

Winning Space Race with Data Science

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Outline

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

Executive Summary

Summary of methodologies

- Data Collection using web scraping and SpaceX API.
- Exploratory Data Analysis (EDA), including data wrangling, data visualization and interactive visual analytics.
- Building an interactive map with Folium
- Building a Dashboard with Plotly Dash
- Machine Learning Prediction.

Summary of all results

- It was possible to collected valuable data from public sources.
- EDA allowed to identify which features are the best to predict success of launchings.
- Machine Learning Prediction showed the best model to predict which characteristics are important to drive this opportunity by the best way, using all collected data.

Introduction

Project background and context

- SpaceX is the most successful company of the commercial space age, making space travel affordable.
- The company 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 SpaceX can reuse the first stage.
- The objective is to evaluate the viability of the new company Space Y to compete with Space X.
- Therefore, if we can determine if the first stage will land, we can determine the cost of a launch.
- Based on public information and machine learning models, we are going to predict if SpaceX will reuse the first stage.

Problems you want to find answers

- How do variables such as payload mass, launch site, number of flights, and orbits affect the success of the first stage landing?
- Does the rate of successful landings increase over the years?
- What is the best algorithm that can be used for binary classification in this case?



Methodology

Executive Summary

- Data collection methodology:
 - Using SpaceX Rest API
 - Using Web Scrapping from Wikipedia
- Perform data wrangling
 - Filtering the data
 - Dealing with missing values
 - Using One Hot Encoding to prepare the data to a binary classification
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - · Building, tuning and evaluation of classification models to ensure the best results

Data Collection

- Data sets were collected from
- Space X API (https://api.spacexdata.com/v4/rockets/ rockets/)
- Wikipedia
 (https://en.wikipedia.org/wiki/List_of_Falcon/_9/_and_Falcon_Heavy_launches)
 using web scraping.

Data Collection - SpaceX API

- SpaceX offers a public API from where data can be obtained and then used;
- This API was used according to the flowchart beside and then data is persisted.
- GitHub URL of the completed SpaceX API calls notebook https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd814 4689f6ce54bb0daab3c7627ec5df37 7/IBM%20Data%20Science/capstone/ jupyter-labs-spacex-data-collectionapi.ipynb



Data Collection - Scraping

- Data from SpaceX launches can also be obtained from Wikipedia;
- Data are downloaded from Wikipedia according to the flowchart and then persisted.

 GitHub URL of the completed web scraping notebook https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd814468
 9f6ce54bb0daab3c7627ec5df377/IBM%2
 OData%20Science/capstone/jupyter-labswebscraping.ipynb Request the Falcon9 Launch Wiki page



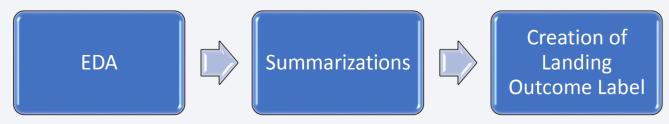
Extract all column/variable names from the HTML table header



Create a data frame by parsing the launch HTML tables

Data Wrangling

- Initially some Exploratory Data Analysis (EDA) was performed on the dataset.
- Then the summaries launches per site, occurrences of each orbit and occurrences of mission outcome per orbit type were calculated.
- Finally, the landing outcome label was created from Outcome column.



• GitHub URL of your completed data wrangling related notebooks https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd8144689f6ce54bb0daab3c7627ec5df377/lBM%20Data%20Science/capstone/labs-jupyter-spacex-Data%20wrangling.ipynb

EDA with Data Visualization

Charts were plotted

- Flight Number vs. Payload Mass, Flight Number vs. Launch Site, Payload Mass vs. Launch Site, Orbit Type vs. Success Rate, Flight Number vs. Orbit Type, Payload Mass vs Orbit Type and Success Rate Yearly Trend
- Scatter plots show the relationship between variables. If a relationship exists, they could be
 used in machine learning model. Bar charts show comparisons among discrete categories. The
 goal is to show the relationship between the specific categories being compared and a
 measured value. Line charts show trends in data over time (time series).
- GitHub URL of your completed EDA with data visualization notebook
 https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd8144689f6ce54bb0daab3c7627ec5df377/IBM%20Data%20Science/capstone/edadataviz.ipynb

