

Winning Space Race with Data Science

Tuan Duong January 2, 2022



Executive Summary

- SpaceY is a new commercial rocket company founded by Billionaire industrialist, Allon Mask, that would like to join the space race.
- The main competitor, SpaceX, advertises launch services starting at \$62 million versus their competitors services upwards of \$165 million dollars.
- The reason for this is SpaceX reuses the first stage of rocket launches, saving millions of dollars.
- By using past public data from the Space X Falcon 9 launches, SpaceY will determine which parameters are important for successful first stage launches and reusability.
- As a result, SpaceY will be able to make more informed bids against SpaceX by using 1st stage landing predictions as the basis for their future launches.



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Introduction: Background

- This project is being presented in the role of a data scientist working for the fictitious new rocket company, SpaceY, that would like to compete with SpaceX founded by Billionaire industrialist Allon Mask.
- It was prepared for the IBM Applied Data Science Capstone course.
- This report includes the data scientist methodology, results, and conclusions on their findings.

Introduction: Business Problem



- SpaceX advertises Falcon 9 rocket launches with a cost of 62 million dollars because they can reuse the first stage of rocket launches.
- The first stage is estimated to cost upwards of 15 million dollars to create.
- This report aims to accurately predict the likelihood of the first stage rocket landing successfully for reusability.



Methodology

Overview of methodologies used by data scientist

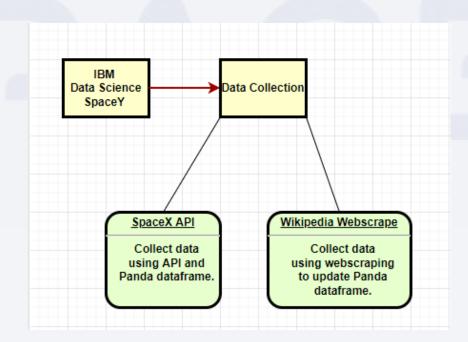
- Data Collection
- Data Wrangling
- Exploratory Data Analysis: Visual Analytics
- Exploratory Data Analysis: SQL
- Data Visualization: Map
- Data Visualization: Dashboard
- Predictive Analysis (Classification)
- Results

Methodology: Data Collection

Two methods were used to collect data.

Open-Source Rest API: Application programming interfaces was used with request library to get data using SpaceX API.

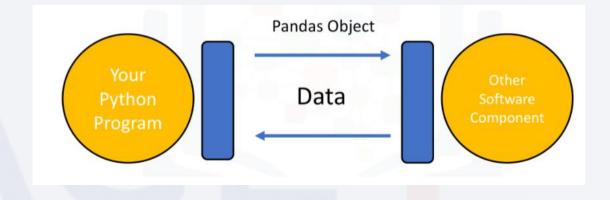
Web scraping: the process of using **beautifulsoup** library to extract content and data from a website. The data was collected using the public website, **Wikipedia**.



Methodology: Data Collection - SpaceX API

SpaceX API

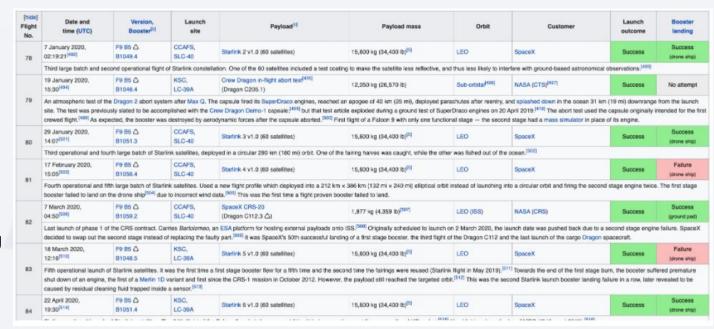
- Gathered historical public data using Open-Source API for SpaceX:
 - Request and parse the SpaceX launch data using the GET request.
 - Loaded information with Get Functions and loaded into Panda Dataframe.
 - Filtered data to only use Falcon 9 launches.



