

Winning Space Race with Data Science

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Outline

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- Introduction
- Methodology
- Results
- Conclusion
- Appendix

Executive Summary

- Introduction
- Summary of methodologies
- Data Collection & Data Wrangling
- EDA & Interactive Visual Analytics Methodology
- Predictive Analysis Methodology
- EDA with Visualization Results
- EDA With SQL Results
- Interactive Map with Folium Results
- Plotly Dashboard Results
- Predictive Analysis (Classification) Results
- Summary of all results

Introduction

Project background and context

SpaceX advertises Falcon 9 rocket launches on its website with a cost of 62 million dollars; other providers cost upwards of 165 million dollars each, much of the savings is because SpaceX can reuse the first stage.

Problems you want to find answers

If we can determine if the first stage will land, we can determine the cost of a launch



Methodology

Executive Summary

- Data collection methodology:
 - SpaceX Rest API
 - Web Scrapping from SpaceX Wikipedia webpage
- Perform data wrangling
 - Hot Encoding data fields
 - Data Cleaning: Correcting and Cleaning null values & irrelevant columns
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Different algorithm such as: Logistic Regression, K-Nearest Neighbors, Support Vector Machine, Decision Tree model have been deployed and evaluated to find the best method.

Data Collection

There are two methods that data are collected:

1

Used SpaceX Rest API Returned SpaceX data Normalized data int a csv file

Data
Consolidation &
Wrangling

2

Got HTML Response from Wikipedia webpage

Used BeautifulSoup to extract Data

Normalized data int a csv file

Data Consolidation & Wrangling

Data Collection – SpaceX API

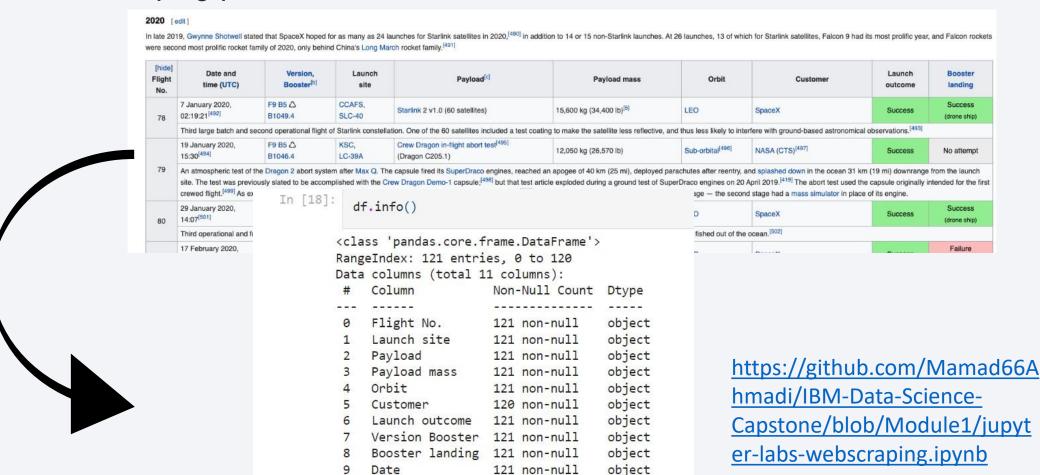
Data collection with SpaceX REST calls

Out[35]:		FlightNumber	Date	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	La
	4	1	2010- 06-04	Falcon 9	NaN	LEO	CCSFS SLC 40	None None	1	False	False	False	
	5	2	2012- 05-22	Falcon 9	525.0	LEO	CCSFS SLC 40	None None	1	False	False	False	
	6	3	2013- 03-01	Falcon 9	677.0	ISS	CCSFS SLC 40	None None	1	False	False	False	
	7	4	2013- 09-29	Falcon 9	500.0	РО	VAFB SLC 4E	False Ocean	1	False	False	False	
	8	5	2013- 12-03	Falcon 9	3170.0	GTO	CCSFS SLC 40	None None	1	False	False	False	
		***	***		***	***	***	***	***	***	***		
	89	86	2020- 09-03	Falcon 9	15600.0	VLEO	KSC LC 39A	True ASDS	2	True	True	True	5e9e3032383ecb6b

