

Winning Space Race with Data Science

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Outline

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

Executive Summary

- In this capstone project, we will predict if the SpaceX Falcon 9 first stage will land successfully using several machine learning classification algorithms.
- The main steps in this project include:
 - Data collection, wrangling, and formatting
 - Exploratory data analysis
 - Interactive data visualization
 - Machine learning prediction
- SpaceY is a recently established player in the commercial rocket launch industry, aiming to compete with SpaceX by bidding against them for launch contracts.
- SpaceX advertises launch services starting at \$67 million, which includes reserving fuel for reusing the firststage rocket booster.
- According to public statements from SpaceX, the cost of building a first-stage Falcon 9 rocket booster is estimated to be over \$15 million, not including R&D expenses or profit margin.
- The report shows that by considering mission parameters like payload mass and desired orbit, the models were able to predict the first-stage rocket booster landing with an accuracy level of 83.3%.
- Using first-stage landing predictions as a cost proxy for launches, SpaceYcan make more informed bids against SpaceX.

Introduction

- In this capstone, we will predict if the Falcon 9 first stage will land successfully.

 SpaceX 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. 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 SpaceX for a rocket launch.
- Most unsuccessful landings are planned. Sometimes, SpaceX will perform a controlled landing in the ocean.
- The main question that we are trying to answer is, for a given set of features about a Falcon 9 rocket launch which includes its payload mass, orbit type, launch site, and so on, will the first stage of the rocket land successfully?



- The overall methodology includes:
 - 1. Data collection, wrangling, and formatting, using:
 - SpaceX API
 - Web scraping
 - 2. Exploratory data analysis (EDA), using:
 - Pandas and NumPy
 - SQL
 - Data visualization, using:
 - Matplotlib and Seaborn
 - Folium
 - Dash
 - 4. Machine learning prediction, using
 - Logistic regression
 - Support vector machine (SVM)
 - Decision tree
 - K-nearest neighbors (KNN)

Data collection and wrangling

SpaceX API

- The API used is https://api.spacexdata.com/v4/rockets/.
- The API provides data about many types of rocket launches done by SpaceX, the data is therefore filtered to include only Falcon 9 launches.
- Every missing value in the data is replaced the mean the column that the missing value belongs to.
- We end up with 90 rows or instances and 17 columns or features. The picture below shows the first few rows of the data:

	FlightNumber	Date	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	LandingPad	Block	ReusedCount	Serial	Longitude	Latitude
4	1	2010- 06- 04	Falcon 9	NaN	LEO	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B0003	-80.577366	28.561857
5	2	2012- 05-22	Falcon 9	525.0	LEO	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B0005	-80.577366	28.561857
6	3	2013- 03-01	Falcon 9	677.0	ISS	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B0007	-80.577366	28.561857
7	4	2013- 09-29	Falcon 9	500.0	РО	VAFB SLC 4E	False Ocean	1	False	False	False	None	1.0	0	B1003	-120.610829	34.632093
8	5	2013- 12-03	Falcon 9	3170.0	GTO	CCSFS SLC 40	None None	1	False	False	False	None	1.0	0	B1004	-80.577366	28.561857

Data collection and wrangling

- Web scraping
 - The data is scraped from https://en.wikipedia.org/w/index.php?title=List of Falcon 9 and Falcon Heavy launches& oldid=1027686922
 - The website contains only the data about Falcon 9 launches.
 - We end up with 121 rows or instances and 11 columns or features. The picture below shows the first few rows of the data:

	Flight No.	Launch site	Payload	Payload mass	Orbit	Customer	Launch outcome	Version Booster	Booster landing	Date	Time
0	1	CCAFS	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success\n	F9 v1.0B0003.1	Failure	4 June 2010	18:45
1	2	CCAFS	Dragon	0	LEO	NASA	Success	F9 v1.0B0004.1	Failure	8 December 2010	15:43
2	3	CCAFS	Dragon	525 kg	LEO	NASA	Success	F9 v1.0B0005.1	No attempt\n	22 May 2012	07:44
3	4	CCAFS	SpaceX CRS-1	4,700 kg	LEO	NASA	Success\n	F9 v1.0B0006.1	No attempt	8 October 2012	00:35
4	5	CCAFS	SpaceX CRS-2	4,877 kg	LEO	NASA	Success\n	F9 v1.0B0007.1	No attempt\n	1 March 2013	15:10

Data collection and wrangling

- The data is later processed so that there are no missing entries and categorical features are encoded using one-hot encoding.
- An extra column called 'Class' is also added to the data frame. The column 'Class' contains 0 if a given launch is failed and 1 if it is successful.
- In the end, we end up with 90 rows or instances and 83 columns or features.