Methodology: Data Collection - Scraping

Webscraping Wikipedia

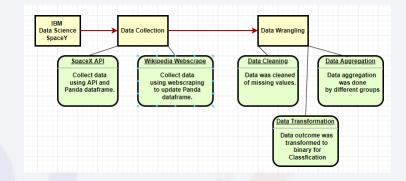
- Webscraping Wikipedia for public knowledge:
 - Request the Falcon9 Launch Wiki page from its URL.
 - Extract all column/variable names from the HTML table header.
 - Create a data frame by parsing the launch HTML tables using Panda's Dataframe



Methodology: Data Wrangling

Data wrangling

- The process of cleaning and unifying messy and complex data sets for easy access and analysis.
 - Data was loaded into Panda's Dataframe
 - Identify and calculate the percentage of the missing values in each attribute
 - Changing missing values and imputing their mean
 - · Identify which columns are numerical and categorical
 - Calculate the number of launches on each site
 - Calculate the number and occurrence of each orbit
 - Calculate the number and occurrence of mission outcome per orbit type
 - Transformed data to binary column, class, where O = unsuccessful, 1 = successful

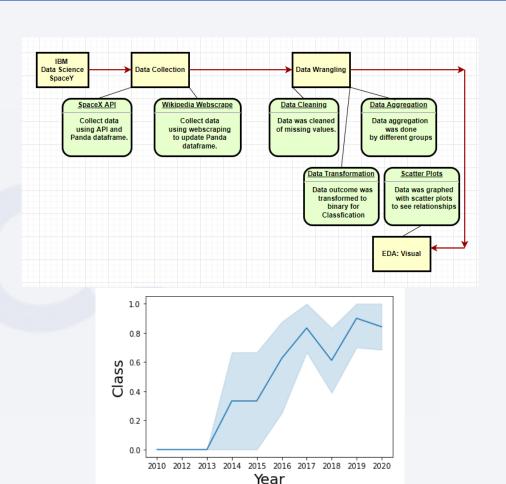




Methodology: EDA with Data Visualization

EDA: Data Visualization

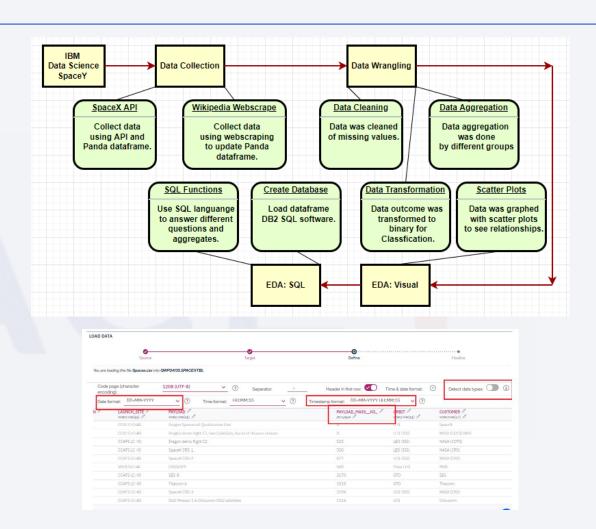
- Exploratory data analysis (EDA) is used by data scientists to analyze and investigate data sets and summarize their main characteristics, often employing data visualization methods.
 - Scatter plots were used to explore the target variable (class) and the different relationships it had with various features.
 - · Bar graphs to see frequency.
 - Line graphs to see trends.
 - Features included: payload in weight, launch-site locations, orbit, time, landing pad, block, grid fins, reused counts, and serials.



Methodology: EDA with SQL

EDA: Data Visualization

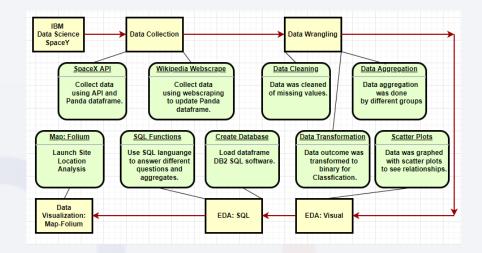
- Exploratory data analysis (EDA) is used by data scientists to analyze and investigate data sets and summarize their main characteristics, this time using SQL (Structured Query Language).
- Loaded dataframe as a database using DB2
- Used SQL to answer questions about features.
 - Ex1. List the total number of successful and failure mission outcomes



Methodology: Build an Interactive Map with Folium

Data Visualization: Launch Sites Locations Analysis with Folium

- Data visualization is the graphical representation of information and data. By using visual elements like charts, graphs, and maps, data visualization tools provide an accessible way to see and understand trends, outliers, and patterns in data.
 - The data scientist wanted to determine if location was an important feature of target variable.
 - Visual map is one approach to see success rate of different launch sites and their surroundings.
 - The data scientist can engage in risk management, transportation, and other location factors.