EDA with SQL

Performed SQL queries:

- Displaying the names of the unique launch sites in the space mission
- Displaying 5 records where launch sites begin with the string 'CCA'
- Displaying the total payload mass carried by boosters launched by NASA (CRS)
- Displaying average payload mass carried by booster version F9 v1.1
- Listing the date when the first successful landing outcome in ground pad was achieved
- · Listing the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- Listing the total number of successful and failure mission outcomes
- · Listing the names of the booster versions which have carried the maximum payload mass
- Listing the failed landing outcomes in drone ship, their booster versions and launch site names for the months in year 2015
- Ranking 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
- GitHub URL of your completed EDA with SQL notebook https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd8144689f6ce54bb0daab3c7627ec5df377/IBM%20Data%20Science/capstone/jupyter-labs-eda-sql-coursera_sqllite.ipynb

Build an Interactive Map with Folium

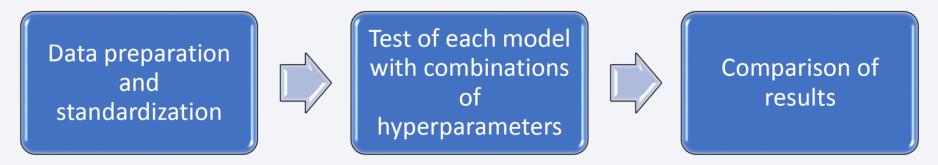
- Markers, circles, lines and marker clusters were used with Folium Maps
 - Markers indicate points like launch sites
 - Circles indicate highlighted areas around specific coordinates, like NASA Johnson Space Center
 - Marker clusters indicates groups of events in each coordinate, like launches in a launch site
 - Lines are used to indicate distances between two coordinates.
- GitHub URL of your completed interactive map with Folium map
 https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd8144689f6ce54bb0daab3c7627ec5df377/IBM
 %20Data%20Science/capstone/lab_jupyter_launch_site_location.ipynb

Build a Dashboard with Plotly Dash

- Launch Sites Dropdown List:
 - Added a dropdown list to enable Launch Site selection.
- Pie Chart showing Success Launches (All Sites/Certain Site):
 - Added a pie chart to show the total successful launches count for all sites and the Success vs. Failed counts for the site, if a specific Launch Site was selected.
- Slider of Payload Mass Range:
 - Added a slider to select Payload range.
- Scatter Chart of Payload Mass vs. Success Rate for the different Booster Versions:
 - Added a scatter chart to show the correlation between Payload and Launch Success.
- GitHub URL of your completed Plotly Dash lab https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd8144689f6ce54bb0daab3c7627ec5df377/IBM%20Data%20Science/capstone/spacex_dash_app.py

Predictive Analysis (Classification)

