# Data Science CapstoneSpaceX

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# **OUTLINE**

- INTRODUCTION
- SUMMARY
- METHODOLOGY
- RESULTS
- CONCLUSION
- APPENDIX





# **Project Background:**

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.

#### **Problem:**

Predicting if the Falcon 9 first stage will land successfully.

#### **EXECUTIVE SUMMARY**

# Methodologies Summary

The research attempts to identify the factors for a successful rocket landing. To make this determination, the following methodologies where used:

- Collect data using SpaceX REST API and web scraping techniques
- Wrangle data to create success/fail outcome variable
- Explore data with data visualization techniques, considering the following factors: payload, launch site, flight number and yearly trend
- Analyze the data with SQL, calculating the following statistics: total partial, payload range for successful launches, and total # of/successful and failed outcomes
- Explore launch site success rates and proximity to geographical markers
- Visualize the launch sites with the most success and successful pay oad ranges
- Build Models to predict landing outcomes using logistic regression, support vector machine (SVM), decision tree and K-nearest neighbor (KNN)

### **EXECUTIVE SUMMARY**

Results Summary

#### **Exploratory Data Analysis:**

- Launch success has improved over time
- KSC LC-39A has the highest success rate among landing sites
- Orbits ES-L1, GEO, HEO, and SSO have a 100% success rate

#### **Visualization/Analytics:**

• Most launch sites are near the equator, and all are close to the coast

#### **Predictive Analytics**

• All models performed similarly on the test set. The decision tree model



- Data Collection
- Data Wrangling
- EDA Data visualization.
- EDA Sql data
- Machine Learning Prediction
- Interactive Visual Analytics with Folium
- Build a Dashboard Using Plotly Dash



- Data Collection
- Request data from SpaceX API (rocket launch data)
- Decode response using .json() and convert to a dataframe using .json\_normalize()
- Request information about the launches from SpaceX API using custom functions
- Create dictionary from the data
- Create dataframe from the dictionary
- Filter dataframe to contain only Falcon 9 launches
- Replace missing values of Payload Mass with calculated mean
- Export data to csv file





Data Collection

**Scraping:** 

Wikipedia has a page that has tables of data about SpaceX launches

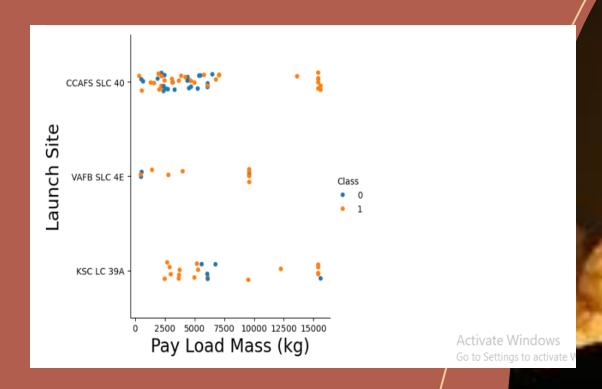
• Link: https://github.com/REDH/WAN0/IBM-Data-Science-Capstone-SpaceX /Web Scraping.ipynb

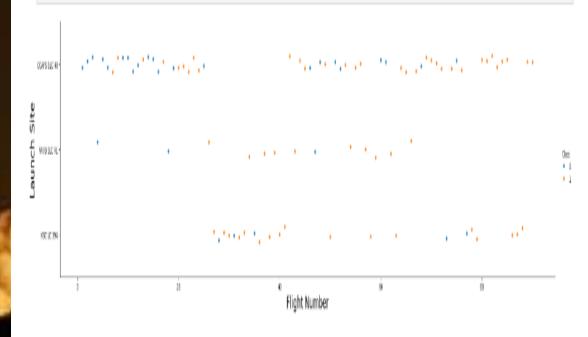


- Data Wrangling
  - Convert outcomes into 1 for a successful landing and 0 for an unsuccessful landing
  - Filtering the data
  - Dealing with missing values
  - Using One Hot Encoding to prepare the data to a binary sification
  - Link: <a href="https://github.com/REDH/WAN0/IBM-Data-science-Capstone-SpaceX/Data-wrangling.ipynb">https://github.com/REDH/WAN0/IBM-Data-science-Capstone-SpaceX/Data-wrangling.ipynb</a>

