# IBM Professional Certificate Data Science Winning Space Race with Data Science

Omer Fatih Kaya 15 May 2023 https://github.com/aeromars/data\_science\_capstone/

#### Capstone Project

#### **Outline**

**Executive Summary** 

Introduction

Methodology

Results

Conclusion

Appendix

#### **Executive Summary**

- Major success came through in 2015 indicating technological advances. These
  rockets are in development phase and success rate expected to increase over time
  as flight numbers increase.
- Launch site distribution are CCAFS SLC 40 (61.11%), KSC LC 39A (24.44%), and VAFB SLC 4E (14.44%), but VAFB SLC 4E discontinued. Since these sites are in similar ranges to equator, launch sites are almost indifferent for space projectile motion as shown in data.
- Orbit distribution are GTO (27), ISS (21), VLEO (14), PO (9), LEO (7), SSO (5), MEO (3), ES-L1 (1), GEO (1), HEO (1), SO (1).
- Rockets have version numbers indicating technological advances. As technological advances increase, other factors such as orbit types, launch sites are expected to be less importance. In another words, as flight numbers increase, other factors mentioned are less importance.

#### **Executive Summary**

• Prediction model for rocket first stage success has accuracy of 83.33%. Remarkably, all prediction models (Logistic Regression, Support Vector Machine, Decision Tree, K-Nearest Neighbors) have same accuracy of 83.33%.

#### Introduction

- Analyzing SpaceX, specifically rocket first stage, makes SpaceX competitive for space cargo. Rocket first stage includes rocket reusability and rocket landing success rate. Market prices for space cargo range from \$62 million offered by SpaceX to \$165 million by other companies.
- We will analyze SpaceX rocket first stage to conclude whether feasible to compete by incumbents or prospective companies. Measurements are rocket first stage success rate, payload mass, orbit types, launch sites.
- Space launch sites are more important compared to air ports because space projectile motion are different from air projectile motion. Since launch sites are near equator, launch sites are insignificant.

#### Flowchart Diagram/Steps Methodology

Flowchart Diagram/Steps proceeding from 1 to 10.

```
[A] Methodology: Data Collection & Wrangling
        [1] Data Collection
        [2] Data Wrangling
[B] Methodology: EDA & Interactive Visualizations
        [3] EDA
        [4] Features Engineering
[C] Methodology: Predictive Analysis (Classification)
        [5] Standardize Data
        [6] Split Train and Test Datasets
        [7] Machine Learning Algorithms
        [8] Train Models
        [9] Test Models
        [10] Feature Selection
```

## Data Collection Methodology: Data Collection & Wrangling Data Collection

- Two processes selected to collect data
  - SpaceX API (Unofficial)
  - Web Scraping (Wikipedia)

#### Data Collection API Methodology: Data Collection & Wrangling Data Collection

- SpaceX API (Unofficial)
  - https://api.spacexdata.com/v4/launches/past
- IBM Skills Network Dataset
  - https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/datasets/API\_call\_spacex\_api.json
- Dataset contains identification number as references. For example, rocket column
  has no information about rocket but only an identification number. These
  identification numbers must be used to extract more information from different API
  endpoints. Specifically rocket, payloads, launchpad, and cores columns are
  extracted.
- Source
  - aeromars Capstone Data Collection API

#### Web Scraping Methodology: Data Collection & Wrangling Data Collection

- requests, pandas, re, unicodedata, BeautifulSoup packages used for web scraping.
- Extract Falcon 9 launch records HTML table from Wikipedia (List of Falcon 9 and Falcon Heavy launches on 9 June 2019).
  - https://en.wikipedia.org/w/index.php?title=List\_of\_Falcon\_9\_and\_Falcon\_Heavy\_launches&oldid=1027686922
- Parse table and convert it to Pandas dataframe.
- Source
  - aeromars Capstone Data Collection via Web Scraping

### Data Wrangling Methodology: Data Collection & Wrangling Data Wrangling

- Pandas and Numpy used for performing basic statistics.
- Replace missing values with mean.
- Create binary (Success = 1, Fail = 0) "Class" column from "Outcome" category for training supervised models label. In another words create a landing outcome label from "Outcome" column.
- Filter only Falcon 9 rockets from dataset.

#### **EDA Basic Statistics Methodology: EDA & Interactive Visualizations EDA**

- Aggregate Site Launches
- Aggregate Orbit Types
- Aggregate Mission Outcome per Orbit Type
- Aggregate Success Rate

### **EDA** with Data Visualization Methodology: **EDA** & Interactive Visualizations **EDA**

- Examining relationships:
  - Flight Number and Launch Sites
  - Payload and Launch Sites
  - Success Rate and Orbit Types
  - Flight Number and Orbit Types
  - Payload and Orbit Types
  - Average Success Rate (Year and Average Success Rate)