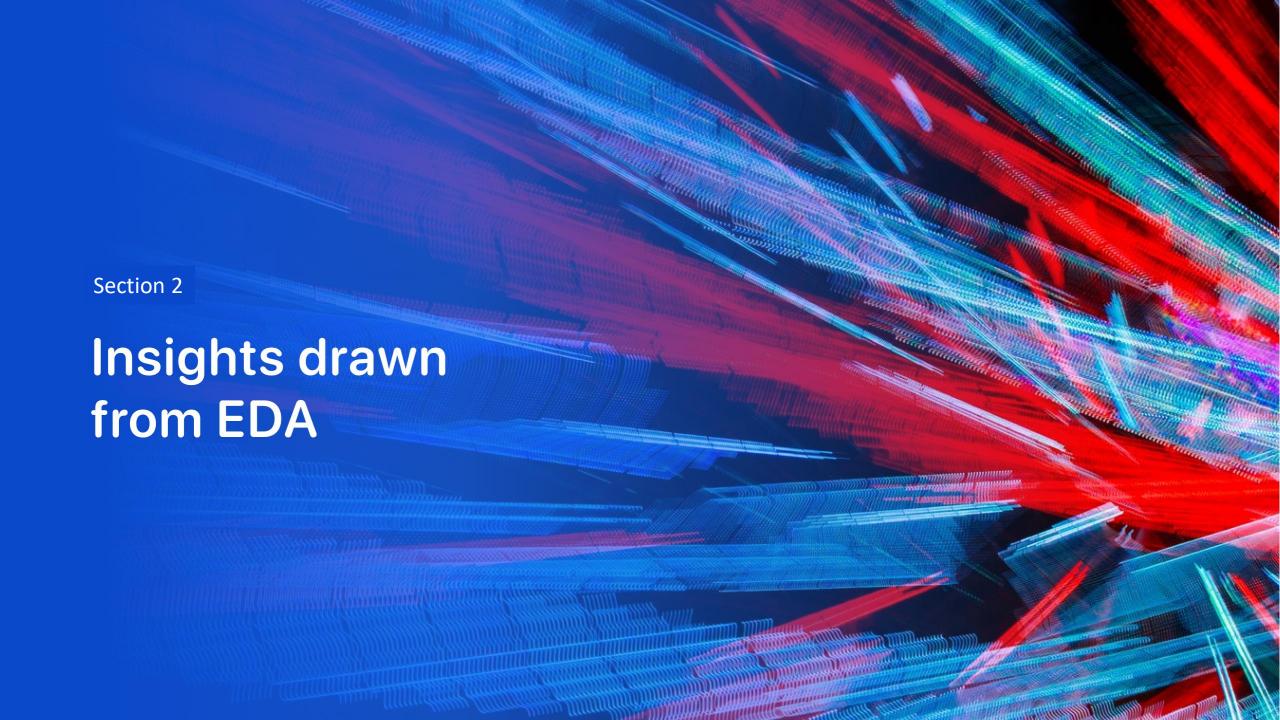
• Four classification models were compared: logistic regression, support vector machine, decision tree and k nearest neighbors.



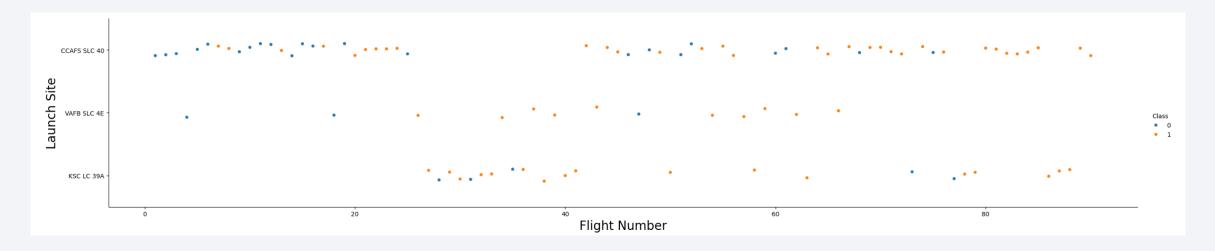
GitHub URL of your completed predictive analysis lab
 https://github.com/Ahmed-Althubyani/Coursera/blob/e70dbd8144689f6ce54bb0daab3c7627ec5df377/lBM%20Data%20Science/capstone/SpaceX_Machine%20Learning%20Prediction_Part_5.ipynb

Results

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

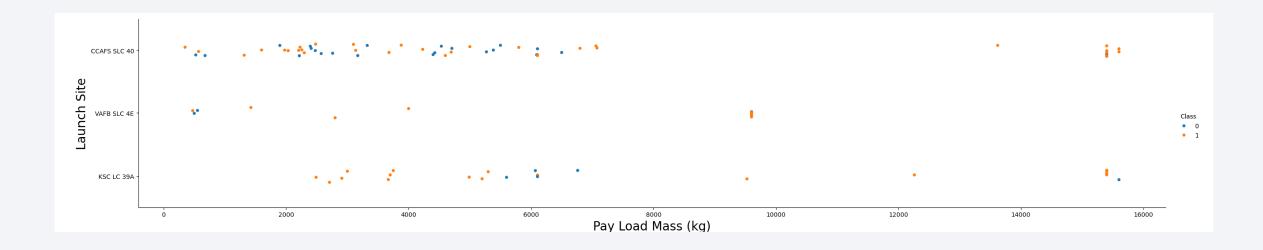


Flight Number vs. Launch Site



- The earliest flights all failed while the latest flights all succeeded.
- The CCAFS SLC 40 launch site has about a half of all launches.
- VAFB SLC 4E and KSC LC 39A have higher success rates.
- It can be assumed that each new launch has a higher rate of success.