Exploratory Data Analysis (EDA)

- Pandas and NumPy
 - Functions from the Pandas and NumPy libraries are used to derive basic information about the data collected, which includes:
 - The number of launches on each launch site
 - The number of occurrence of each orbit
 - The number and occurrence of each mission outcome
- SQL
 - The data is queried using SQL to answer several questions about the data such as:
 - The names of the unique launch sites in the space mission
 - The total payload mass carried by boosters launched by NASA (CRS)
 - The average payload mass carried by booster version F9 v1.1

Data Visualization

Matplotlib and Seaborn

- Functions from the Matplotlib and Seaborn libraries are used to visualize the data through scatterplots, bar charts, and line charts.
- The plots and charts are used to understand more about the relationships between several features, such as:
 - The relationship between flight number and launch site
 - The relationship between payload mass and launch site
 - The relationship between success rate and orbit type

Folium

- Functions from the Folium libraries are used to visualize the data through interactive maps.
- The Folium library is used to:
 - Mark all launch sites on a map
 - Mark the succeeded launches and failed launches for each site on the map
 - Mark the distances between a launch site to its proximities such as the nearest city, railway, or highway

Data Visualization

Dash

- Functions from Dash are used to generate an interactive site where we can toggle the input using a dropdown menu and a range slider.
- Using a pie chart and a scatterplot, the interactive site shows:
 - The total success launches from each launch site
 - The correlation between payload mass and mission outcome (success or failure) for each launch site

Machine Learning Prediction

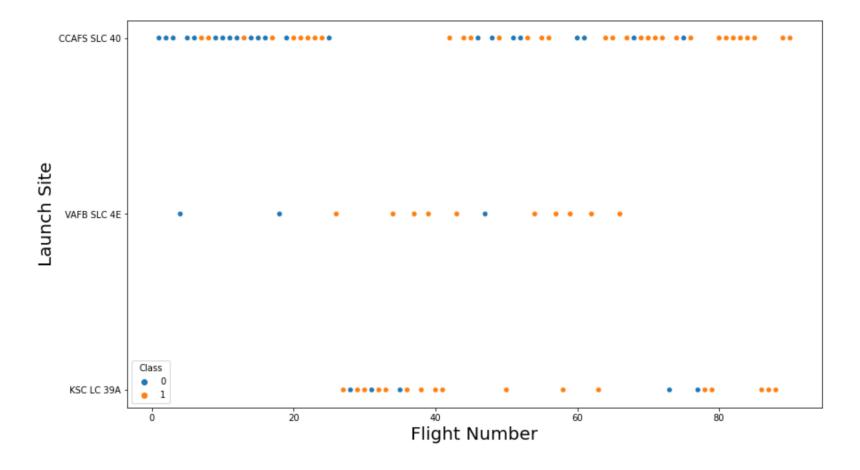
- Functions from the Scikit-learn library are used to create our machine learning models.
- The machine learning prediction phase include the following steps:
 - Standardizing the data
 - Splitting the data into training and test data
 - Creating machine learning models, which include:
 - Logistic regression
 - Support vector machine (SVM)
 - Decision tree
 - K nearest neighbors (KNN)
 - Fit the models on the training set
 - Find the best combination of hyperparameters for each model
 - Evaluate the models based on their accuracy scores and confusion matrix

- The results are split into 5 sections:
 - SQL (EDA with SQL)
 - Matplotlib and Seaborn (EDA with Visualization)
 - Folium
 - Dash
 - Predictive Analysis
- In all of the graphs that follow, class 0 represents a failed launch outcome while class 1 represents a successful launch outcome.



