

Winning Space Race with Data Science

IBM Data Science Capstone Project Conrad Kleykamp 7/24/24



OUTLINE

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- Methodology
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EXECUTIVE SUMMARY

Methodologies

- Data Collection
 - API, Web Scraping (BeautifulSoup)
- Data Wrangling
 - Missing values
 - Training label → 'Outcome' column
 - One Hot Encoding
- Exploratory Data Analysis (EDA)
 - SQL Queries
 - Visualization w/ Matplotlib, Seaborn

- <u>Launch Sites Locations Analysis</u>
 - Folium
- Launch Records Dashboard
 - Plotly Dash
- Predictive Analysis
 - Logistic Regression
 - Support Vector Machine
 - Decision Tree Classifier
 - K Nearest Neighbors Classifier

EXECUTIVE SUMMARY

General Results

- Binary Classification
 - Outcome' → 1 (Successful Landing)
 - Outcome' → 0 (Unsuccessful Landing)
- ~ 83.33% test set accuracy across all predictive methods
- ~ 88.88% train set accuracy for the Decision Tree Classifier

INTRODUCTION

Background

- Completed as part of IBM's Applied Data Science Capstone course
- SpaceY is a fictional commercial rocket launch provider that wishes to compete against SpaceX
- SpaceX's Falcon 9 rocket launches advertised to cost \$62 million
 - Other rocket launch providers cost ~\$165 million
 - SpaceX savings are due to the ability to land and reuse the first stage

INTRODUCTION

Business Problem

- How can a company like SpaceY bid against SpaceX for a rocket launch?
 - Must deduce the cost of a launch
- Using public SpaceX data, if we can determine whether the first stage will land, we can determine the cost of a launch
- This project will leverage multiple machine learning models to accurately predict the likelihood of a successful landing of the first stage rocket
- Important variables to consider:
 - Payload mass, launch site, orbit, number of flights



METHODOLOGY

Overview

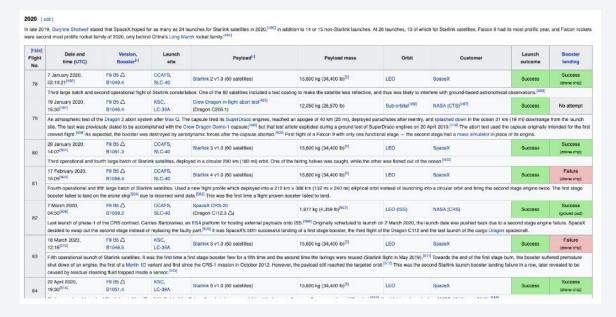
- 1: Data Collection
 - API, Web Scraping (BeautifulSoup)
- 2: Data Wrangling
 - Missing values
 - Training label → 'Outcome' column
 - One Hot Encoding
- 3: Exploratory Data Analysis (EDA)
 - SQL Queries
 - Visualization w/ Matplotlib, Seaborn

- 4: Launch Sites Locations Analysis
 - Folium
- 5: Launch Records Dashboard
 - Plotly Dash
- 6: Predictive Analysis
 - Logistic Regression
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DATA COLLECTION

Overview

- SpaceX API
 - Historical launch data from SpaceX REST API
- Web Scraping
 - Historical launch data from Wikipedia page
 - <u>'List of Falcon 9 and Falcon</u>
 <u>Heavy Launches'</u>



Example of launch records in HTML format prior to web scraping

DATA COLLECTION

SpaceX API

- Task 1:
 - Requested and parsed the SpaceX launch data using the GET request
- Task 2:
 - Filtered the dataframe to only include Falcon 9 launches
- Task 3:
 - Replaced null values in 'PayloadMass' column with mean 'PayloadMass' value

DATA COLLECTION

Web Scraping (Beautiful Soup)

- Task 1:
 - Requested the Falcon 9 Launch Wiki page from its URL
- Task 2:
 - Extracted all column/variable names from the HTML table header
- Task 3:
 - Created a dataframe by parsing the launch HTML tables

DATA WRANGLING

- Task 1:
 - Calculated the number of launches on each site
- Task 2:
 - Calculated the number and occurrence of each orbit
- Task 3:
 - Calculated the number and occurrence of mission outcome of the orbits
- Task 4:
 - Created a landing outcome label from 'Outcome' column
 - 'Class' → 1 if successful, 0 if unsuccessful

| True ASDS | 41 | |
|----------------|--------|-------|
| None None | 19 | |
| True RTLS | 14 | |
| False ASDS | 6 | |
| True Ocean | 5 | |
| False Ocean | 2 | |
| None ASDS | 2 | |
| False RTLS | 1 | |
| Name: Outcome, | dtype: | int64 |
| | | |