• EDA with Data Visualization

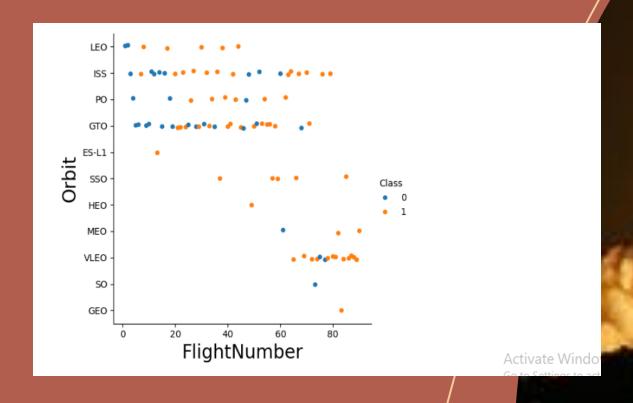
The following charts were created to look at Launch Site trends

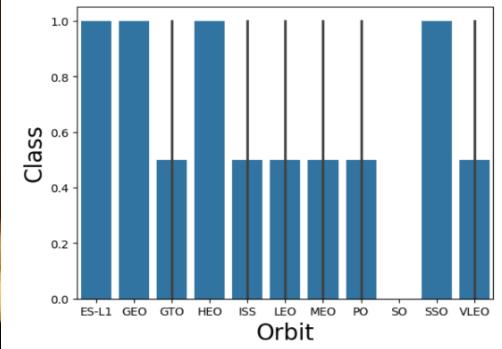




• EDA with Data Visualization

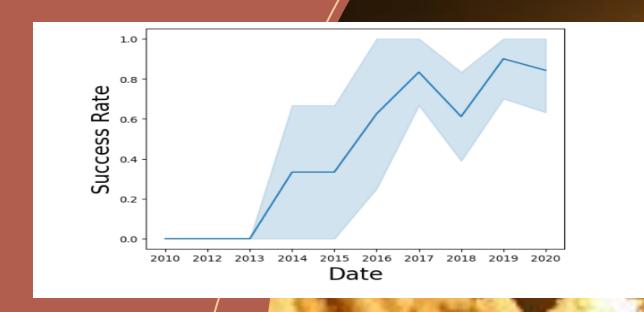
The following charts were created to look at Orbit Type trends





• EDA with Data Visualization

The following chart was created to look at trends based on time



GitHub URL (EDA with Data Visualization): : <a href="https://github.com/REDHWANO/IBM-Data-Science-Capstone-SpaceX/EDA-Data visualization.ipynb">https://github.com/REDHWANO/IBM-Data-Science-Capstone-SpaceX/EDA-Data visualization.ipynb</a>



• EDA Sql data

Queries were written to extract information about:

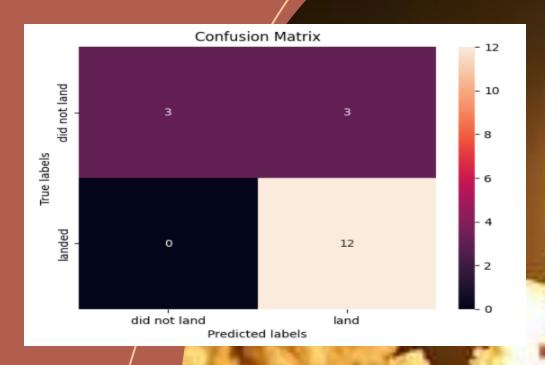
- Launch sites
- Payload masses
- Dates
- Booster types
- Mission outcomes