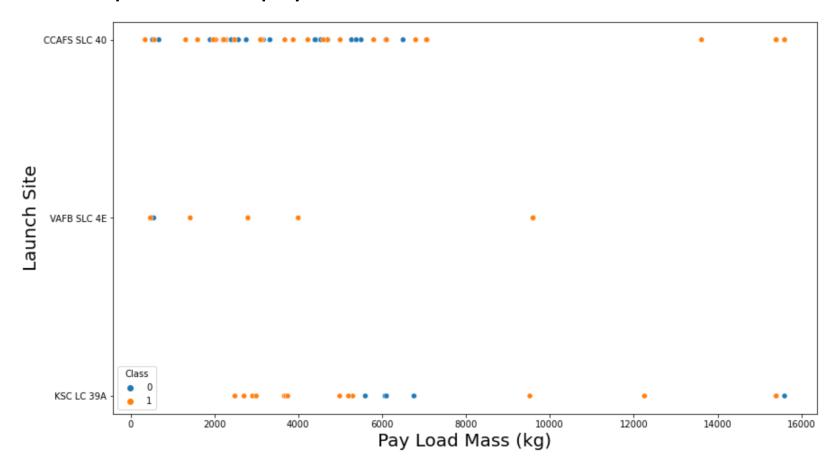
Matplotlib and Seaborn (EDA with Visualization)

• The relationship between flight number and launch site



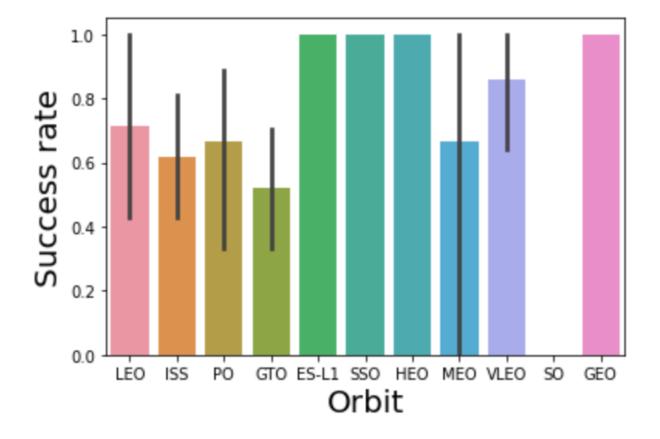
Matplotlib and Seaborn (EDA with Visualization)

The relationship between payload mass and launch site



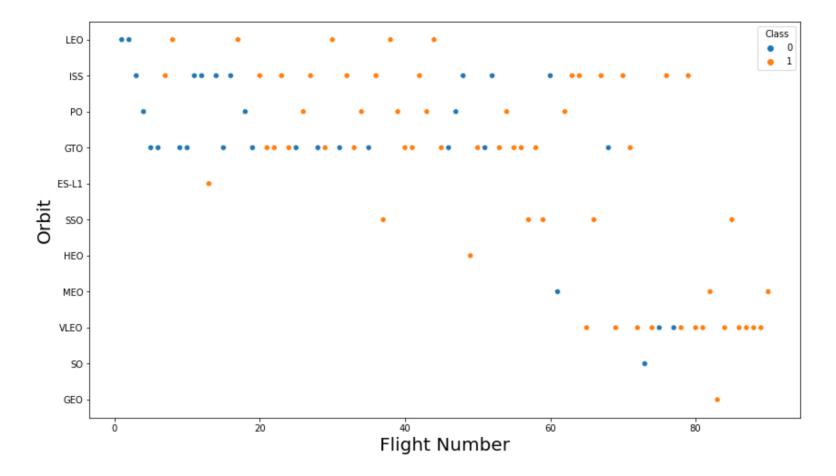
Matplotlib and Seaborn (EDA with Visualization)

• The relationship between success rate and orbit type



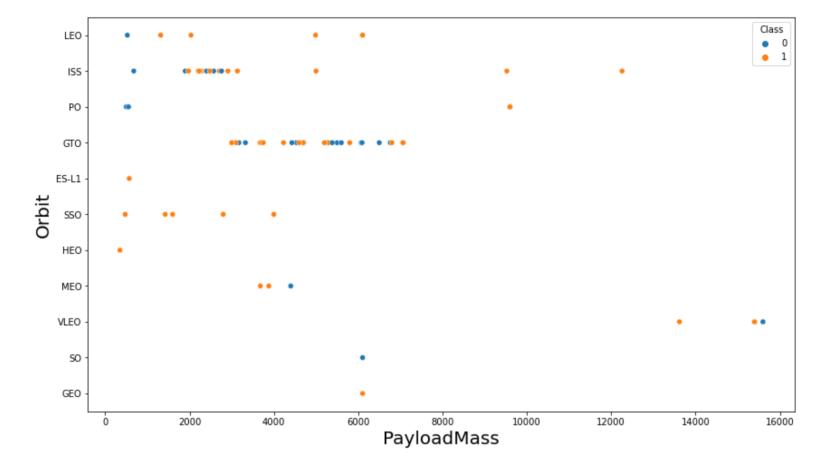
Matplotlib and Seaborn (EDA with Visualization)