Landing Outcomes, n=90 ~66% Success Rate

EDA

SQL

- Loaded data into IBM DB2 instance
- Completed queries:
 - 1: Displayed the names of unique launch sites
 - 2: Displayed 5 records where launch sites began with the string 'CCA'
 - 3: Displayed the total payload mass carried by boosters launched by NASA (CRS)
 - 4: Displayed the average payload mass carried by booster version F9 v1.1
 - 5: Listed the date when the first successful landing outcome in ground pad was achieved
 - 6: Listed the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
 - 7: Listed the total number of successful and failure mission outcomes
 - 8: Listed the names of the booster versions which have carried the maximum payload mass
 - 9: Listed the failed landing outcomes in drone ship, along with their booster versions and launch site names for the year
 2015
 - o 10: Ranked the count of landing outcomes between the date 2010-06-04 and 2017-03-20 in descending order

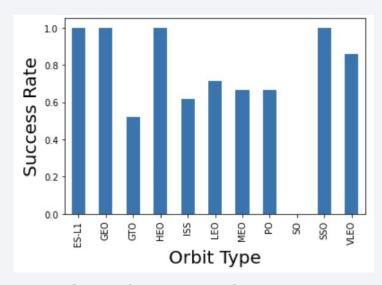
mission_outcometotal_numberFailure (in flight)1Success99Success (payload status unclear)1

SQL Query #7

EDA

Data Visualization

- Scatter Point Charts (w/ landing outcome overlaid)
 - Flight Number x Launch Site
 - Payload x Launch Site
 - Flight Number x Orbit Type
 - Payload x Orbit Type
- Bar Chart
 - Orbit Type x Success Rate
- Line Chart
 - Year x Success Rate



Bar Chart: Orbit Type x Success Rate

LAUNCH SITES LOCATIONS ANALYSIS

Folium

- Task 1:
 - Marked all launch sites on a map
- Task 2:
 - Marked the successful/unsuccessful launches for each site on the map
- Task 3:
 - Calculated the distances between a launch site to its proximities



Interactive map showing successful/unsuccessful launches at the CCAFS SLC-40 launch site

LAUNCH RECORDS DASHBOARD

Plotly Dash

- Task 1:
 - Added a Launch Site drop-down input component
- Task 2:
 - Added a pie chart showing Success Rate based upon selected Launch Site
- Task 3:
 - Added a range slider to select Payload Mass
- Task 4:
 - Added a scatter chart showing Payload Mass x Landing Outcome

PREDICTIVE ANALYSIS

- Task 1:
 - Created a NumPy array from the 'Class' column and assigned it to the variable Y
- Task 2:
 - Standardized the data and assigned it to the variable X
- Task 3:
 - Split the X and Y data into training and test sets, with test size as 20%
- Task 4:
 - Fit the training data to the following models
 - Logistic Regression
 - Support Vector Machine
 - Decision Tree Classifier
 - K Nearest Neighbors Classifier

- Task 5:
 - Used GridSearchCV to identify best parameters for each model
- Task 6:
 - Calculated the accuracy on the test data using the optimal parameters for each model
- Task 7:
 - Plotted confusion matrices for each model's predictions
- Task 8:
 - Compared the accuracies of each model



Launch Site Names

```
In [5]:
 %sql
 SELECT DISTINCT LAUNCH_SITE
 FROM SPACEXTBL;
* ibm_db_sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.clogj3sd0tgtu0lqde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out[5]:
   launch_site
 CCAFS LC-40
CCAFS SLC-40
  KSC LC-39A
  VAFB SLC-4E
```

Launch Site Names Begin with 'CCA'

```
In [9]:
 %%sql
 SELECT LAUNCH_SITE
 FROM SPACEXTBL
 WHERE LAUNCH_SITE LIKE 'CCA%'
 LIMIT 5;
* ibm_db_sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.c1ogj3sd0tgtu0lqde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out[9]:
 launch_site
CCAFS LC-40
CCAFS LC-40
CCAFS LC-40
CCAFS LC-40
CCAFS LC-40
```

Total Payload Mass (NASA CRS Boosters)

```
In [10]:
 %%sql
 SELECT SUM(PAYLOAD_MASS__KG_)
 FROM SPACEXTBL
 WHERE Customer = 'NASA (CRS)';
 * ibm_db_sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.c1ogj3sd0tgtu0lgde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out[10]:
45596
```