^{• &}lt;a href="https://github.com/Mamad66Ahmadi/IBM-Data-Science-Capstone/blob/Module1/jupyter-labs-spacex-data-collection-api.ipynb">https://github.com/Mamad66Ahmadi/IBM-Data-Science-Capstone/blob/Module1/jupyter-labs-spacex-data-collection-api.ipynb

Data Collection - Scraping

Web scraping process



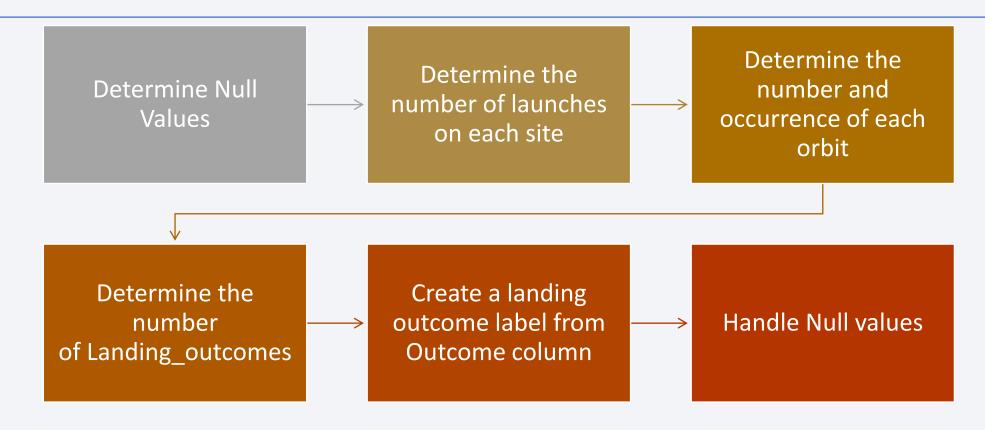
121 non-null

object

Time

dtypes: object(11)
memory usage: 10.5+ KB

Data Wrangling



https://github.com/Mamad66Ahmadi/IBM-Data-Science-Capstone/blob/Module1/labs-jupyter-spacex-data wrangling jupyterlite.jupyter

EDA with Data Visualization

• For machine learning models we should get insight into the relationship between parameters. Therefore, various type of plots such as: Scatter plots, line charts, and bar plots are used to show the relation between different parameters.

Flight Number vs. Payload Mass

Flight Number vs. Launch Site

Payload Mass vs. Launch Site

Orbit vs. Success Rate

Flight Number vs. Orbit

Payload vs Orbit

Success Yearly Trend

https://github.com/Mamad66Ahmadi/IBM-Data-Science-

<u>Capstone/blob/Module2/jupyter-labs-eda-dataviz.ipynb.jupyterlite.ipynb</u>

EDA with SQL

- Displayed the names of the unique launch sites in the space mission
- Displayed 5 records where launch sites begin with the string 'CCA'
- Displayed the total payload mass carried by boosters launched by NASA (CRS)
- Displayed average payload mass carried by booster version F9 v1.1
- Listed the date when the first successful landing outcome in ground pad was acheived.
- Listed the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- Listed the total number of successful and failure mission outcomes
- Listed the names of the booster_versions which have carried the maximum payload mass. Use a subquery
- Listed the records which will display the month names, failure landing_outcomes in drone ship ,booster versions, launch_site for the months in year 2015.
- Ranked the count of successful landing_outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

Build an Interactive Map with Folium

with aim to finding an optimal location for building a launch site:

- Calculated the distances between a launch site to its proximities
- Marked down a point on the closest coastline using MousePosition and calculated the distance between the coastline point and the launch site
- Drew a PolyLine between a launch site to the selected coastline point
- Created a marker with distance to a closest city, railway, highway, etc.
- Drew a line between the marker to the launch site

https://github.com/Mamad66Ahmadi/IBM-Data-Science-Capstone/blob/Module3/lab jupyter launch site location.jupyterlite.ipynb

Build a Dashboard with Plotly Dash

Dashboard with a selectable pie chart and a scatter plot.