• The relationship between flight number and orbit type



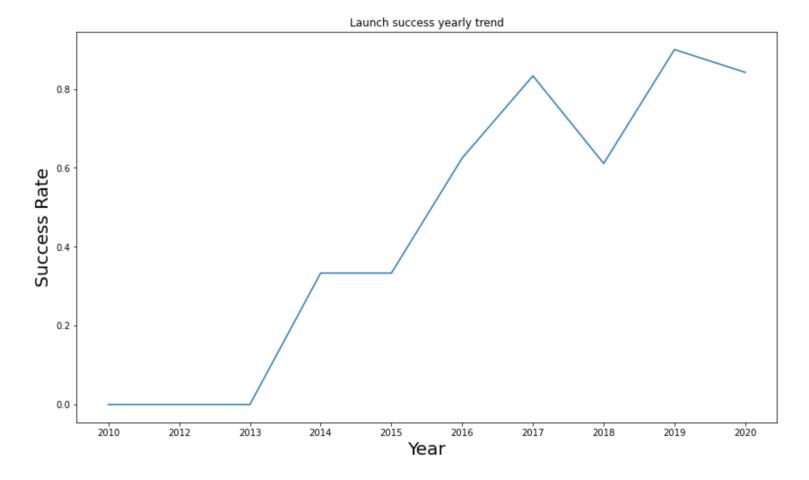
Matplotlib and Seaborn (EDA with Visualization)

• The relationship between payload mass and orbit type



Matplotlib and Seaborn (EDA with Visualization)

The launch success yearly trend



• The names of the unique launch sites in the space mission

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

• 5 records where launch sites begin with 'CCA'

DATE	timeutc_	booster_version	launch_site	payload	payload_masskg_	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

The total payload mass carried by boosters launched by NASA (CRS)

Total payload mass by NASA (CRS) 45596

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

Average payload mass by Booster Version F9 v1.1

2928

The date when the first successful landing outcome in ground pad was achieved

Date of first successful landing outcome in ground pad

2015-12-22

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

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

• The total number of successful and failure mission outcomes

```
number_of_success_outcomes number_of_failure_outcomes 100 1
```

The names of the booster versions which have carried the maximum payload mass

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

 The failed landing outcomes in drone ship, their booster versions, and launch site names for in year 2015

launch_site	booster_version	DATE
CCAFS LC-40	F9 v1.1 B1012	2015-01-10
CCAFS LC-40	F9 v1.1 B1015	2015-04-14

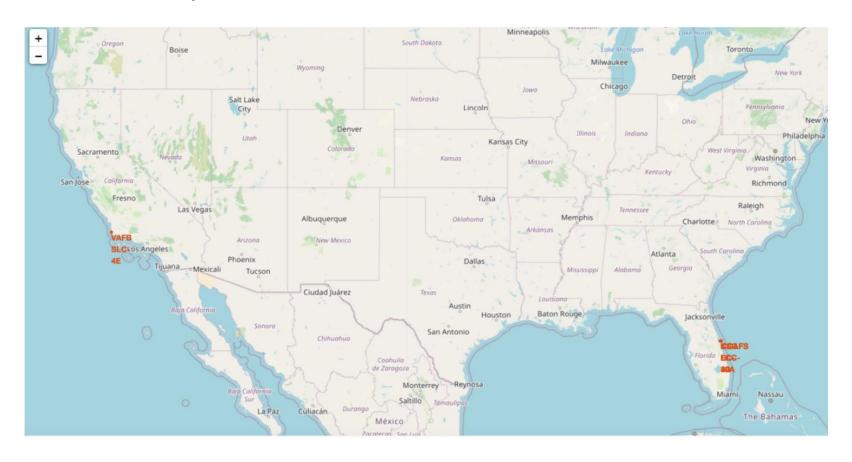
• The count of landing outcomes between the date 2010-06-04 and 2017-03-20, in descending order

landing_outcome	landing_count
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