- Task 1
  - %sql SELECT DISTINCT(Launch Site) FROM SPACEXTBL;
- Task 2
  - %sql SELECT \* FROM SPACEXTBL WHERE Launch\_Site LIKE 'CCA%' LIMIT 5;
- Task 3
  - %sql SELECT SUM("PAYLOAD\_MASS\_\_KG\_") FROM SPACEXTBL WHERE Customer
     = 'NASA (CRS)';
- Task 4
  - %sql SELECT AVG("PAYLOAD\_MASS\_\_KG\_") FROM SPACEXTBL WHERE "Booster\_Version" = 'F9 v1.1';

- Task 5
  - %sql SELECT MIN(Date) FROM SPACEXTBL WHERE "Landing \_Outcome" = 'Success (ground pad)';
- Task 6
  - %sql SELECT "Booster\_Version" FROM SPACEXTBL WHERE "Landing \_Outcome" = "Success (drone ship)" AND ("PAYLOAD\_MASS\_\_KG\_" BETWEEN 4000 AND 6000);
- Task 7
  - %sql SELECT "Mission\_Outcome", COUNT("Mission\_Outcome") FROM SPACEXTBL GROUP BY "Mission\_Outcome";

- Task 8
  - %sql SELECT Booster\_Version FROM SPACEXTBL WHERE PAYLOAD\_MASS\_\_KG\_ = (SELECT MAX(PAYLOAD MASS\_KG\_) FROM SPACEXTBL);
- Task 9
  - Skills Network

```
%sql SELECT substr(Date, 4, 2) AS Dates, "Booster_Version", "Launch_Site", "Landing _Outcome" FROM SPACEXTBL WHERE "Landing _Outcome" = 'Failure (drone ship)' AND substr(Date, 7, 4) = '2015';
```

Watson

%sql SELECT "Date", "Booster\_Version", "Launch\_Site", "Landing \_Outcome" FROM SPACEXTBL WHERE "Landing \_Outcome" = 'Failure (drone ship)' AND YEAR("Date") = 2015;

- Task 10
  - Skills Network

%sql SELECT "Landing \_Outcome", COUNT(\*) AS Counter FROM SPACEXTBL WHERE substr(Date, 4, 2) BETWEEN '04062010' AND '20032017' GROUP BY "Landing \_Outcome" ORDER BY Counter DESC;

Watson

%sql SELECT "Landing \_Outcome", COUNT(\*) AS Counter FROM SPACEXTBL WHERE "Date" BETWEEN '2010-06-04' AND '2017-03-20' GROUP BY "Landing \_Outcome" ORDER BY Counter DESC;

#### Interactive Map with Folium Methodology: EDA & Interactive Visualizations EDA

- Mark all launch site locations on map. All launch sites located near equator.
- Launch outcomes shown on map for easy statistic on the fly. In our case we only have success or failure outcomes, but we can optionally display other features on map.
- Explore map for possible collateral near launch site locations.

#### Dashboard with Plotly Dash Methodology: EDA & Interactive Visualizations EDA

- Launch sites and success rate graphs are compared for best locations.
- Payload and success rate graph examined for best rocket selections.

## Features Engineering Methodology: EDA & Interactive Visualizations Features Engineering

- Use dummy variables on categorical columns.
- Cast all numeric columns to float64 type.

### Standardize Data Methodology: Predictive Analysis (Classification) Standardize Data

- pandas, numpy, matplotlib, seaborn, preprocessing (sklearn), train\_test\_split, GridSearchCV, LogisticRegression, SVC, DecisionTreeClassifier, KNeighborsClassifier packages used.
- Synchronize two datasets
- Dataset links
  - https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/datasets/dataset\_part\_2csv
  - https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/datasets/dataset part 3.csv

### Standardize Data Methodology: Predictive Analysis (Classification) Standardize Data

Standardize data using sklearn

```
standardization_prep = preprocessing.StandardScaler().fit(X)
X_standard = standardization_prep.transform(X)
X = X standard
```

### Split Train and Test Datasets Methodology: Predictive Analysis (Classification) Split Train and Test Datasets

• Split dataset into train and test datasets using sklearn's train\_test\_split function with parameters test size = 0.2 and random state = 2.

```
X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.2, random_state = 2)
```

## Machine Learning Algorithms Methodology: Predictive Analysis (Classification) Machine Learning Algorithms

- Four algorithms used:
  - Logistic Regression
  - Support Vector Machines
  - Decision Tree
  - K-Nearest Neighbors

#### Train Models Methodology: Predictive Analysis (Classification) Train Models

- Best hyperparameters via sklearn's GridSearchCV with parameters
  - Logistic Regression parameters parameters = {"C":[0.01,0.1,1],'penalty':['l2'], 'solver':['lbfgs']}# |1 | lasso |2 ridge
  - Support Vector Machines parameters

```
parameters = {'kernel':('linear', 'rbf', 'poly', 'rbf', 'sigmoid'), 'C': np.logspace(-3, 3, 5), 'gamma':np.logspace(-3, 3, 5)}
```

Decision Tree

```
parameters = {'criterion': ['gini', 'entropy'], 'splitter': ['best', 'random'], 'max_depth': [2*n for n in range(1,10)], 'max_features': ['auto', 'sqrt'], 'min_samples_leaf': [1, 2, 4], 'min_samples_split': [2, 5, 10]}
```

#### Train Models Methodology: Predictive Analysis (Classification) Train Models

K-Nearest Neighbors

```
parameters = {'n_neighbors': [1, 2, 3, 4, 5, 6, 7, 8, 9, 10], 'algorithm': ['auto', 'ball tree', 'kd tree', 'brute'], 'p': [1,2]}
```

- Fit the models using training data
- Training set accuracy via best\_score\_ function.