- Pie chart shows distribution of successful landings across all launch sites and can be selected to show individual launch site success rates.
- The pie chart is used to visualize launch site success rate.
- Scatter plot takes two inputs: All sites or individual site and payload mass on a slider between 0 and 10000 kg.
- The scatter plot can help us see how success varies across launch sites, payload mass, and booster version category.

https://github.com/Mamad66Ahmadi/IBM-Data-Science-Capstone/blob/Module3/spacex_dash_app.py

Predictive Analysis (Classification)

Standardize the data

Use the function train_test_split to split the data X and Y into training and test data

Create different model objects then create a GridSearchCV

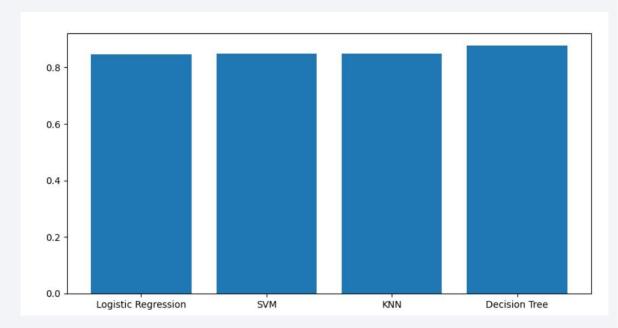
Calculate the accuracy on the test data using the method

Examining the confusion matrix

https://github.com/Mamad66Ahmadi/IBM-Data-Science-Capstone/blob/Module4/SpaceX Machine Learning Prediction Part 5.jupyterlite.ipynb

Results

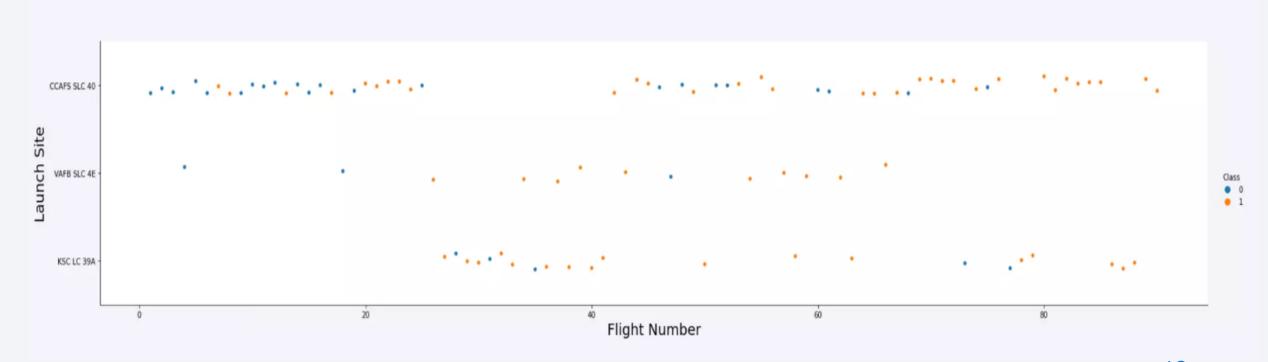
- The SVM, KNN, and Logistic Regression models are the best in terms of prediction accuracy for this dataset.
- Low weighted payloads perform better than the heavier payloads.
- The success rates for SpaceX launches is directly proportional time in years they will eventually perfect the launches.
- KSC LC 39A had the most successful launches from all the sites.
- Orbit GEO, HEO, SSO, ES L1 has the best Success Rate.





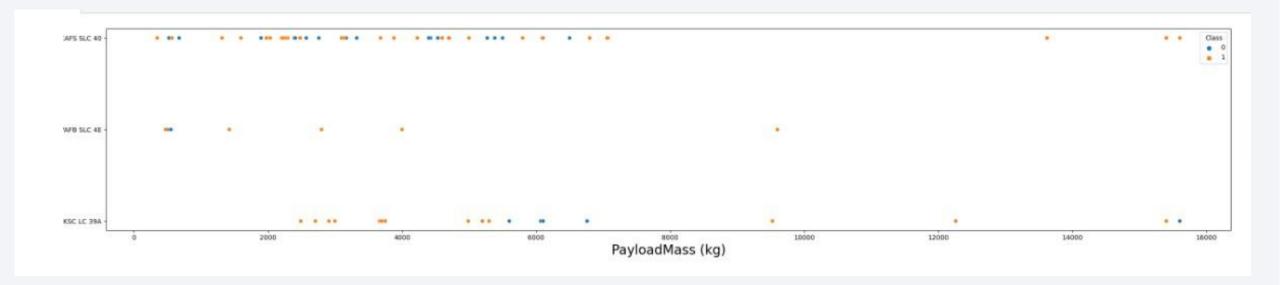
Flight Number vs. Launch Site

 Launches from the site of CCAFS SLC 40 are significantly higher than lunches from other sites