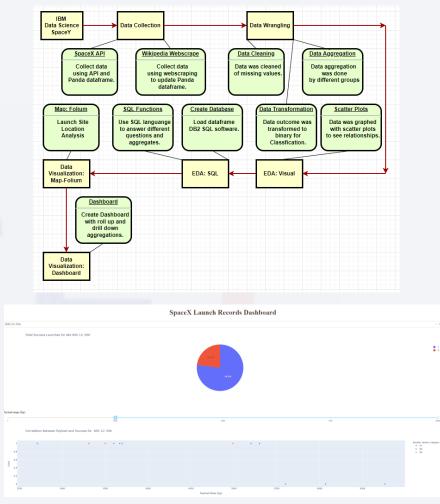




Methodology: Build a Dashboard with Plotly Dash

Data Visualization: Dashboard

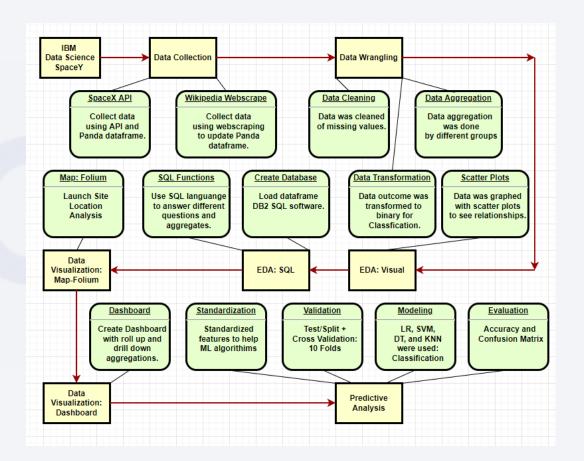
- A dashboard is a type of graphical user interface which often provides at-a-glance views of key performance indicators (KPIs) relevant to a particular objective or business process.
 - Dashboard was created to help stakeholders visualize findings through roll up/drill down and different menus to see data at different angles.
 - Strong features and aggregations were shown with biggest impact on target variable.



Methodology: Predictive Analysis (Classification)

Predictive analytics: Classification

- Use of data, statistical algorithms and machine learning techniques to identify the likelihood of future outcomes based on historical data.
 - Classification was used since the data scientist used a binary target variable with success rate outcome O = unsuccessful and 1 = successful
 - · Standardized features to help ML algorithm train
 - Train/Test Split and Cross Validation: 10 folds were used to validate the data
 - Tried 4 different ML methods of logistic regression, sector vector machine, decision tree, and k nearest neighbors
 - Evaluation methods included accuracy check of classification and confusion matrix





Results

- EDA: Visualization Results
- EDA: SQL Results
- Data Visualization: Map Results
- Data Visualization: Dashboard
- Predictive Analysis
- Evaluation

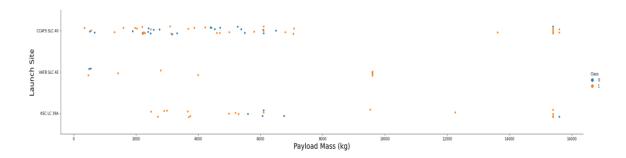


NSC UC 384-



Flight Number vs. Launch Site

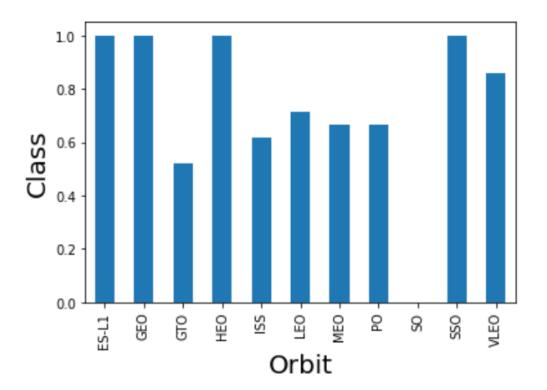
- The data scientist used an EDA visualization scatter plot graph using Seaborn and Matplotlib, where x = Flight Number and Y = Launch Site.
- Blue dots were unsuccessful, orange dots were successful.
- Success rate affected by Location feature?
- Visually, yes, more orange/blue dot ratio on VAFB and KSC locations. Need further investigation.