#### Test Models Methodology: Predictive Analysis (Classification) Test Models

- Training set accuracy via best\_score\_ function.
- Test models accuracy using test data.
- Compare all models' accuracy.
- Compare all models' confusion matrices.

#### Feature Selection Methodology: Predictive Analysis (Classification) Feature Selection

• Evaluate which features for optimizing future models. From analyzing data, rocket versions, payload mass, and launches suitable for building future models while remaining features are insignificant.

#### **EDA Basic Statistics Results: EDA & Interactive Visualizations EDA**

- Aggregate Success Rate: 0.6666
- Source:
  - aeromars Capstone Data Wrangling

```
Aggregate Site Launches
CCAFS SLC 40: 55
KSC LC 39A: 22
VAFB SLC 4E: 13
```

```
Aggregate Orbit Types
GTO: 27
ISS: 21
VLEO: 14
PO: 9
LEO: 7
SSO: 5
MEO: 3
ES-L1: 1
GEO: 1
HEO: 1
SO: 1
```

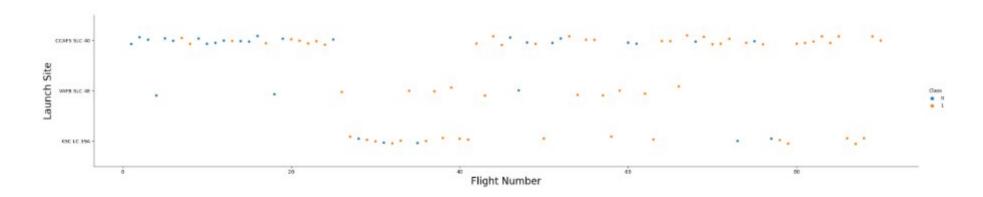
```
Aggregate Mission Outcome per Orbit Types
True ASDS: 41
None None: 19
True RTLS: 14
False ASDS: 6
True Ocean: 5
False Ocean: 2
None ASDS: 2
False RTLS: 1
```

#### Flight Numbers and Launch Sites Results: EDA & Interactive Visualizations EDA

- We see that different launch sites have different success rates. CCAFS LC-40 has a success rate of 60% while KSC LC-39A and VAFB SLC-4E has a success rate of 77%.
- Initial launches from CCAFS SLC are less successful and continuously success rate increase where flight number over 80 are all success. CCAFS SLC site is also highest number of launches occurred.
- VAFB SLC site is the least used site with very high success rate.

#### Flight Numbers and Launch Sites Results: EDA & Interactive Visualizations EDA

- Source:
  - aeromars Capstone EDA via Visualizations

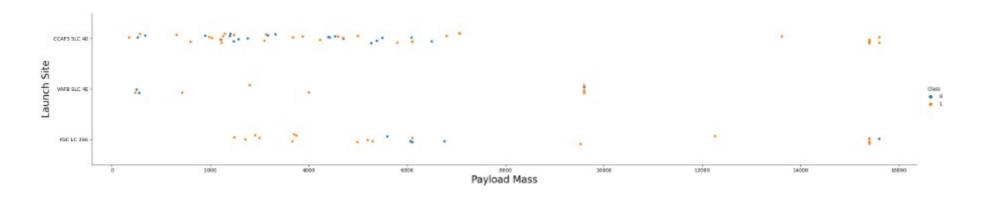


#### Payload and Launch Sites Results: EDA & Interactive Visualizations EDA

- VAFB-SLC launch site there are no rockets launched for heavy payload mass greater than 10000.
- CCAFS SLC has more failures compared to other sites and also most used site.
- There aren't many rocket launches for payload mass greater than 7000.
- Hardly any failures for payload mass over 9000.

#### Payload and Launch Sites Results: EDA & Interactive Visualizations EDA

- Source:
  - aeromars Capstone EDA via Visualizations

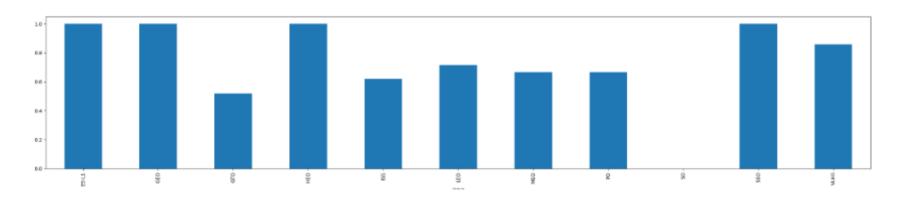


#### Success Rate and Orbit Types Results: EDA & Interactive Visualizations EDA

- ES-L1, GEO, HEO, SSO orbit types have perfect success rate. VLEO orbit type is the next success rate. SO orbit type has no success rate. GTO orbit type have least success rate with over 55%.
- Aggregate success rate for orbit types seems to be over 80%.