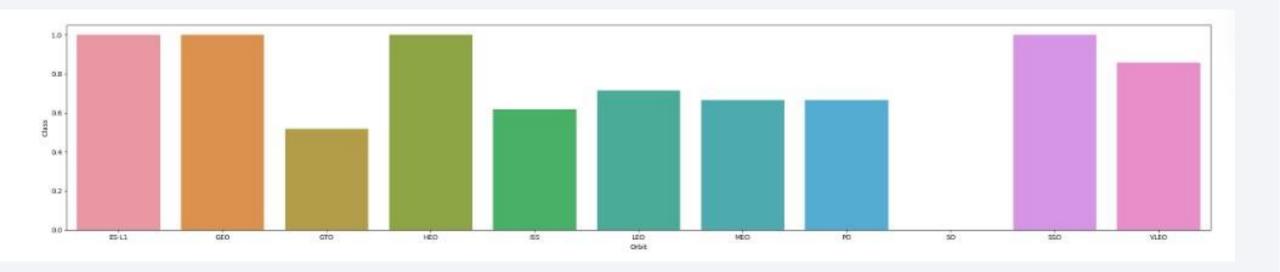
Payload vs. Launch Site

• The majority of Pay Loads with lower Mass have been launched from CCAFS SLC 40.



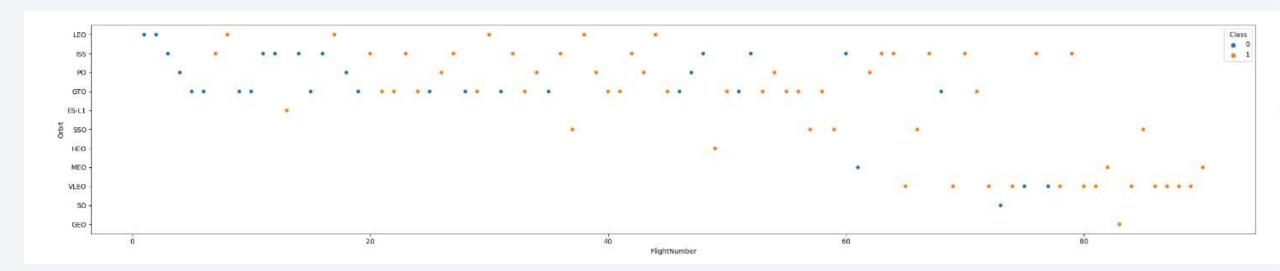
Success Rate vs. Orbit Type

• ESL1, GEO, HEO and SSO orbits had the highest success rate.



Flight Number vs. Orbit Type

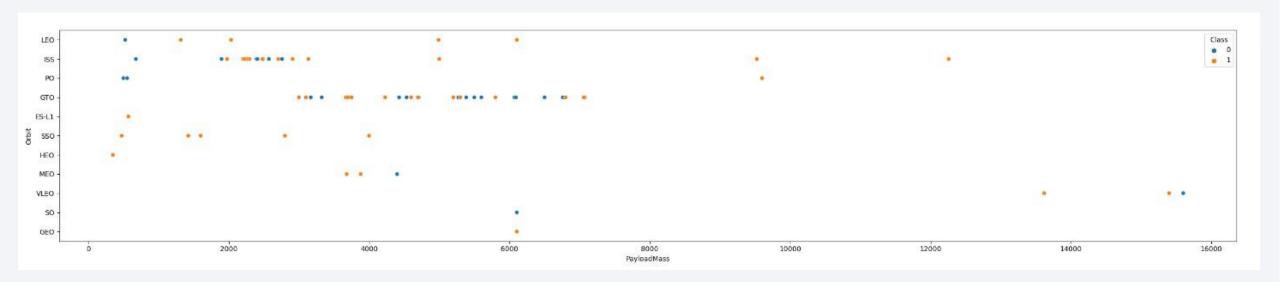
• There is an increasing rate of VLEO launches in recent years.