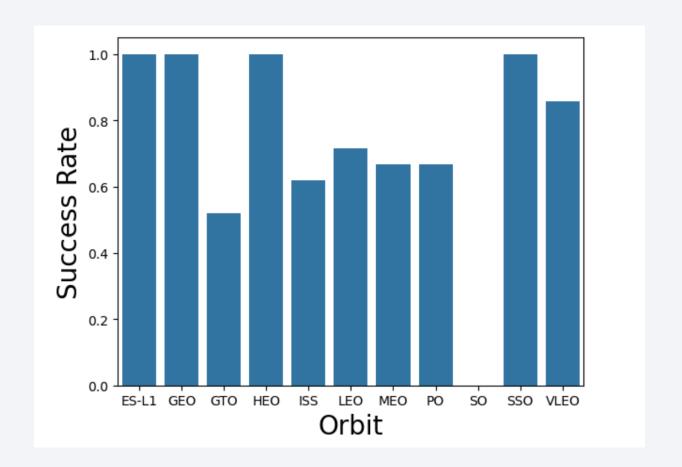
Payload vs. Launch Site



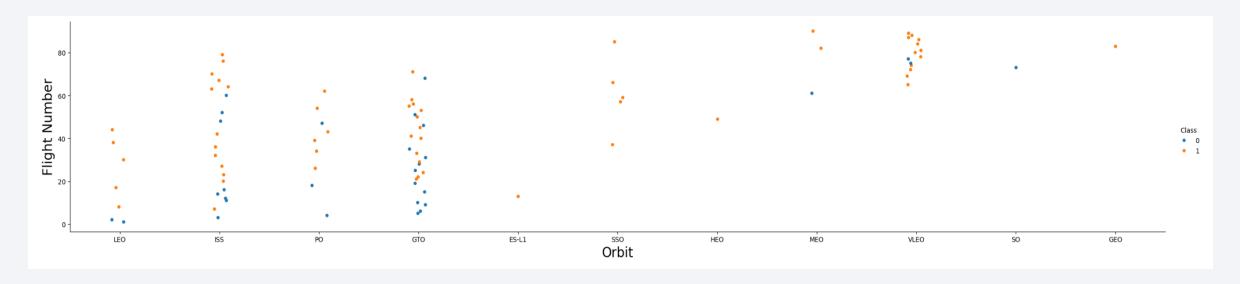
- Payloads over 9,000kg (about the weight of a school bus) have excellent success rate;
- Payloads over 12,000kg seems to be possible only on CCAFS SLC 40 and KSC LC 39A launch sites.

Success Rate vs. Orbit Type

- Orbits with 100% success rate:
 - ES-L1, GEO, HEO, SSO
- Orbits with 0% success rate:
 - SO
- Orbits with success rate between 50% and 85%:
 - GTO, ISS, LEO, MEO, PO

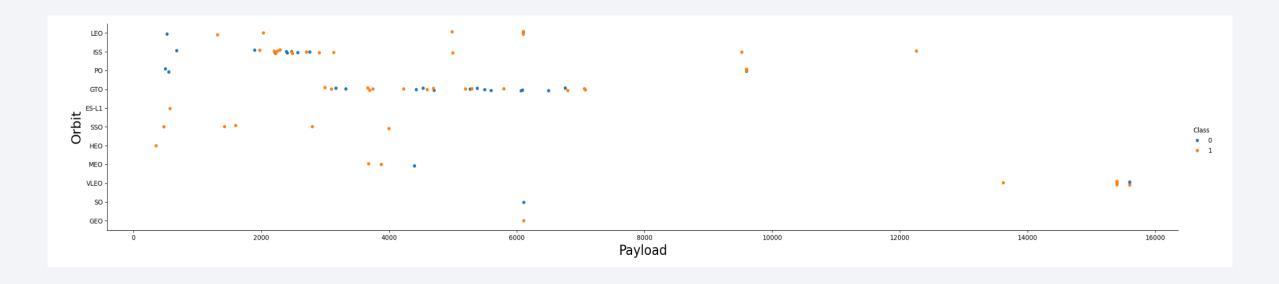


Flight Number vs. Orbit Type



- Apparently, success rate improved over time to all orbits;
- VLEO orbit seems a new business opportunity, due to recent increase of its frequency.