#### Success Rate and Orbit Types Results: EDA & Interactive Visualizations EDA

- Source:
  - aeromars Capstone EDA via Visualizations

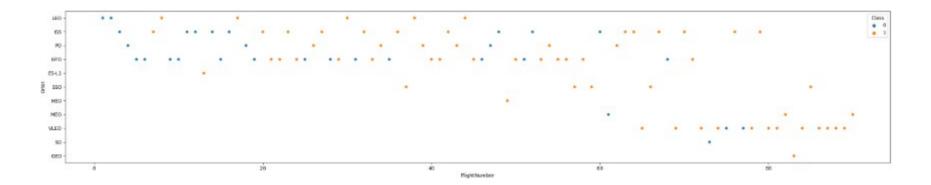


### Flight Number and Orbit Types Results: EDA & Interactive Visualizations EDA

• LEO orbit type have success appears related to number of flights; on the other hand, there seems to be no relationship between flight number when in GTO orbit.

### Flight Number and Orbit Types Results: EDA & Interactive Visualizations EDA

- Source:
  - aeromars Capstone EDA via Visualizations

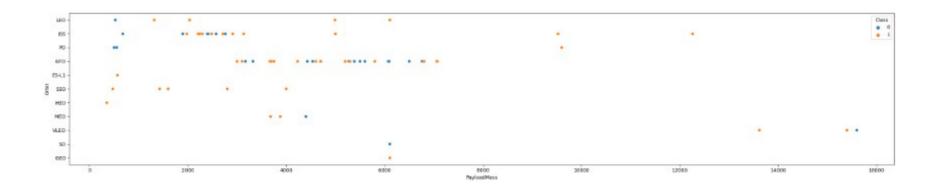


#### Payload and Orbit Types Results: EDA & Interactive Visualizations EDA

- With heavy payloads the successful landing or positive landing rate are more for Polar, LEO, ISS orbit types. However for GTO orbit type we can't distinguish this well as both positive landing rate and negative landing (unsuccessful mission) are both there.
- GTO orbit type have most frequent number of successes and failures followed by ISS orbit type.

### Payload and Orbit Types Results: EDA & Interactive Visualizations EDA

- Source:
  - aeromars Capstone EDA via Visualizations

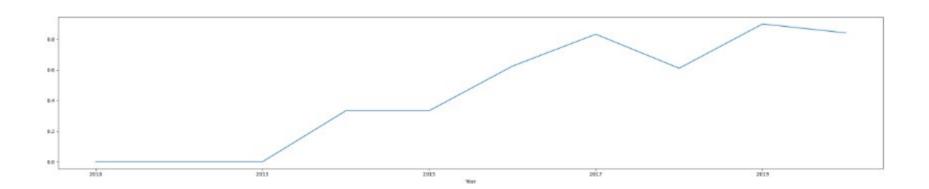


### Average Success Rate (Year and Average Success Rate) Results: EDA & Interactive Visualizations EDA

- You can observe that success rate since 2013 increasing until 2020.
- Average success rate increased reaching max in year 2019 with over 80% value.

### Average Success Rate (Year and Average Success Rate) Results: EDA & Interactive Visualizations EDA

- Source:
  - aeromars Capstone EDA via Visualizations



```
TASK 1

Description:
Display the names of the unique launch sites in the space mission.
Query:
%sql SELECT DISTINCT(Launch_Site) FROM SPACEXTBL;
Result:
CCAFFS LC-40, VAFB SLC-4E, KSC LC-39A, CCAFS SLC-40

TASK 2

Description:
Display 5 records where launch sites begin with the string 'CCA'.
Query:
%sql SELECT * FROM SPACEXTBL WHERE Launch_Site LIKE 'CCA%' LIMIT 5;
Result:
```

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASSKG_	Orbit	Customer	Mission_Outcome	Landi _Outco
04- 06- 2010	18:45:00	F9 v1.0 B0003	CCAFS LC- 40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Faili (parachu
08- 12- 2010	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	Failı (parachu
22- 05- 2012	07:44:00	F9 v1.0 B0005	CCAFS LC- 40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No atter
08- 10- 2012	00:35:00	F9 v1.0 B0006	CCAFS LC- 40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No atter
01- 03- 2013	15:10:00	F9 v1.0 B0007	CCAFS LC- 40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No atter

```
Description:
Display the total payload mass carried by boosters launched by NASA (CRS).
Query:
%sql SELECT SUM("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE Customer = 'NASA (CRS)';
Result:
45596

TASK 4

Description:
Display average payload mass carried by booster version F9 v1.1.
Query:
%sql SELECT AVG("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE "Booster_Version" = 'F9 v1.1';
Result:
2928
```

```
TASK 5
Description:
List the date when the first successful landing outcome in ground pad was achieved.
Querv:
%sql SELECT MIN(Date) FROM SPACEXTBL WHERE "Landing Outcome" = 'Success (ground pad)';
Result:
01-05-2017
TASK 6
Description:
List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less
than 6000.
Query:
%sql SELECT "Booster Version" FROM SPACEXTBL WHERE "Landing Outcome" = "Success (drone ship)" AND
("PAYLOAD MASS KG " BETWEEN 4000 AND 6000);
Result:
F9 FT B1022
F9 FT B1026
F9 FT B1021.2
F9 FT B1031.2
```

```
Description:
List the total number of successful and failure mission outcomes.
Query:
%sql SELECT "Mission_Outcome", COUNT("Mission_Outcome") FROM SPACEXTBL GROUP BY "Mission_Outcome";
Result:
Failure (in flight): 1
Success: 98
Success: 1
Success (payload status unclear): 1
```