 GitHub URL (EDA/Sql data) m/RADHWAN0/IBM-Data

Science-Capstone-SpadeX%20/EDA-Sql%20gata.ipynb

- Machine Learning Prediction
  - The dataset was split into training and testing sets.
  - Logistic Regression, SVM (Support Vector Machine), Decision Tree, and KNN (k-Nearest Neighbørs) machine learning model were trained on the training data set.
  - Hyper-parameters were evaluated using GridSearchCV and the best was selected using '.best\_params\_'.
  - Using the best hyper-parameters, each of the four models were scored on accuracy by using the testing data set.oad

• Machine Learning Prediction



• GitHub URL (Machine Learning Prediction): https://github.com/RADHWANO/BM-Data-Science-Capstone-SpaceX%20/Machine Learning Prediction.ipynb

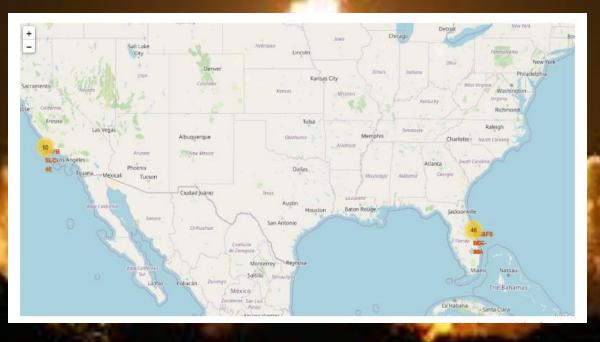
- Interactive Visual Analytics with Folium

  Summarize what map objects such as markers, circles, lines, etc. you created and added to a folium map
  - Markers were added for launch sites and for the NASA Johnson Space Center
  - Circles were added for the launch sites.
  - Lines were added to show the distance to the nearby featu
    - Distance from CCAFS LC-40 to the coastline
    - Distance from CCAFS/LC-40 to the rail line
    - Distance from CCAFS LC-40 to the perimeter road

• Interactive Visual Analytics with Folium/

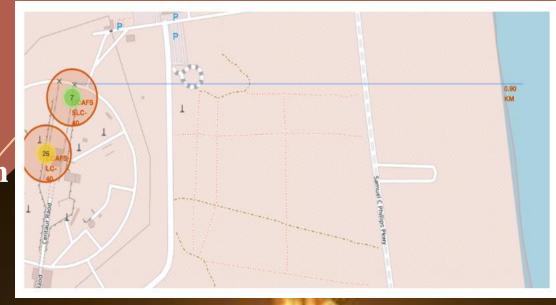


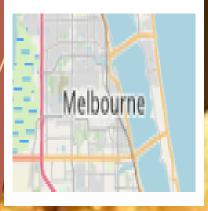




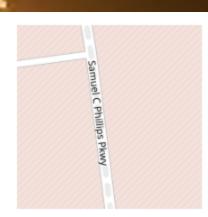
Interactive Visual Analytics with Folium









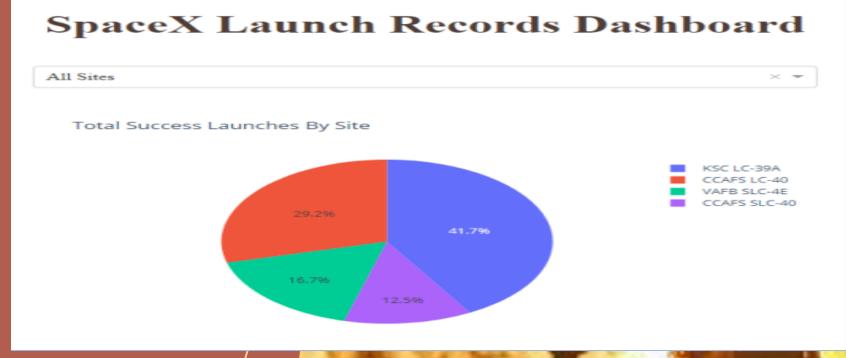


• GitHub URL (Interactive Visual Analytics with

Folium): <a href="https://github.com/R/ADHWAN0/IBM-Data-Science-Capstone-SpaceX%20/Interactive%20Visual%20Analytics%20with%20Folium.ipynb">https://github.com/R/ADHWAN0/IBM-Data-Science-Capstone-SpaceX%20/Interactive%20Visual%20Analytics%20with%20Folium.ipynb</a>