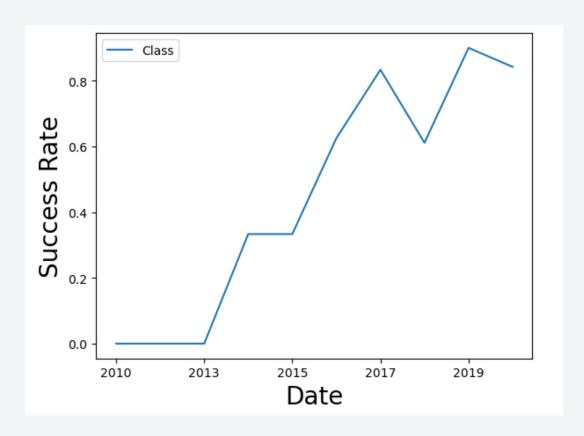
Payload vs. Orbit Type



- Apparently, there is no relation between payload and success rate to orbit GTO;
- ISS orbit has the widest range of payload and a good rate of success;
- There are few launches to the orbits SO and GEO.

Launch Success Yearly Trend

• The success rate since 2013 kept increasing till 2020.



All Launch Site Names

• Displaying the names of the unique launch sites in the space mission.

Launch_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

Launch Site Names Begin with 'CCA'

• 5 records where launch sites begin 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	7:44:00	F9 v1.0 B0005	CCAFS LC- 40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
2012- 10-08	0:35:00	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

Calculate the total payload carried by boosters from NASA

SUM(PAYLOAD_MASS__KG_)

45596

Average Payload Mass by F9 v1.1

Calculate the average payload mass carried by booster version F9 v1.1

AVG(PAYLOAD_MASS__KG_)

2928.4

First Successful Ground Landing Date

• Find the dates of the first successful landing outcome on ground pad

min(Date)

2015-12-22

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



Total Number of Successful and Failure Mission Outcomes

• Calculate the total number of successful and failure mission outcomes

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

Boosters Carried Maximum Payload

• List the names of the booster which have carried the maximum payload mass

ter_	Version
B5	B1048.4
B5	B1049.4
B5	B1051.3
B5	B1056.4
B5	B1048.5
B5	B1051.4
В5	B1049.5
B5	B1060.2
В5	B1058.3
B5	B1051.6
B5	B1060.3
B5	B1049.7
	B5 B5 B5 B5 B5 B5 B5 B5 B5

2015 Launch Records

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

MONTH	DATE	booster_version	launch_site	landing_outcome
January	2015-01-10	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
April	2015-04-14	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

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

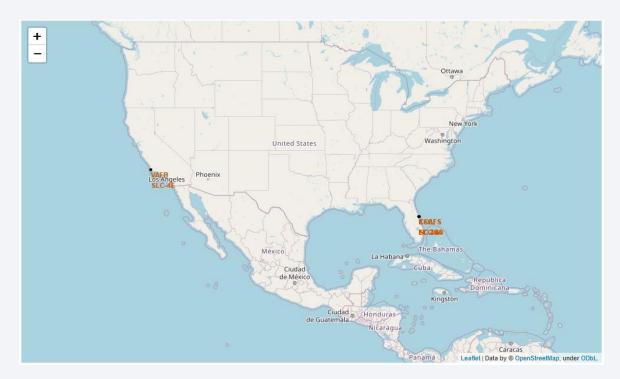
 Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order

landing_outcome	count_outcomes
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