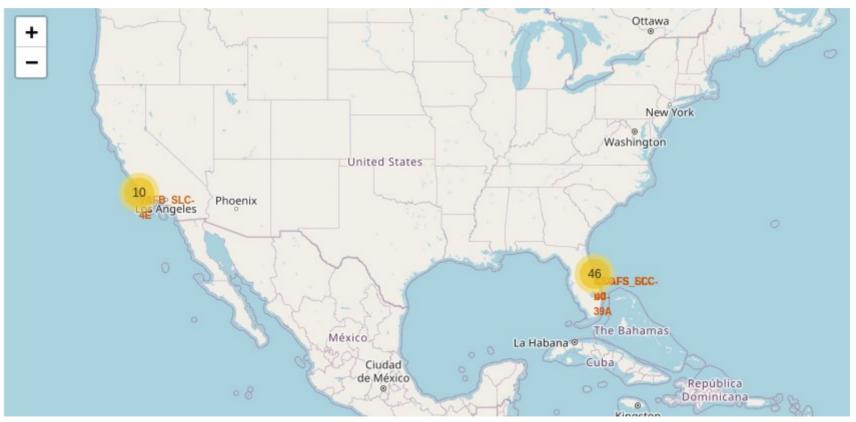
```
TASK 8
Description:
List the names of the booster versions which have carried the maximum payload mass. Use a subquery.
Query:
%sql SELECT Booster Version FROM SPACEXTBL WHERE PAYLOAD MASS KG = (SELECT MAX(PAYLOAD MASS KG) FROM SPACEXTBL);
Result:
F9 B5 B1048.4
F9 B5 B1049.4
F9 B5 B1051.3
F9 B5 B1056.4
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
```

```
TASK 9
Description:
List the records which will display the month names, failure landing outcomes in drone ship, booster versions,
launch site for the months in year 2015.
Query:
# Skills Network solution
%sql SELECT substr(Date, 4, 2) AS Dates, "Booster Version", "Launch Site", "Landing Outcome" FROM SPACEXTBL WHERE
"Landing Outcome" = 'Failure (drone ship)' AND substr(Date, 7, 4) = '2015';
# Watson solution
# %sql SELECT "Date", "Booster Version", "Launch Site", "Landing Outcome" FROM SPACEXTBL WHERE "Landing Outcome"
= 'Failure (drone ship)' AND YEAR("Date") = 2015;
Result:
Dates Booster Version Launch Site
                                       Landing Outcome
       F9 v1.1 B1012 CCAFS LC-40
                                     Failure (drone ship)
01
       F9 v1.1 B1015 CCAFS LC-40
                                      Failure (drone ship)
04
```

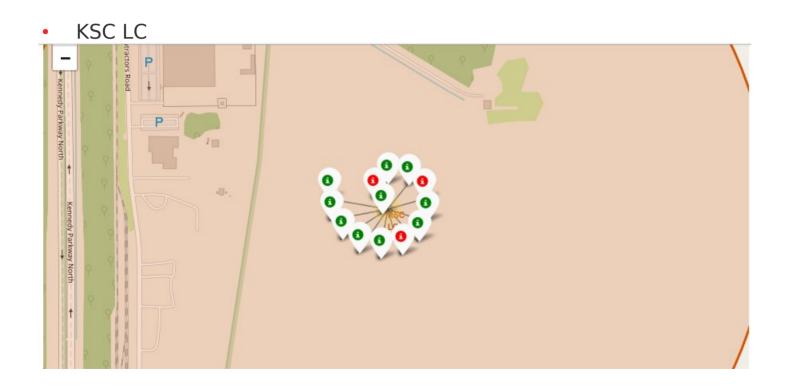
```
TASK 10
Description:
Rank the count of successful landing outcomes between the date 04-06-2010 and 20-03-2017 in decending order.
Query:
# Skills Network solution
%sql SELECT "Landing Outcome", COUNT(*) AS Counter FROM SPACEXTBL WHERE substr(Date, 4, 2) BETWEEN '04062010' AND
'20032017' GROUP BY "Landing Outcome" ORDER BY Counter DESC;
# Watson solution
# %sql SELECT "Landing Outcome", COUNT(*) AS Counter FROM SPACEXTBL WHERE "Date" BETWEEN '2010-06-04' AND
'2017-03-20' GROUP BY "Landing Outcome" ORDER BY Counter DESC;
Result:
Success: 31
No attempt: 11
Success (drone ship): 10
Success (ground pad): 7
Uncontrolled (ocean): 2
Failure (parachute): 2
Controlled (ocean): 2
Precluded (drone ship): 1
No attempt: 1
Failure (drone ship): 1
Failure: 1
```