Payload vs. Orbit Type

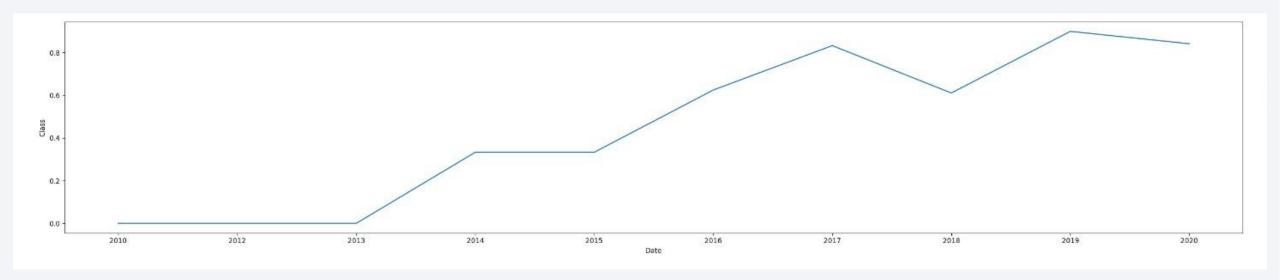
At ISS most of payloads were around 2000

At GTO most of payload were between 3000 and 7000



Launch Success Yearly Trend

• Success rate has increased incredibly in recent years



All Launch Site Names

```
In [7]: %sql select distinct launch_site from SPACEXTBL
        * sqlite:///my_data1.db
       Done.
Out[7]:
          Launch_Site
          CCAFS LC-40
          VAFB SLC-4E
           KSC LC-39A
         CCAFS SLC-40
                None
```

Launch Site Names Begin with 'CCA'

In [14]: %sql select * from SPACEXTBL where launch site like 'CCA%' limit 5 * sqlite:///my data1.db Done. Out[14]: Booster Version Launch Site Payload PAYLOAD MASS KG Orbit Customer Mission Outcome Landing Outc Date Dragon CCAFS LC-Spacecraft 06/04/2010 18:45:00 F9 v1.0 B0003 0.0 LEO SpaceX Success Failure (paracl Qualification Unit Dragon demo flight NASA C1, two CCAFS LC-LEO 12/08/2010 15:43:00 F9 v1.0 B0004 CubeSats, (COTS) Failure (paracl Success (ISS) barrel of NRO Brouere cheese Dragon CCAFS LC-LEO NASA F9 v1.0 B0005 7:44:00 demo flight 525.0 Success 22/05/2012 No atte (ISS) (COTS) CCAFS LC-SpaceX LEO NASA 10/08/2012 0:35:00 F9 v1.0 B0006 500.0 Success No atte CRS-1 (ISS) (CRS) CCAFS LC-SpaceX NASA LEO 03/01/2013 15:10:00 F9 v1.0 B0007 677.0 No atte Success CRS-2 (ISS) (CRS)

Total Payload Mass

Average Payload Mass by F9 v1.1

First Successful Ground Landing Date

Successful Drone Ship Landing with Payload between 4000 and 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

```
In [27]:
          %sql select Mission_Outcome, count(*) from SPACEXTBL group by Mission_Outcome
         * sqlite:///my_data1.db
        Done.
Out[27]:
                     Mission Outcome count(*)
                                            898
                                 None
                        Failure (in flight)
                                             98
                               Success
                               Success
          Success (payload status unclear)
```