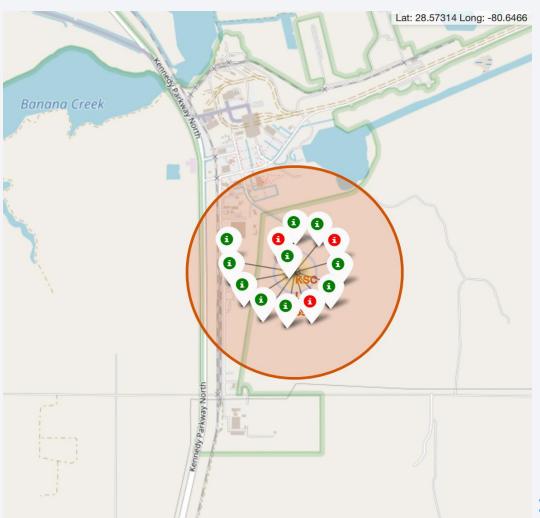
All launch sites

• Launch sites are near sea, probably by safety, but not too far from roads and railroads.



Color-labeled launch records on the map

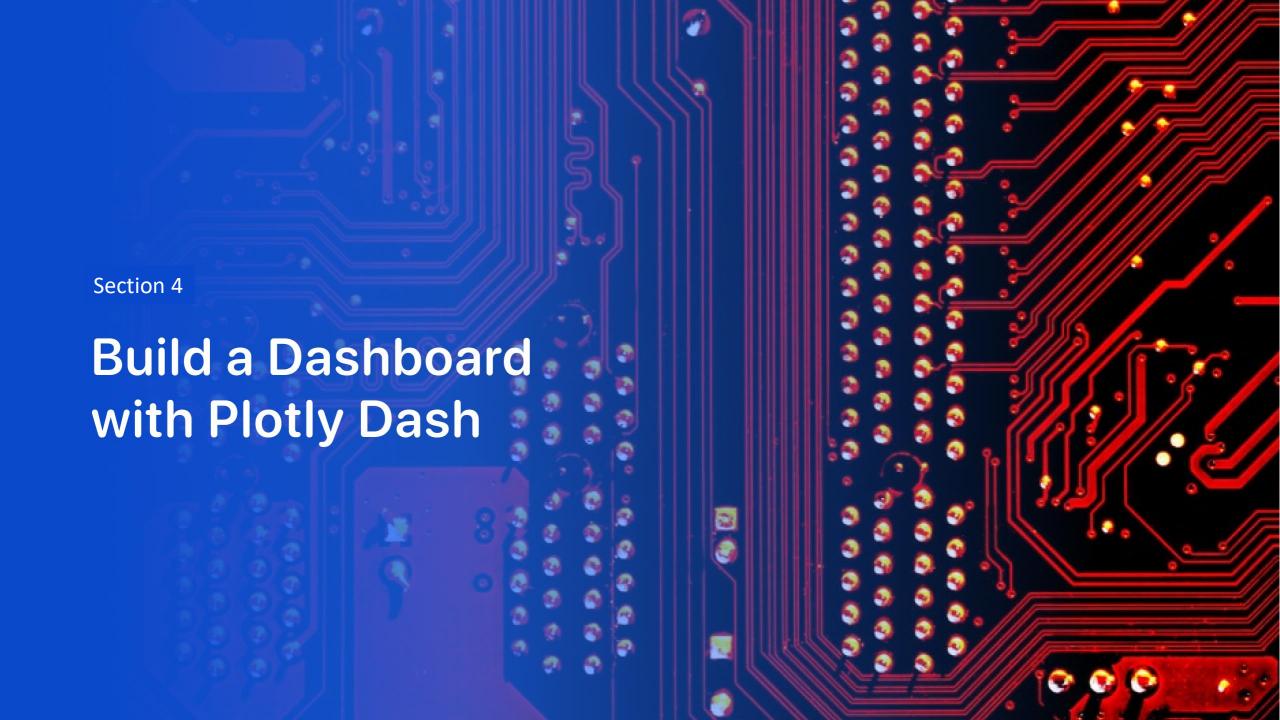
- From the color-labeled markers we should be able to easily identify which launch sites have relatively high success rates.
 - Green Marker = Successful Launch
 - Red Marker = Failed Launch
- Launch Site KSC LC-39A has a very high Success Rate.