- Build a Dashboard Using Plotly Dash
  - The input dropdown is used to select one or all launch sites for the pie chart and scatterplot.
  - The pie chart displays one of two things:
  - For All Sites the distribution of successful Falcon 9 set stage landings between the sites
  - For One Site the distribution of successful and faile lcon 9 first stage landings for that site
  - The input slider is used to filter the payload masses for the seatterplot. The scatterplot displays the distribution of Falcon 9 first stage landings split by payload mass, mission outcome and by booster version category.

• Build a Dashboard Using Plotly Dash



• GitHub URL (Build a Dashboard Using Plotly Dash): https://github.com/RADHWANG/IBM-Data-Science-Capstone-SpaceX%20/plotly\_dash.py

#### **RESULTS**

# **Exploratory Data Analysis:**

- Launch success has improved over time
- KSC LC-39A has the highest success rate among landing sites
- Orbits ES-L1, GEO, HEO, and SSO have a 100% success rate

# **Visualization / Analytics:**

Most launch sites are near the equator, and all are close to the past

# **Predictive Analytics**

• All models performed similarly on the test set. The decision treatmodel slightly outperformed when looking at .best\_score\_



- SpaceX does not have a perfect track record of Falcon 9 first stage landing outcomes.
- SpaceX's Falcon 9 first stage landing outcomes have been trending towards greater success as more launches are made.
- The machine learning models can be used to predict future the ceX Falcon 9 first stage landing outcomes



#### **APPENDIX**

- Data Sets (.csv files):
- GitHub URL (CSV 1):

  <a href="https://github.com/RADHWAN0/IBM-Data-Sgience-Capstone-SpaceX/dataset\_part\_1.csv">https://github.com/RADHWAN0/IBM-Data-Sgience-Capstone-SpaceX/dataset\_part\_1.csv</a>
- GitHub URL (CSV 2):

  https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX /dataset\_part\_2.csv
- GitHub URL (CSV 3):

  <a href="https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX/dataset\_part\_3.csv">https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX/dataset\_part\_3.csv</a>
- GitHub URL (SpaceX):

  https://github.com/RADHWAN0/IBM/Data-Science-Canstone-SpaceX /Spacex.csv
- GitHub URL (Launch Geo):

  https://github.com/RADHWAN0/IBM-Data-Science-Cassione-SpaceX /plotly\_dash.py
- GitHub URL (Plotly Dash):

  https://github.com/RADHWAN0/IBM-Data-Science-Capatone-SpaceX /plotly\_dash.py/

#### **APPENDIX**

# Jupyter Notebooks and Dashboard Python File

- GitHub URL (Data Collection): <a href="https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX/Data-collection-api.ipynb">https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX/Data-collection-api.ipynb</a>
- GitHub URL (Web Scraping): <a href="https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX/Web Scraping.ipynb">https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX/Web Scraping.ipynb</a>
- GitHub URL (Data Wrangling): https://github.com/RADHWAN0/IBM-Data-Science-Capstone-SpaceX /Data wrangling.ipynb
- GitHub URL (EDA Sql data): https://github.com/RADHWANO/IBM Data-Science-Capstone-SpaceX /EDA-Sql data.ipynb
- GitHub URL (EDA with Data Visualization): <a href="https://github.com/RADHWANO/IBM-Data-science-Capstone-SpaceX/EDA-Data-visualization.jpwnb">https://github.com/RADHWANO/IBM-Data-Science-Capstone-SpaceX/EDA-Data-visualization.jpwnb</a>
- GitHub URL (Folium Maps): https://gubub.com/RADHWANO/IBM-Data-Science-Capstone-SpaceX /Interactive Visual Analytics with Folium in your
- GitHub URL (Plotly Dash): https://github.com/RADHWANO/IBM-Data-Science-Capstone-SpaceX/plotly\_dash.py
- CitHub IIRI (Machine Learning). https://github.com/PADHWANO/IBM-Data-Science