Boosters Carried Maximum Payload

```
In [28]:
           %sql select distinct Booster_Version from SPACEXTBL where PAYLOAD_MASS__KG_ = (select max(PAYLOAD_MASS__KG_) from SPACEXTBL
         * sqlite:///my_data1.db
        Done.
Out[28]: 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

%sql select substr(Date, 4, 2) as month, "Landing_Outcome", Booster_Version , Launch_Site from SPACEXTBL where "Landing_Outcome" = 'Failure (drone ship)' and substr(Date,7,4)='2015'

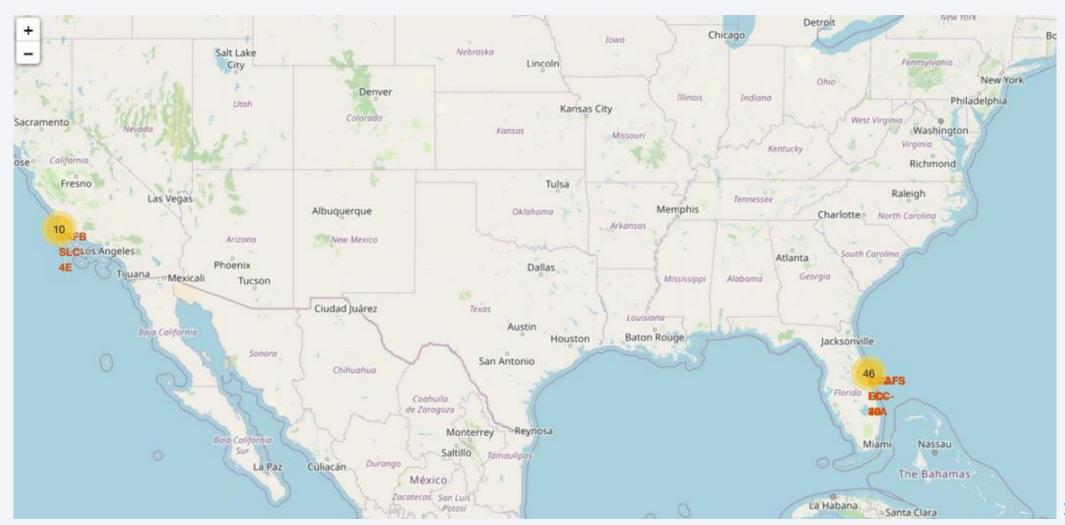
Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