Distance from the launch site CCAFS SLC-40 to its proximities

- From the visual analysis of the launch site CCAFS SLC-40 we can clearly see that it is:
 - relatively close to railway (1.26 km)
 - relatively close to highway (0.59 km)
 - relatively close to coastline (0.86 km)



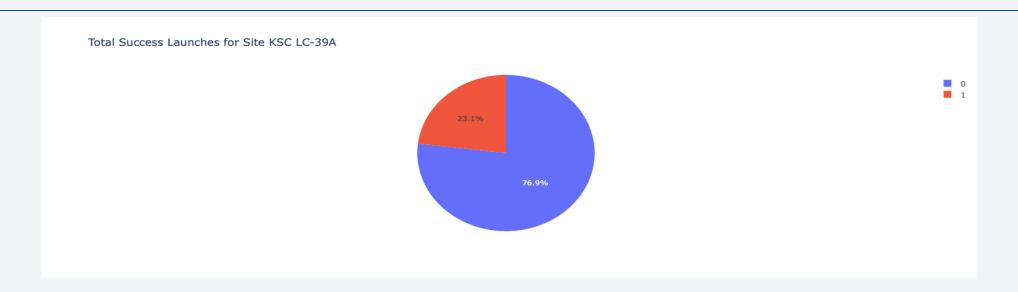


Launch success count for all sites



• The chart clearly shows that from all the sites, KSC LC-39A has the most successful launches.

Launch site with highest launch success ratio



 KSC LC-39A has the highest launch success rate (76.9%) with 10 successful and only 3 failed landings.

< Dashboard Screenshot 3>

• The charts show that payloads between 2000 and 5500 kg have the highest success rate.







Classification Accuracy

- Based on the scores of the Test Set, we can not confirm which method performs best.
- Same Test Set scores may be due to the small test sample size (18 samples).
 Therefore, we tested all methods based on the whole Dataset.
- The scores of the whole Dataset confirm that the best model is the Decision Tree Model. This model has not only higher scores, but also the highest accuracy.

Scores and Accuracy of the Test Set

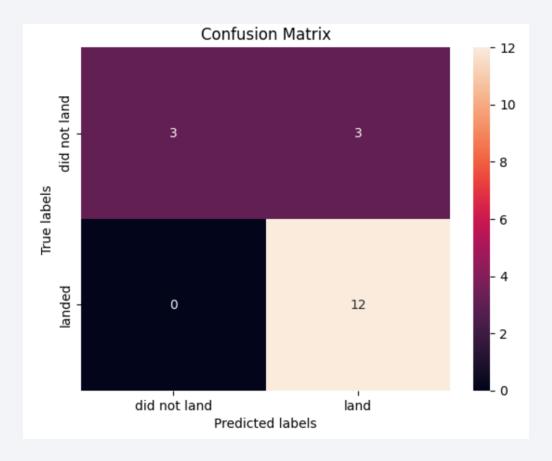
	LogReg	SVM	Tree	KNN
Jaccard_Score	0.800000	0.800000	0.800000	0.800000
F1_Score	0.888889	0.888889	0.888889	0.888889
Accuracy	0.833333	0.833333	0.833333	0.833333

Scores and Accuracy of the Entire Data Set

	LogReg	SVM	Tree	KNN
Jaccard_Score	0.833333	0.845070	0.882353	0.819444
F1_Score	0.909091	0.916031	0.937500	0.900763
Accuracy	0.866667	0.877778	0.911111	0.855556

Confusion Matrix

• 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

- Decision Tree Model is the best algorithm for this dataset.
- Launches with a low payload mass show better results than launches with a larger payload mass.
- Most of launch sites are in proximity to the Equator line and all the sites are in very close proximity to the coast.
- The success rate of launches increases over the years.
- KSC LC-39A has the highest success rate of the launches from all the sites.
- Orbits ES-L1, GEO, HEO and SSO have 100% success rate.

Appendix

Special Thanks to:

<u>Instructors</u>

Coursera

IBM