- Source:
  - aeromars Capstone EDA via SQL

- All launch sites located in southern states due to proximity to equator. For example, there are no launch sites in Chicago.
- Source:
  - aeromars Capstone Data Visualization via Folium



- Easy to compare statistics on the fly with map.
- Source:
  - aeromars Capstone Data Visualization via Folium



**CCAFS SLC** 

CCAFS LC

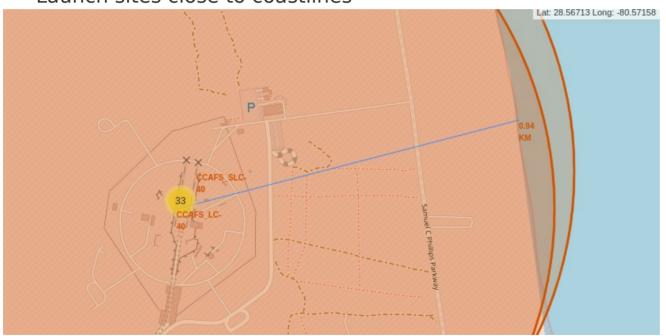


VAFB SLC

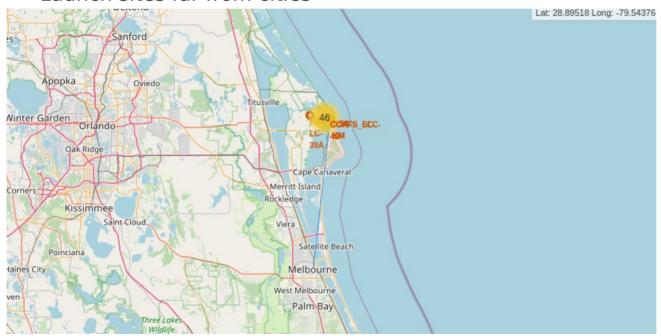


- For safety reasons, all launch sites are in reserved areas far from population.
   Coastlines seem ideal for rockets in test phrase as oceans capture failed rocket debris.
- Source:
  - aeromars Capstone Data Visualization via Folium

Launch sites close to coastlines

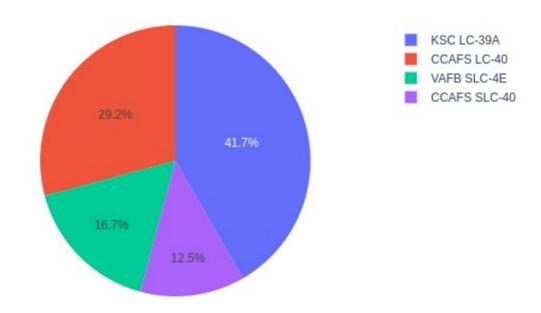


Launch sites far from cities



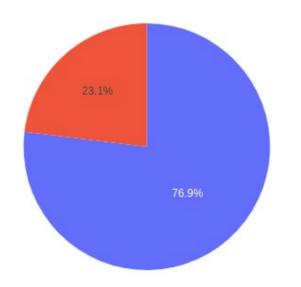
- Aggregate launch success for all launch sites
- Aggregate launch success for specific sites.
- Source:
  - aeromars Capstone Data Visualization via Dash

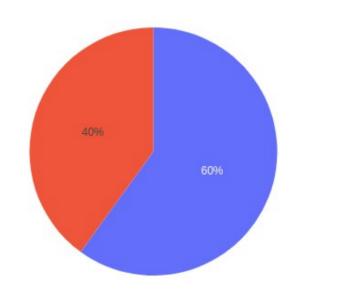
Aggregate Success Launches By Site



- KSC LC-39A has highest success ratio.
- 1 = Success, 0 = Fail

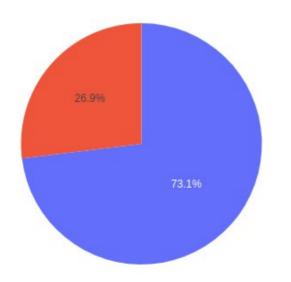
Aggregate Success Launches for launch site KSC LC-39A Aggregate Success Launches for launch site VAFB SLC-4E