%sql SELECT "Landing_Outcome", count(*) AS count, RANK() OVER (ORDER BY count(*) DESC) AS rank FROM SPACEXTBL WHERE "Landing_Outcome" like 'Success%' and Date between '04-06-2010' and '20-03-2017' GROUP BY "Landing_Outcome" ORDER BY count DESC

% sql SELECT "Landi	ng_Out	come",	count(*) AS count, RANK() OVER (ORDER BY count(*) DESC) AS rank FROM SPACEXTBL WHERE "L
* sqlite:///my_data	1. <mark>d</mark> b		
one. Landing Outcome	count	unnle	
Landing_Outcome	count	rank	
Success	20	1	
Success (drone ship)	8	2	
Success (ground pad)	7	3	



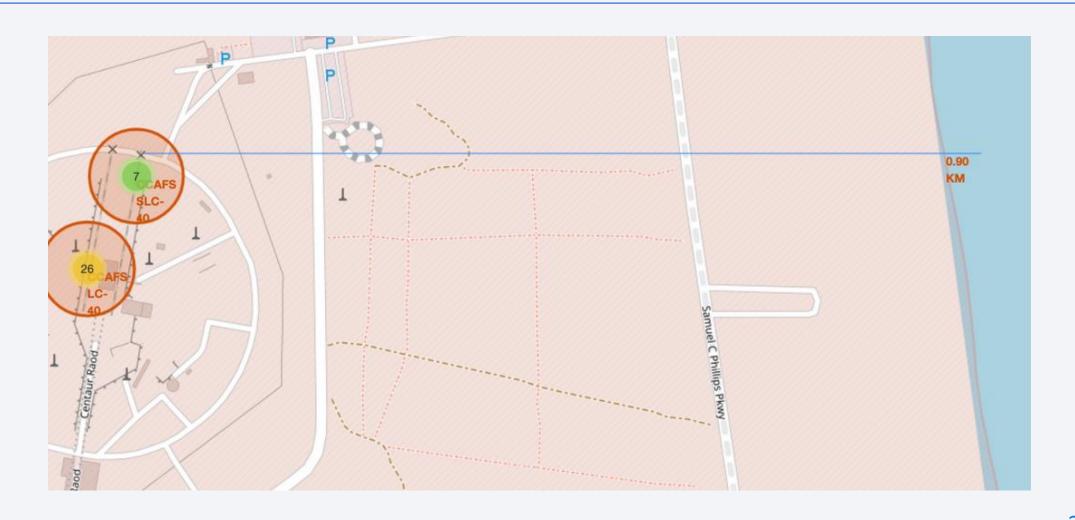
<Folium Map Screenshot 1>



<Folium Map Screenshot 2>

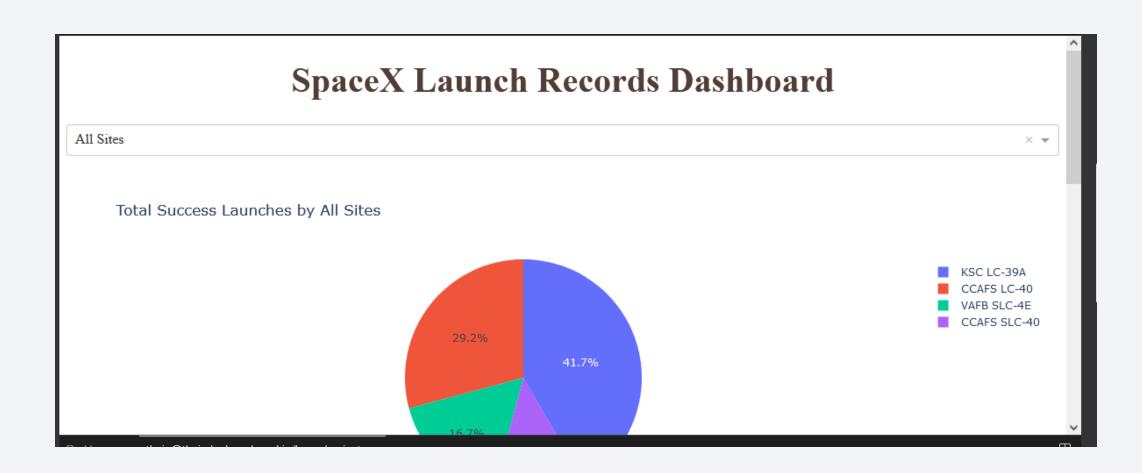


<Folium Map Screenshot 3>





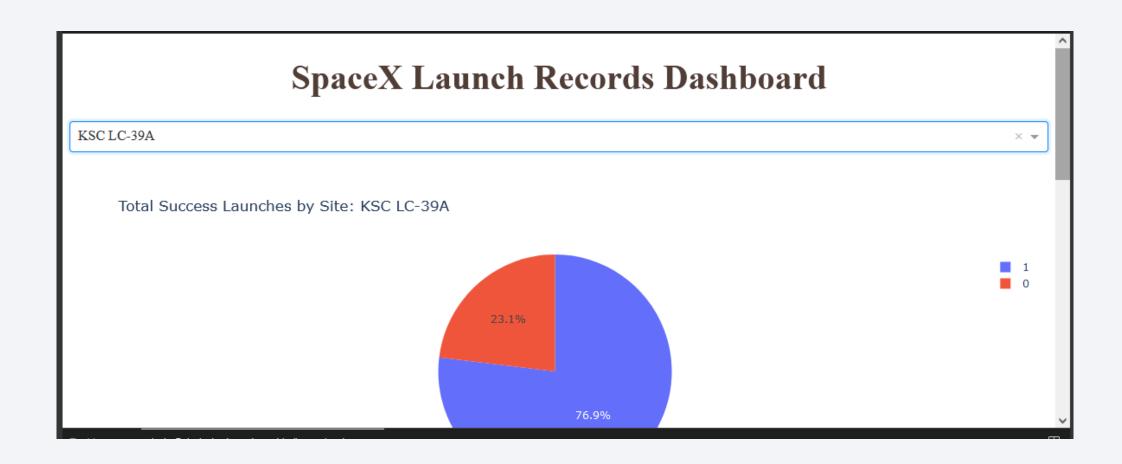