Payload vs. Launch Site

- The data scientist used an EDA visualization scatter plot graph using Seaborn and Matplotlib, where x = Payload Mass (kg) and Y = Launch Site.
- Blue dots were unsuccessful, orange dots were successful.
- Success rate affected by Location/Weight features?
- Visually, yes, more orange/blue dot ratio on VAFB and KSC locations. Need further investigation.



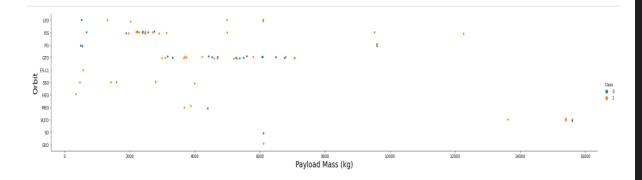
Success Rate vs. Orbit Type

- The data scientist used an EDA visualization bar graph using Seaborn and Matplotlib, where x = Orbit and Y = Success Rate.
- Success rate affected by Orbit feature?
- Visually, not enough information. Not enough samples on some categories.



Flight Number vs. Orbit Type

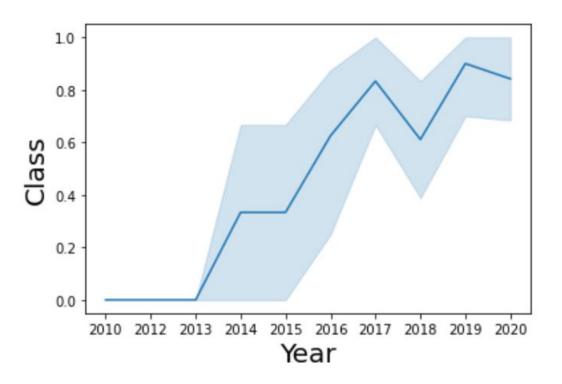
- The data scientist used an EDA visualization scatter plot graph using Seaborn and Matplotlib, where x = Flight Number and Y = Orbit.
- Blue dots were unsuccessful, orange dots were successful.
- Success rate affected by Orbit feature?
- Visually, not enough information. Not enough samples on some categories.





Payload vs. Orbit Type

- The data scientist used an EDA visualization scatter plot graph using Seaborn and Matplotlib, where x = Payload Mass (kg) and Y = Orbit.
- Blue dots were unsuccessful, orange dots were successful.
- Success rate affected by Weight/Orbit features?
- Visually, not enough information. Not enough samples on some categories.



Launch Success Yearly Trend

- The data scientist used an EDA visualization line graph using Seaborn and Matplotlib, where x = Time and Y = Success Rate.
- Success rate affected by Time feature?
- Visually, definite yes, but still need further investigation.

```
%%sql
SELECT DISTINCT launch_site
FROM spacex
```

* ibm_db_sa://tjf73799:***@:
Done.

launch_site

CCAFS LC-40

CCAFS SLC-40

KSC LC-39A

VAFB SLC-4E

All Launch Site Names

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- Display the names of the unique launch sites in the space mission?
- Answer:
 - CCAFS LC-40
 - CCAFS SLC-40
 - KSC LC-39A
 - VAFB SLC-4E

%%sql SELECT launch_site FROM spacex WHERE launch_site LIKE 'CCA%' LIMIT 5

* ibm_db_sa://tjf73799:***@21 Done.

launch_site

CCAFS LC-40

CCAFS LC-40

CCAFS LC-40

CCAFS LC-40

CCAFS LC-40

Launch Site Names Begin with 'CCA'

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- Display 5 records where launch sites begin with the string 'CCA'
- Answer:
 - CCAFS LC-40

```
%%sql
SELECT SUM(payload_mass__kg_)
FROM spacex
WHERE customer = 'NASA (CRS)'
```

```
* ibm_db_sa://tjf73799:***@2f
Done.
```

1

45596

Total Payload Mass

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- Display the total payload mass carried by boosters launched by NASA (CRS)
- Answer:
 - 45596

```
%%sql
SELECT avg(payload_mass__kg_)
FROM spacex
WHERE booster_version = 'F9 v1.1'
```

* ibm_db_sa://tjf73799:***@2f3279
Done.

