOUR

ML Method	Accuracy Score (%)
Support Vector Machine	83.333333
Logistic Regression	83.333333
K Nearest Neighbour	83.333333
Decision Tree	83.333333

Examining the confusion matrix, we see that logistic regression can distinguish between the different classes. We see that the major problem is false positives.

Conclusions

- Point 1
- Point 2
- Point 3
- Point 4

• ...

Appendix

• Include any relevant assets like Python code snippets, SQL queries, charts, Notebook outputs, or data sets that you may have created during this project