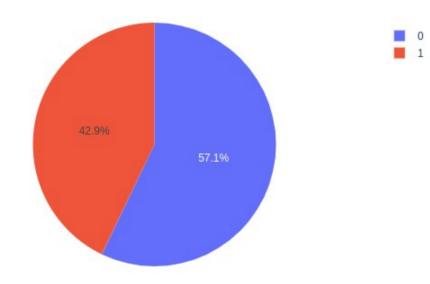




• 1 = Success, 0 = Fail

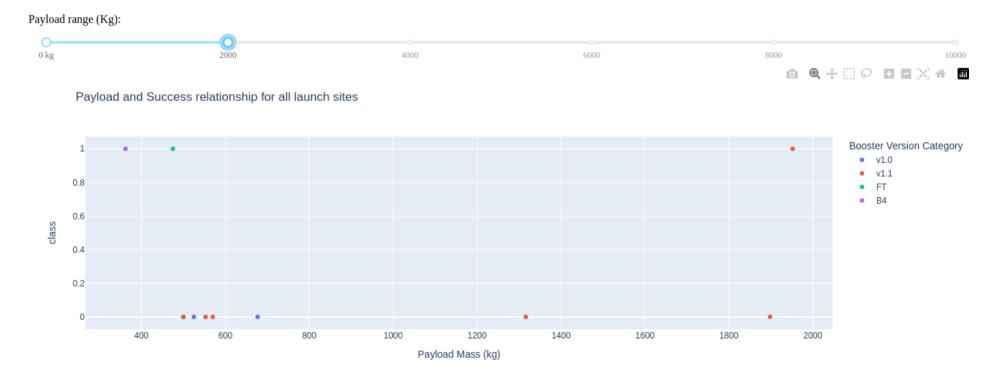
Aggregate Success Launches for launch site CCAFS LC-40 Aggregate Success Launches for launch site CCAFS SLC-40

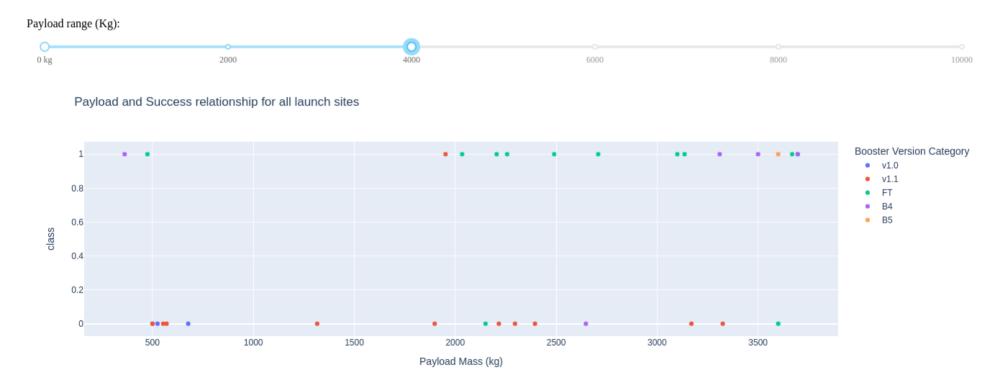


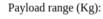


#### Payload and Launch Outcome scatter plot for all launch sites with different payload selected in the range slider.

- Booster Version FT has best overall success rating from light payload to heavy payload compared to other Booster Versions. Booster Version v1.1 has worst overall success rating from light payload to heavy payload.
- Source:
  - aeromars Capstone Data Visualization via Dash

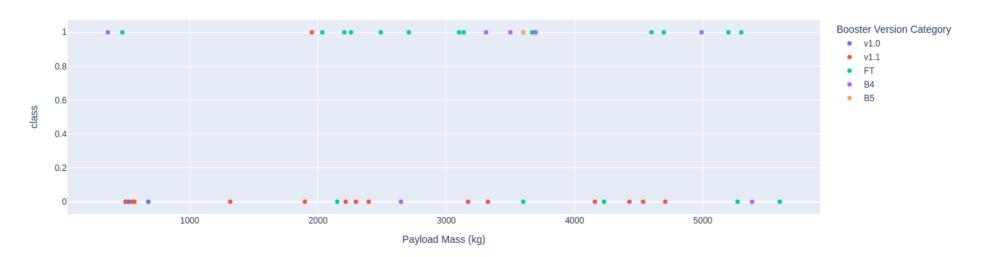








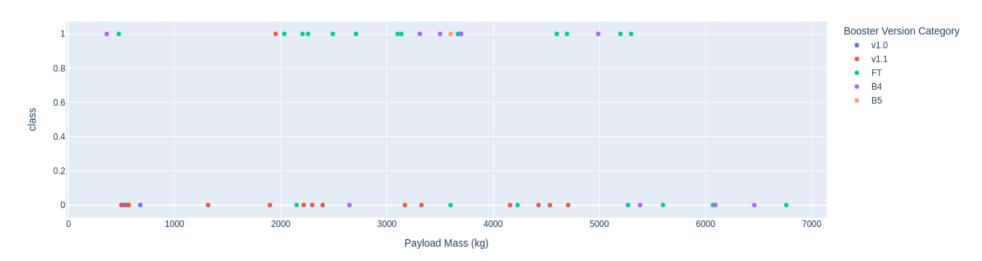
#### Payload and Success relationship for all launch sites