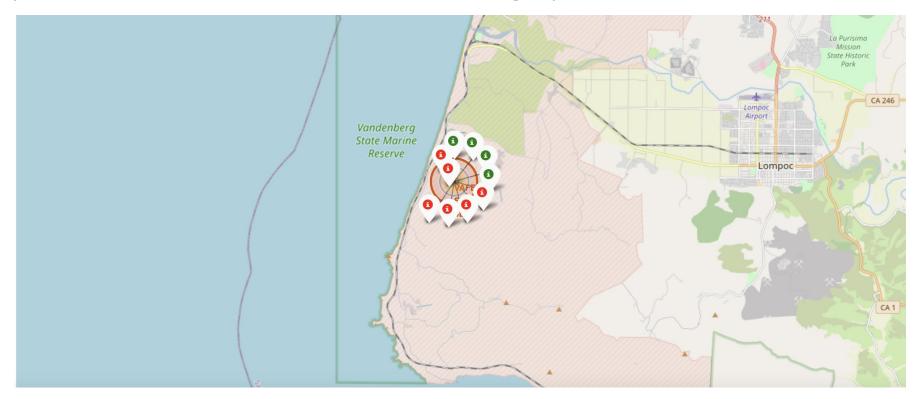
RESULTS Folium

All launch sites on map



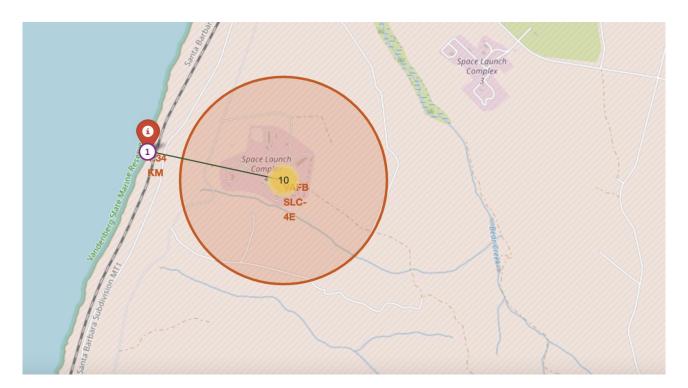
RESULTS Folium

- The succeeded launches and failed launches for each site on map
 - If we zoom in on one of the launch site, we can see green and red tags. Each green tag represents a successful launch while each red tag represents a failed launch



RESULTS Folium

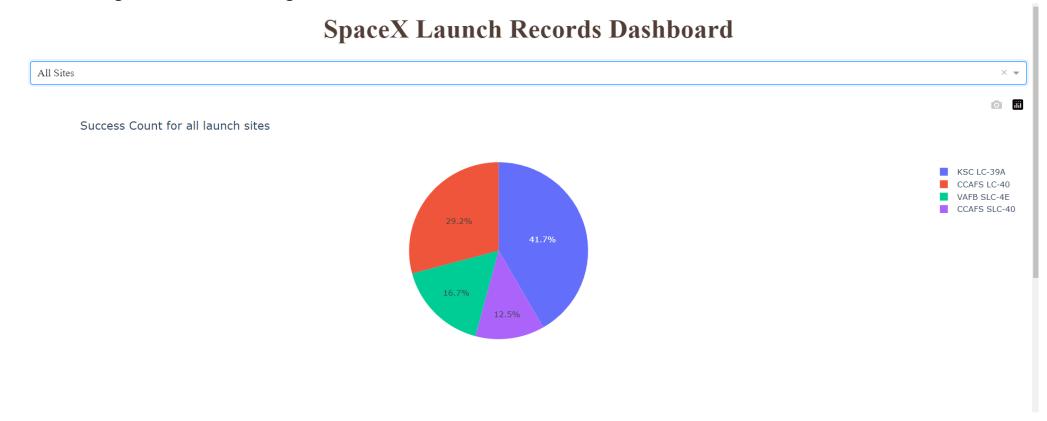
- The distances between a launch site to its proximities such as the nearest city, railway, or highway
 - The picture below shows the distance between the VAFB SLC-4E launch site and the nearest coastline