Average Payload Mass by F9 v1.1 Booster

```
In [11]:
 %%sql
 SELECT AVG(PAYLOAD_MASS__KG_)
 FROM SPACEXTBL
 WHERE Booster_Version LIKE 'F9 v1.0%';
* ibm_db_sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.c1ogj3sd0tgtu0lqde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out[11]:
340
```

First Successful Ground Landing Date

```
In [30]:
 %sql
 SELECT MIN(Date)
 FROM SPACEXTBL
 WHERE Landing__Outcome = 'Success (ground pad)';
* ibm_db_sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.c1ogj3sd0tgtu0lqde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out [30]:
2015-12-22
```

Successful Drone Ship Landing with Payload Between 4000 and 6000 Kg

```
In [33]:
 %sal
 SELECT BOOSTER_VERSION
 FROM SPACEXTBL
 WHERE LANDING OUTCOME = 'Success (drone ship)'
     AND 4000 < PAYLOAD_MASS__KG_ < 6000;
* ibm_db_sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.c1ogj3sd0tgtu0lqde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out[33]:
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
```

Total Number of Successful and Failure Mission Outcomes

```
In [44]:
 %sql
 SELECT MISSION_OUTCOME, COUNT(MISSION_OUTCOME) AS TOTAL_NUMBER
 FROM SPACEXTBL
 GROUP BY MISSION OUTCOME;
 * ibm db sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.clogj3sd0tgtu0lgde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out [44]:
            mission_outcome total_number
              Failure (in flight)
                                       99
                     Success
Success (payload status unclear)
```

Boosters Carrying Maximum Payload

```
In [57]:
 %%sql
 SELECT DISTINCT BOOSTER_VERSION
 FROM SPACEXTBL
 WHERE PAYLOAD MASS KG = (
     SELECT MAX(PAYLOAD MASS KG)
     FROM SPACEXTBL);
* ibm_db_sa://xcq80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.clogj3sd0tgtu0lqde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out[57]:
 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
```

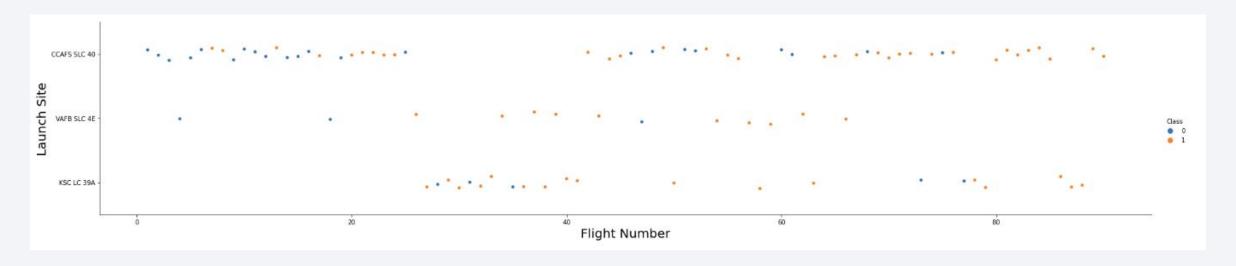
2015 Launch Records

```
In [62]:
 %%sql
 SELECT LANDING OUTCOME, BOOSTER VERSION, LAUNCH SITE
 FROM SPACEXTBL
 WHERE Landing_Outcome = 'Failure (drone ship)'
     AND YEAR(DATE) = 2015;
* ibm db sa://xcq80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.clogj3sd0tgtu0lgde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out [62]:
landing__outcome booster_version
                                   launch_site
 Failure (drone ship)
                     F9 v1.1 B1012 CCAFS LC-40
 Failure (drone ship)
                     F9 v1.1 B1015 CCAFS LC-40
```

Ranking Landing Outcomes (2010-06-04 – 2017-03-20)