< Dashboard Screenshot 1>



< Dashboard Screenshot 2>

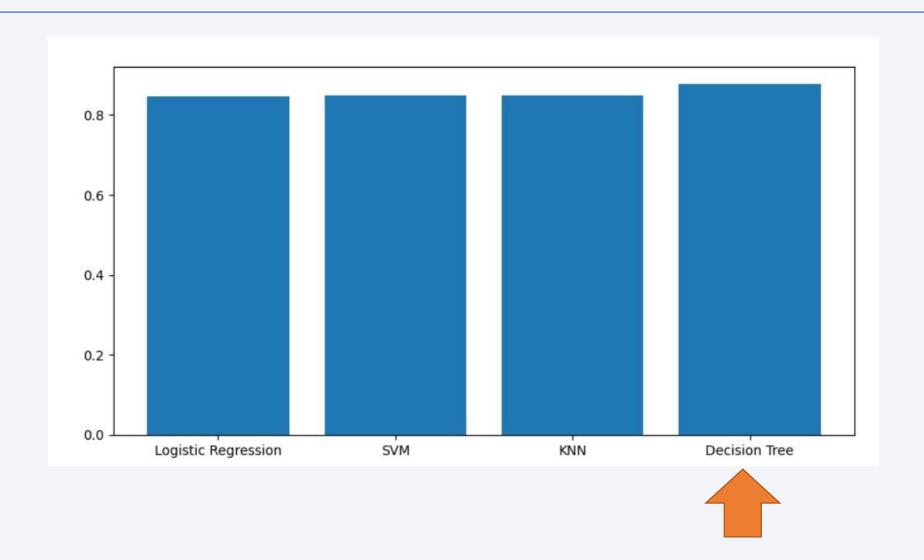


< Dashboard Screenshot 3>

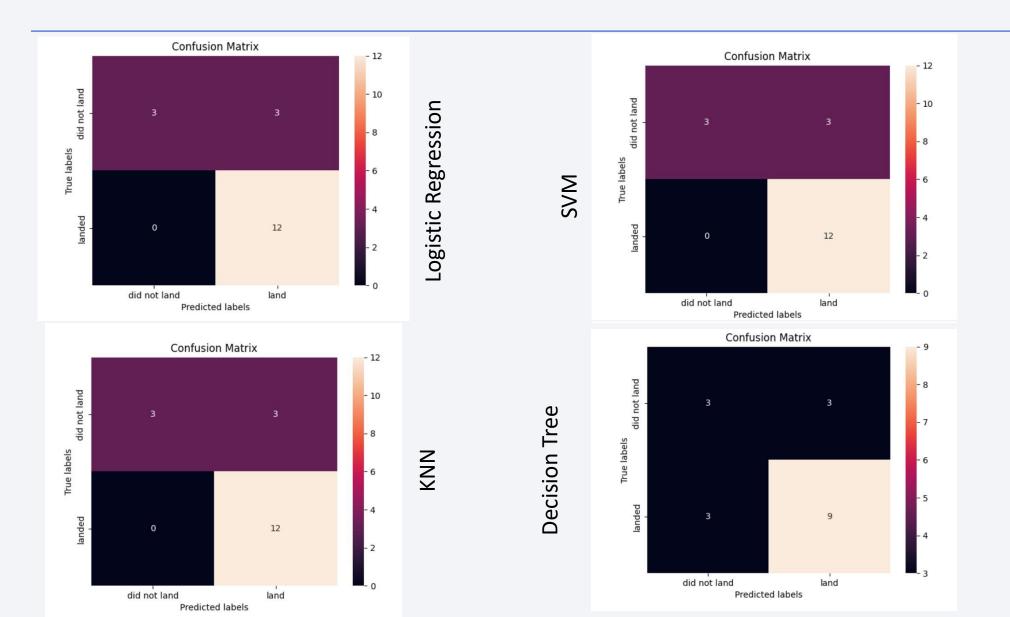




Classification Accuracy



Confusion Matrix



Conclusions

- The Decision Tree model has the best prediction accuracy.
- Low weighted payloads has a better success rate than the heavier payloads.
- The success rates for SpaceX launches have increased significantly in recent years.
- KSC LC 39A had the most successful launches among all other sites.
- Orbit GEO, HEO, SSO, ES L1 had the best Success Rate.

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