1

2928

Average Payload Mass by F9 v1.1

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- Display average payload mass carried by booster version F9 v1.1
- Answer:
 - 2928

```
%%sql
SELECT MIN(date)
FROM spacex
WHERE landing__outcome = 'Success (ground pad)'
```

* ibm_db_sa://tjf73799:***@2f3279a5-73d1-4859-

1

2015-12-22

First Successful Ground Landing Date

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- List the date when the first successful landing outcome in ground pad was achieved.
- Answer:
 - 12/22/2015

```
%%sql
SELECT booster_version
FROM spacex
WHERE landing_outcome = 'Success (drone ship)'
AND 4000 < payload_mass__kg_ < 6000</pre>
```

* ibm_db_sa://tjf73799:***@2f3279a5-73d1-4859-& Done.

booster_version

F9 FT B1021.1

F9 FT B1023.1

F9 FT B1029.2

F9 FT B1038.1

F9 B4 B1042.1

F9 B4 B1045.1

F9 B5 B1046.1

Successful Drone Ship Landing with Payload between 4000 and 6000

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- Answer:
 - F9 FT B1021.1
 - F9 FT B1023.1
 - F9 FT B1029.2
 - F9 FT B1038.1
 - F9 B4 B1042.1
 - F9 B4 B1045.1
 - F9 B4 B1046.1

```
%%sql
SELECT mission_outcome, COUNT(*)
FROM spacex
GROUP BY mission_outcome
```

* ibm_db_sa://tjf73799:***@2f327 Done.

```
mission_outcome 2
Failure (in flight) 1
Success 99
Success (payload status unclear) 1
```

Total Number of Successful and Failure Mission Outcomes

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- List the total number of successful and failure mission outcomes
- Answer:

• Failure: 1

• Success: 99

• Unclear: 1

```
%%sql
SELECT booster_version
FROM spacex
WHERE payload_mass__kg_ IN (SELECT MAX(payload_mass__kg_)
FROM spacex)
```

* ibm_db_sa://tjf73799:***@2f3279a5-73d1-4859-88f0-a6c3e6

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

Boosters Carried Maximum Payload

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- List the names of the booster_versions which have carried the maximum payload mass. Use a subquery
- Answer:
 - F9 FT B1048.4
 - F9 FT B1048.4
 - F9 FT B1051.3
 - F9 FT B1056.4
 - F9 FT B1048.5
 - F9 FT B1051.4
 - F9 FT B1049.5
 - F9 FT B1060.2
 - F9 FT B1058.3
 - F9 FT B1051.6
 - F9 FT B1060.3
 - F9 FT B1049.7

```
%%sql
SELECT landing__outcome, booster_version, launch_site, date
FROM spacex
WHERE landing__outcome = 'Failure (drone ship)'
AND YEAR(date) = 2015
```

* ibm_db_sa://tjf73799:***@2f3279a5-73d1-4859-88f0-a6c3e6b4 Done.

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

2015 Launch Records

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- List the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- Answer:
 - Failure (drone ship) F9 v1.1 B1012 CCAFS LC-40 2015-01-10
 - Failure (drone ship) F9 v1.1 B1015
 CCAFS LC-40 2015-04-14

%%sql SELECT landing__outcome, count(landing__outcome) FROM spacex WHERE date BETWEEN '2010-06-04' AND '2017-03-20' GROUP BY landing__outcome ORDER BY 2 DESC

* ibm_db_sa://tjf73799:***@2f3279a5-73d1-4859-88 Done.