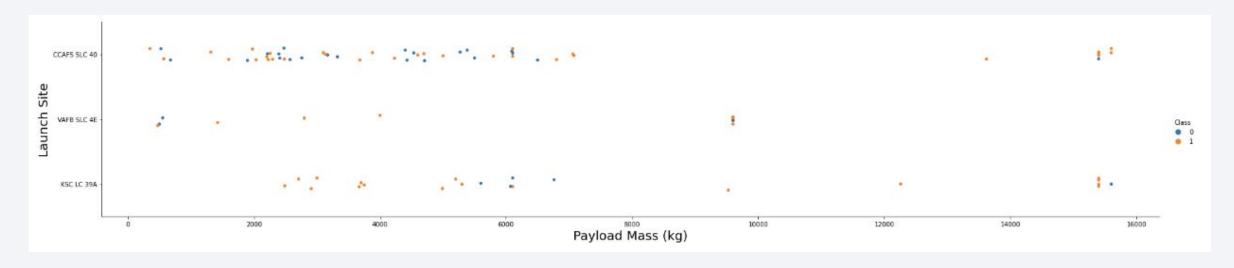
```
In [70]:
 %sql
 SELECT LANDING__OUTCOME, COUNT(LANDING__OUTCOME) AS TOTAL_NUMBER
 FROM SPACEXTBL
 WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20'
 GROUP BY LANDING__OUTCOME
 ORDER BY TOTAL_NUMBER DESC
* ibm_db_sa://xcg80731:***@ba99a9e6-d59e-4883-8fc0-d6a8c9f7a08f.c1ogj3sd0tgtu0lqde0
0.databases.appdomain.cloud:31321/bludb
Done.
Out[70]:
   landing_outcome total_number
                               10
          No attempt
   Failure (drone ship)
                                5
  Success (drone ship)
                                5
    Controlled (ocean)
                                3
 Success (ground pad)
                                3
    Failure (parachute)
                                2
  Uncontrolled (ocean)
Precluded (drone ship)
```

Flight Number x Launch Site



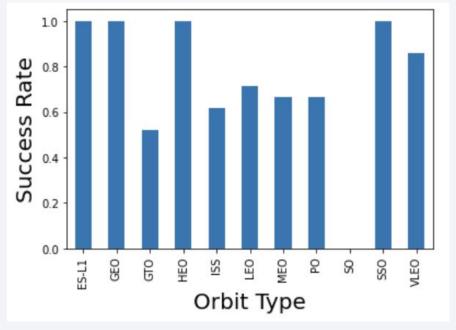
- The earliest flights tended to fail more often, while the latest flights tended to succeed
- The CCAFS SLC 40 launch site had the majority of launches

Payload x Launch Site



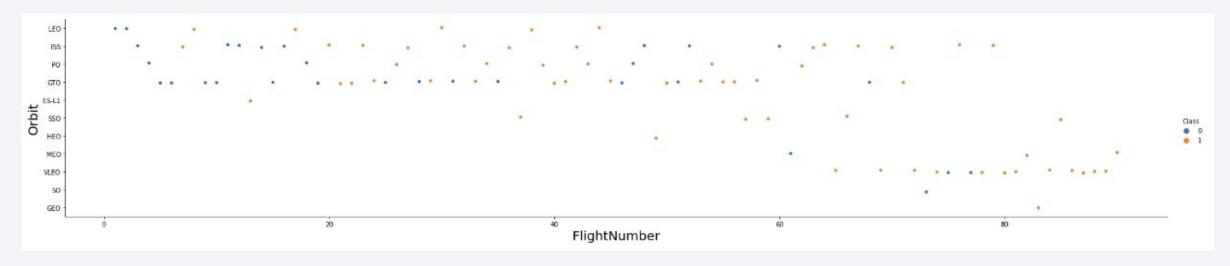
- Overall, a higher payload mass appears to coincide with a higher success rate
- The majority of launches with payload mass over 8000 kg were successful
- KSC LC 39A had a 100% success rate with payload mass under ~5500 kg

Orbit Type x Success Rate



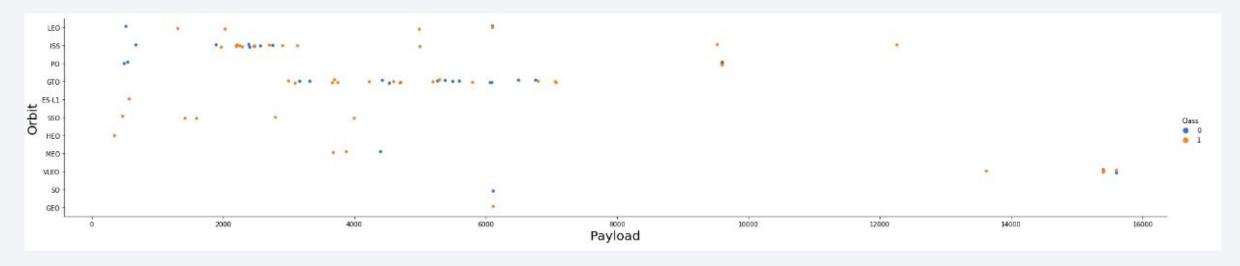
- ES-L1, GEO, HEO, and SSO all had a 100% success rate
- SO had a 0% success rate