Launch Records Dashboard

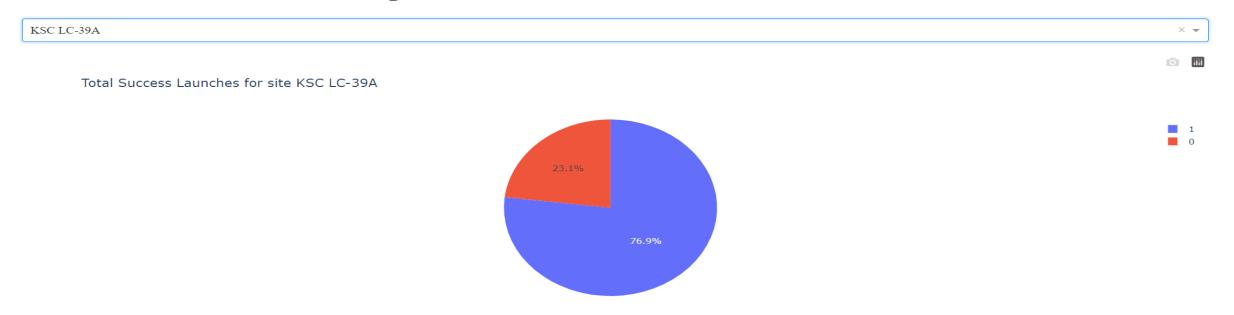
- Drop-down menu to choose between all sites and individual launch sites.
- Color coded by launch sites.
- Pie chart showing booster landing success rate.



Dashboard Observations

- Enabling stakeholders to explore and manipulate the data in an interactive and real-time way
- KSC LC 39A has the highest booster landing success rate.

SpaceX Launch Records Dashboard



Dashboard Observations

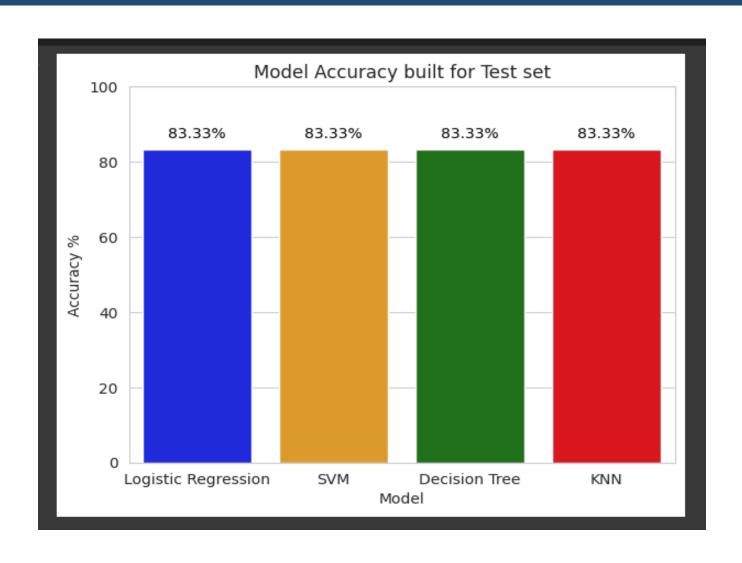
- The picture below shows a scatterplot when the payload mass range is set to be from 2000kg to 8000kg.
- Class 0 represents failed launches while class 1 represents successful launches.