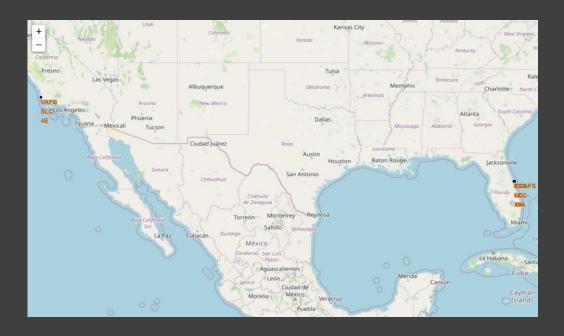
landing_outcome	2
No attempt	10
Failure (drone ship)	5
Success (drone ship)	5
Controlled (ocean)	3
Success (ground pad)	3
Failure (parachute)	2
Uncontrolled (ocean)	2
Precluded (drone ship)	1

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

- The data scientist used an EDA SQL Database and SQL query to answer the following question:
- 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
- Answer:

•	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

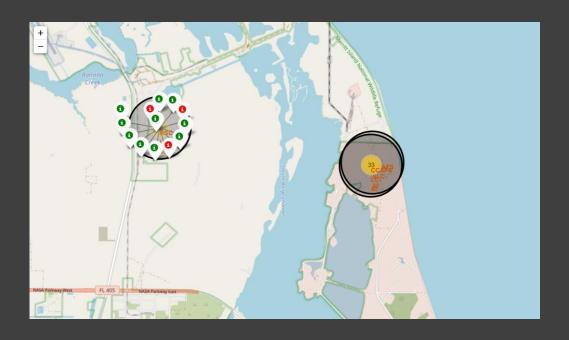






All Launch Sites

- This map was produced by using the panda library: folium and inputting the data location through the data frame.
- The importance of data visualization is to help the data scientist understand if location has any effect on successful launches and their risks.
- This map shows all the locations of the launches.





Success Rate by Launch-Site

- This map was produced by using the panda library of folium and inputting the data location through the data frame.
- The importance of data visualization is to help the data scientist understand if location has any effect on successful launches and their risks.
- This map shows the success/failed launches of each site by clicking on any circled marker, green means success and red means failure.





Nearest Risks/Transportation

- This map was produced by using the panda library of folium and inputting the data location through the data frame.
- The importance of data visualization is to help the data scientist understand if location has any effect on successful launches and their risks.
- This map shows the nearest railroads, highways, and cities to determine the risks and modes of transportations from CCAFS LC 40 launch site.
 - Nearest Railroad = 0.35130446057664205 km
 - Nearest Highway = 0.5890306859613954 km
 - Nearest City = 25.065251129546912 km



SpaceX Launch Records Dashboard



Successful Launches All Sites

- The data scientist made this dashboard using dash and plotly libraries.
- The following Pie Chart represents all of the successful launches through the 4 sites.
- Descending Order

• KSC LC-39A: 41.7%

• CCAFS LC-40: 29.2%

• VAFB SLC-4E: 16.7%

CCAFS SLC-40: 12.5%

SpaceX Launch Records Dashboard

Total Success Launches for site KSC LC-39A

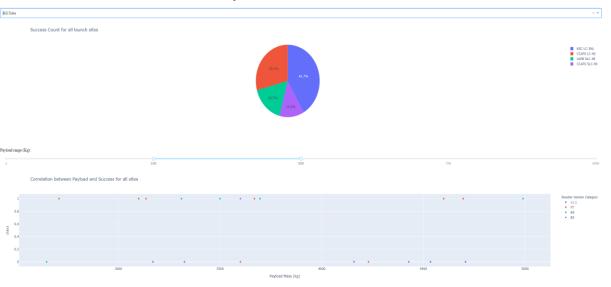


KSC LC-39A Pie Chart

- The data scientist made this dashboard using dash and plotly libraries.
- The following Pie Chart represents the launch site with the highest success ratio.
- They had 3 failed launches and 10 successful launches.
- KSC LC-39A had the highest success rate of 10/13 = 76.9%.