Flight Number x Orbit Type



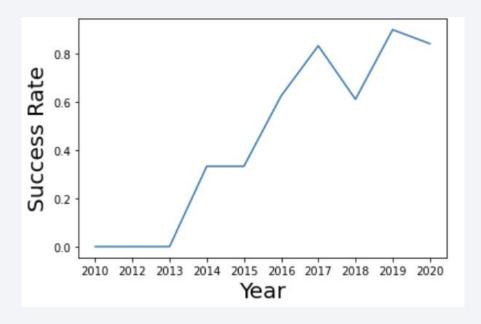
- LEO showed a general increase in success rate as the number of flights increased
- VLEO showed a high rate of success at a larger number of flights

Payload x Orbit Type



- GTO appeared to have a lower rate of success at higher payloads (>5000 kg)
- SSO had a 100% success rate at payloads less than 4000 kg

Launch Success Yearly Trend

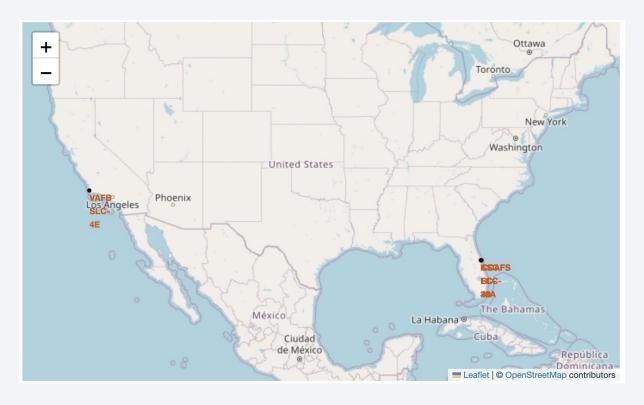


- Overall, success rate appeared to increase year over year
- 2014 to 2015 showed little to no increase in success rate
- Success rate decreased from 2017 to 2018 and from 2019 onwards



LAUNCH SITES LOCATIONS ANALYSIS

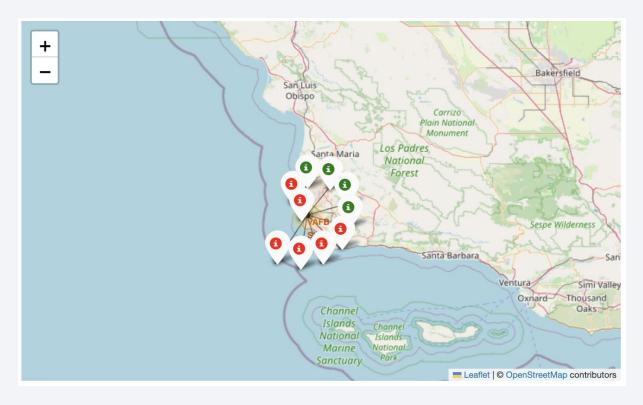
Launch Site Location Markers



The map above shows the locations (marked in red) of the launch sites. Note that these locations are as far south as possible (in the U.S.) and are close to the coast.

LAUNCH SITES LOCATIONS ANALYSIS

Color-Labeled Launch Outcomes



The map above shows the outcomes of launches at VAFB SLC 4E. Red represents an unsuccessful landing, while green represents a successful landing.

LAUNCH SITES LOCATIONS ANALYSIS

Launch Site Proximities



The map above shows the proximity of the CCAFS SLC 40 launch site to several landmarks, such as the closest railway (1.28 km), highway (0.58 km), and coastline (0.86 km).



LAUNCH RECORDS DASHBOARD

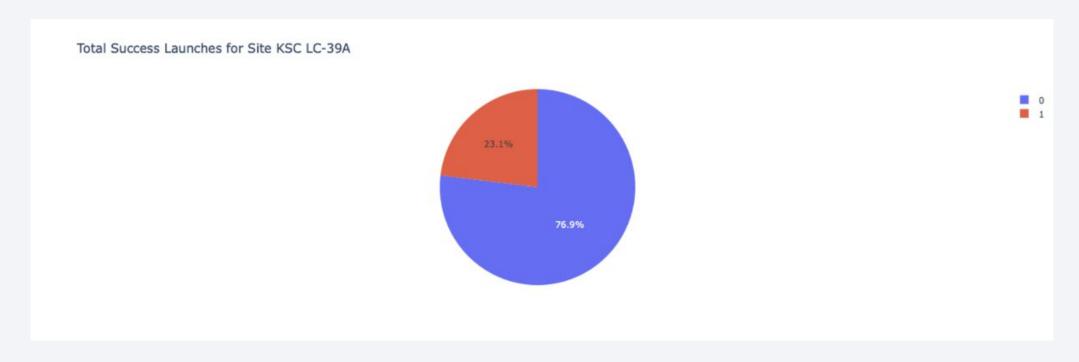
Launch Success Count for all Sites



- The pie chart above shows the launch success count for all launch sites
- Here, we can see that KSC LC-39A had the most successful launches