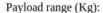






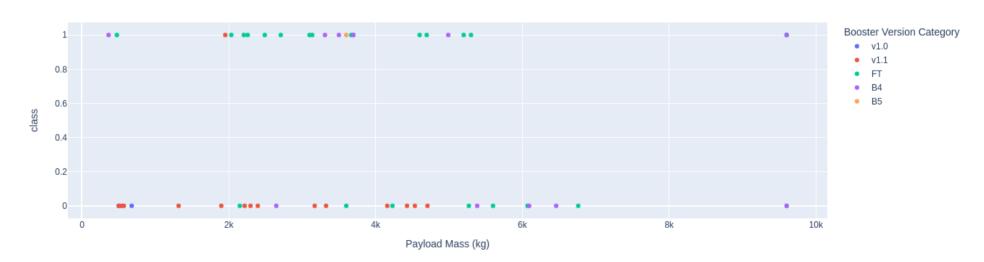
#### Payload and Success relationship for all launch sites







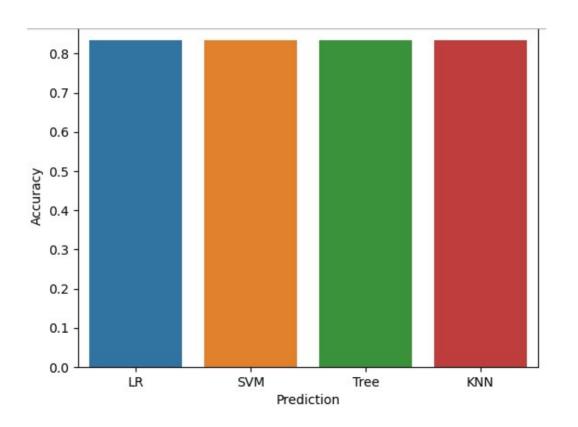
Payload and Success relationship for all launch sites



#### Classification Accuracy Results: Predictive Analysis (Classification) Predictive Analysis

- All models have accuracy of 0.83333.
- Accuracy of 0.83333 is acceptable but a better model (besides mentioned four models) highly recommended.
- Source:
  - aeromars Capstone Predictions

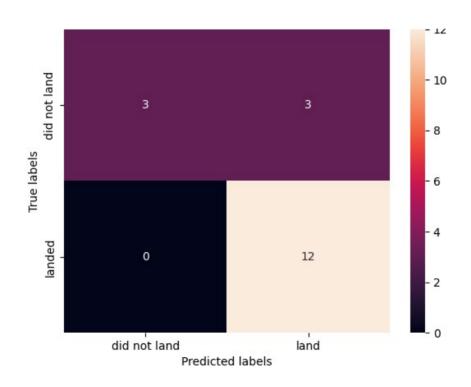
#### Classification Accuracy Results: Predictive Analysis (Classification) Predictive Analysis



## Confusion Matrix Results: Predictive Analysis (Classification) Predictive Analysis

- All models have same confusion matrix.
- Confusion matrix show prediction model "did not land" as "landed" about 20% of the time.
- Source:
  - aeromars Capstone Predictions

# Confusion Matrix Results: Predictive Analysis (Classification) Predictive Analysis



#### Feature Selection Results: Predictive Analysis (Classification) Feature Selection

- All prediction models (Linear Regression, Support Vector Machines, Decision Tree, K-Nearest Neighbor) result in same accuracy of 0.8333 and confusion matrix. 0.8333 accuracy is acceptable, but a better model highly exists using different machine learning algorithms.
- Most important features are rocket versions, flight numbers, and payload mass.
   Remaining features are important but may be excluded to test new models for better model accuracy.

#### **Conclusion**

- Obvious that success rate mostly depends on product development denoted by rocket versions but also payload mass, flight numbers. As rocket versions increase, payload mass also increased, orbit types and launch sites become irrelevant.
- Rocket versions denote technological advances in particular rocket first stage. Later rocket versions have higher success rate.
- All prediction models (Linear Regression, Support Vector Machines, Decision Tree, K-Nearest Neighbor) result in same accuracy of 0.8333.

#### **Appendix**

- Capstone Project
  - https://github.com/aeromars/data science capstone/
- Capstone Project Data Collection API
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_01
     -Data\_Collection\_API.ipynb
- Capstone Project Data Web Scraping
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_02
     -Data\_Collection\_via\_Web\_Scraping.ipynb
- Capstone Project Data Wrangling
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_03
     -Data\_Wrangling.ipynb

#### **Appendix**

- Capstone Project EDA via SQL
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_04
     EDA via SQL.ipynb
- Capstone Project EDA via Visualization
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_05
     -EDA\_via\_Visualizations.ipynb
- Capstone Project Data Visualization via Folium
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_06
     -Data\_Visualization\_via\_Folium.ipynb
- Capstone Project Data Visualization via Dash
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_07
     -Data\_Visualization\_via\_Dash.py

#### **Appendix**

- Capstone Project Predictions
  - https://github.com/aeromars/data\_science\_capstone/blob/main/Capstone\_08
    -Predictions.ipynb
- Capstone Project Presentation