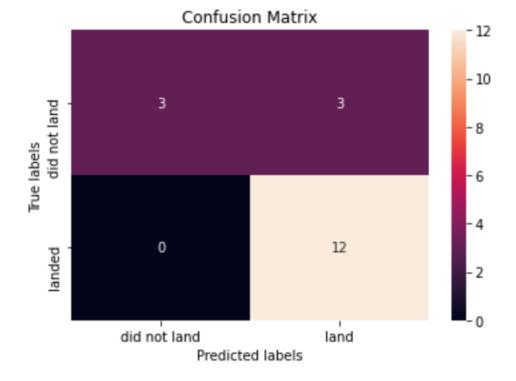


Classification Accuracy

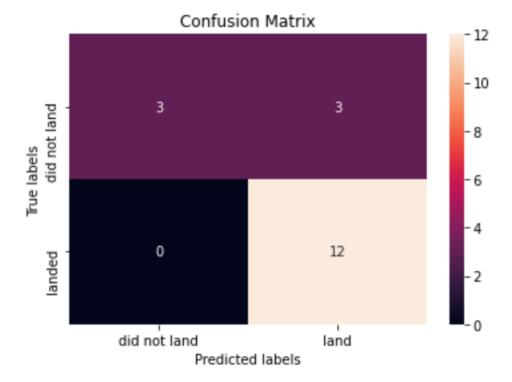
- All the models built on the test set has an accuracy score of 83.33%.
- Therefore, their GridSearchCV best scores are used to rank them instead. Based on the GridSearchCV best scores, the models are ranked in the following order with the first being the best and the last one being the worst:
 - 1. Decision tree (GridSearchCV best score: 0.8892857142857142)
 - 2. K nearest neighbors, KNN (GridSearchCV best score: 0.8482142857142858)
 - 3. Support vector machine, SVM (GridSearchCV best score: 0.8482142857142856)
 - 4. Logistic regression (GridSearchCV best score: 0.8464285714285713)



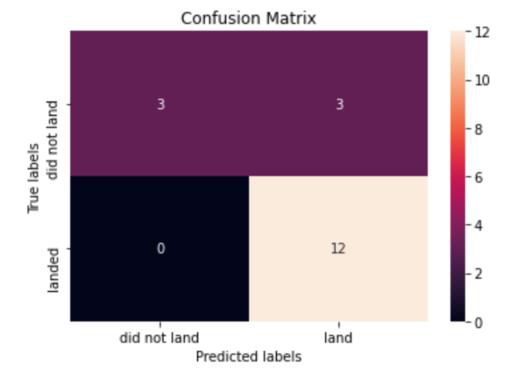
- Logistic regression
 - GridSearchCV best score: 0.8464285714285713
 - Accuracy score on test set: 0.8333333333333333
 - Confusion matrix:



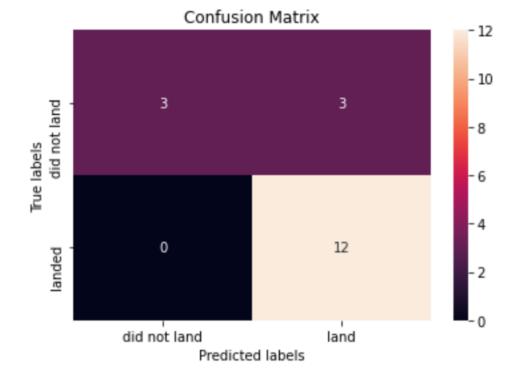
- Support vector machine (SVM)
 - GridSearchCV best score: 0.8482142857142856
 - Accuracy score on test set: 0.8333333333333333
 - Confusion matrix:



- Decision tree
 - GridSearchCV best score: 0.8892857142857142
 - Accuracy score on test set: 0.8333333333333333
 - Confusion matrix:



- K nearest neighbors (KNN)
 - GridSearchCV best score: 0.8482142857142858
 - Accuracy score on test set: 0.8333333333333333
 - Confusion matrix:



CONCLUSION

- This report provides models that SpaceY can use to make predictions about the successful landing of the 1st stage booster by SpaceX These models have an accuracy of 83 3 meaning that they can predict the outcome with a high level of confidence
- SpaceY will be able to make more informed bids against SpaceX This is because SpaceY will have a more accurate idea of when they can expect the SpaceX bid to include the cost of a sacrificed 1st stage booster, allowing them to better plan and allocate their resources
- In this project, we try to predict if the first stage of a given Falcon 9 launch will land in order to determine the cost of a launch.
- Each feature of a Falcon 9 launch, such as its payload mass or orbit type, may affect the mission outcome in a certain way.
- Several machine learning algorithms are employed to learn the patterns of past Falcon 9 launch data to produce predictive models that can be used to predict the outcome of a Falcon 9 launch.
- The predictive model produced by the decision tree algorithm performed the best among the 4 machine learning algorithms employed.

Appendix

▶ Py	thon code snippets, SQL queries, charts, Notebook outputs, datasets analysis, and models:
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/01-%20Data%20Collection%20API.ipynb
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/02- %20Data%20Collection%20with%20Web%20Scraping.ipynb
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/03-%20EDA.ipynb
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/04-%20EDA%20with%20SQL.ipynb
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/05-%20EDA%20Visualization.ipynb
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/06-
	%206Interactive%20Visual%20Analytics%20with%20Folium%20lab.ipynb
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/07-%20spacex_dash_app.py
	https://github.com/Emad-eldeen-Elsayed/IBM-Applied-Data-Science-Capstone-Project/blob/main/08-
	%20Machine%20Learning%20Prediction.ipynb