LAUNCH RECORDS DASHBOARD

Launch Site with the Highest Launch Success Ratio



- The pie chart above shows the launch success ratio for KSC LC-39A
- Here, we can see that this site had a 76.9% success rate

LAUNCH RECORDS DASHBOARD

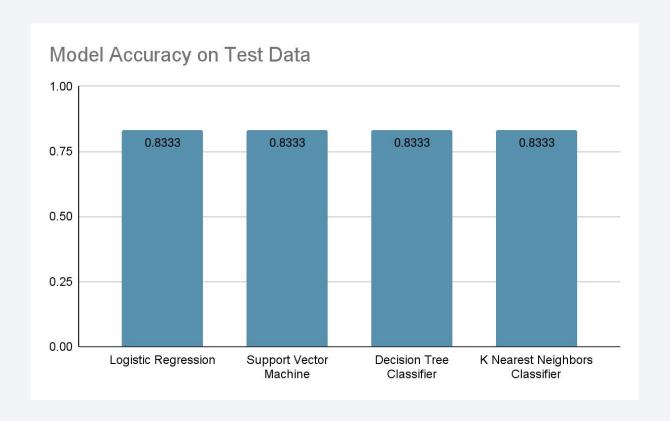
Payload x Launch Outcome for all Launch Sites



- The chart above shows the relationship between Payload Mass and Outcome
- Payloads less than 5300 kg tended to have a higher success rate
- Payloads greater than 5300 kg tended to have lower success rate

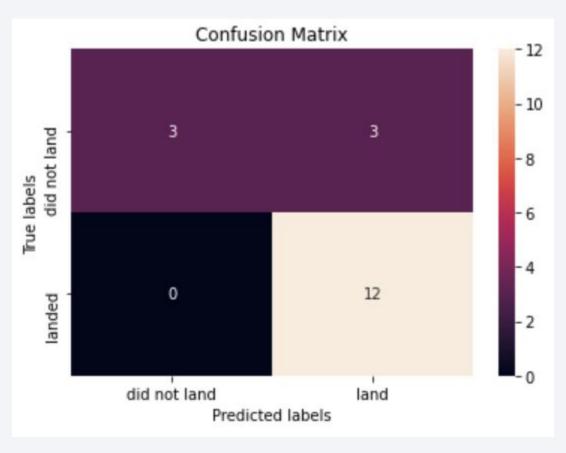


CLASSIFICATION ACCURACY



The chart above shows that each model yielded an accuracy score of ~83.33%

CONFUSION MATRIX



The confusion matrix above represents the performance of each model, as they all performed identically. Here, we can see that each model correctly yielded a true positive 12 times, and a true negative 3 times. However, the model also yielded 3 false positives.

CONCLUSION

- All models yielded the same accuracy on the test set (~83.33%)
 - The Decision Tree Classifier yielded the highest accuracy on the training set (~88.88%)
- SpaceY can predict a successful landing of a SpaceX first stage booster with 83.33% accuracy
- Potential future work can be done to improve this accuracy
 - Further collection of data to increase training and testing dataset sizes
 - Revisit the parameter tuning of the Decision Tree Classifier, as it yielded the highest accuracy on the training set
 - Employ more powerful classifier models, such as <u>LightGBM</u>

APPENDIX

- Notebooks used to complete this project:
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/1-Data-Collection-API.ipynb
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/2-Data-Collection-with-Web-Scraping.ipynb
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/3-Data-Wrangling.ipynb
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/4-EDA-with-S QL.ipynb
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/5-EDA-with-Visualization.ipynb
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/6-Launch-Sites-Locations-Analysis-with-Folium.ipynb
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/7-Interactive-Dashboard-with-Plotly-Dash.py
 - https://github.com/ConradKleykamp/IBM-Applied-Data-Science-Capstone/blob/main/8-Machine-Learning-Prediction.ipynb