SpaceX Launch Records Dashboard



All Sites, 2500-5000 Payload + FT Booster

- The data scientist made this dashboard using dash and plotly libraries.
- The following Line Chart represents the launch site with the highest success ratio with payloads and boosters.
- The following success rate of payload are as follows:

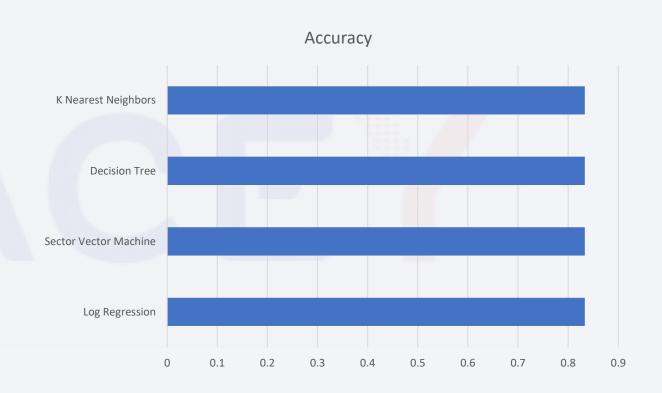
•	0 – 2500	Success/Failure	7/12	36%
•	2500-5000	Success/Failure	11/9	55%
•	5000-7500	Success/Failure	2/7	22%
•	7500-10000	Success/Failure	1/1	50%

 The most successful booster within the payload range of 2500-5000 is the FT booster.



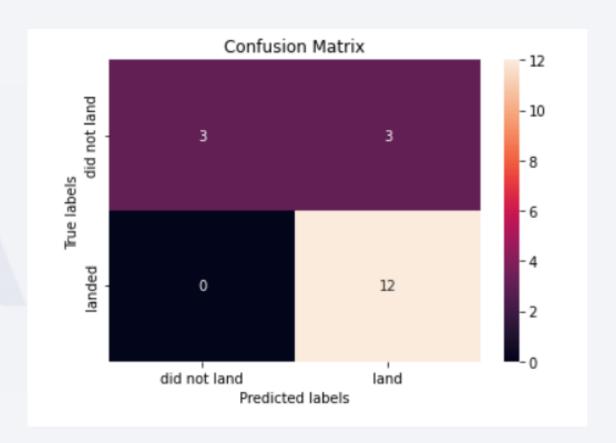
Classification Accuracy

- The following bar chart represents the accuracy of all 4 Machine Learning algorithms used for classification.
- K Nearest Neighbors, Decision Tree, Sector Vector Machine, and Log Regression.
- All of them have the same accuracy rating of 83.3%.



Confusion Matrix

- 4 confusion matrices were used for each classification machine learning method, with all 4 methods having the same endproduct.
- The matrix shows that there were 12 true positives, 3 false positives, 0 false negatives, and 3 true negatives.
- This means all 4 ML methods had 83.3% accurate at classifying each event besides false positives.



Conclusions

- Classification models from this report will help SpaceY predict successful 1st stage landings with 83.3% accuracy.
- SpaceX indicates the 1st stage booster costs upwards of \$15 million to build.
- Statistically, these findings can help SpaceY attribute each launch to \$2.5 million with this accuracy. (15 * .167 = \sim 2.5)
- Understanding the data with the best parameters will also increase the success rate.
 - Some of the findings suggest the organization can optimize: launch sites, weight, booster, etc.
- This data and models should be iterative and can be further approved with every launch.

Appendix & Acknowledgements

Appendix

- The following links were produced on Juypter Notebooks using Python code and libraries. These libraries include: Panda, Numpy, Matplotlib, Seaborn, Dash, Request, Beautifulsoup, API, Sk-Learn, SQL, DB2, SqlAlchemy, IBM, Ipython, Datetime, Folium, and Wget.
- https://github.com/viettwoone/ibm_data_science/blob/main/SpaceX_Machine%20Learning%20Prediction_Part_5.ipynb
- https://github.com/viettwoone/ibm_data_science/blob/main/jupyter-labs-eda-dataviz.ipynb
- https://github.com/viettwoone/ibm data science/blob/main/jupyter-labs-eda-sql-coursera.ipynb
- https://github.com/viettwoone/ibm_data_science/blob/main/jupyter-labs-spacex-data-collection-api.ipynb
- https://github.com/viettwoone/ibm data science/blob/main/jupyter-labs-webscraping.ipynb
- https://github.com/viettwoone/ibm data science/blob/main/lab jupyter launch site location.jpynb
- https://github.com/viettwoone/ibm data science/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb
- https://github.com/viettwoone/ibm_data_science/blob/main/spacex_dash_app.ipynb

Acknowledgements

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