

Winning Space Race with Data Science

KDennisL 2023.06.09



Outline

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

Executive Summary

- The following Methodologies were used through this study:
 - Data Collection using APIs
 - Data Collection using Web Scraping Technologies
 - Data Wrangling
 - Exploratory Data Analysis using SQL
 - Exploratory Data Analysis using Data Visualization Technologies
 - Interactive Visual Analytics with Folium
 - Interactive Visual Analytics with Plotly Sash
 - Machine Learning for Predicting Results
- Summary of all results from
 - Exploratory Data Analysis
 - Interactive Visual Analytics
 - Predicting using Machine Learning

Introduction

- Project background and context
 - Commercial Space is evolving and makes travel to space affordable for everyone
 - The costs for launches are relatively expensive for most companies with 165 million dollars
 - SpaceX on the other hand only pays 62 million dollars as they can reuse the first stage
- Problems you want to find answers
 - What factors are important for successfully landing a Falcon 9?
 - How are these connected?
 - Can we predict the outcome of a landing?



Methodology

Executive Summary

- Data collection methodology:
 - Data was collected from two urls:
 - https://api.spacexdata.com/v4/launches/past using a get request &
 - https://en.wikipedia.org/wiki/List of Falcon 9 and Falcon Heavy launches using Webscraping approaches
- Perform data wrangling
 - Data was cleaned and a class label was created:
 - Class = O for unsuccessful landing
 - Class = 1 for successful landing
- Perform exploratory data analysis (EDA) using visualization and SQL
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Methodology

Executive Summary

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Collected data was scaled and split into Training and Test Dataset
 - Four different ML models were optimized by GridSearch
 - Evaluation was performed using Confusion Matrix and test data accuracy

Data Collection

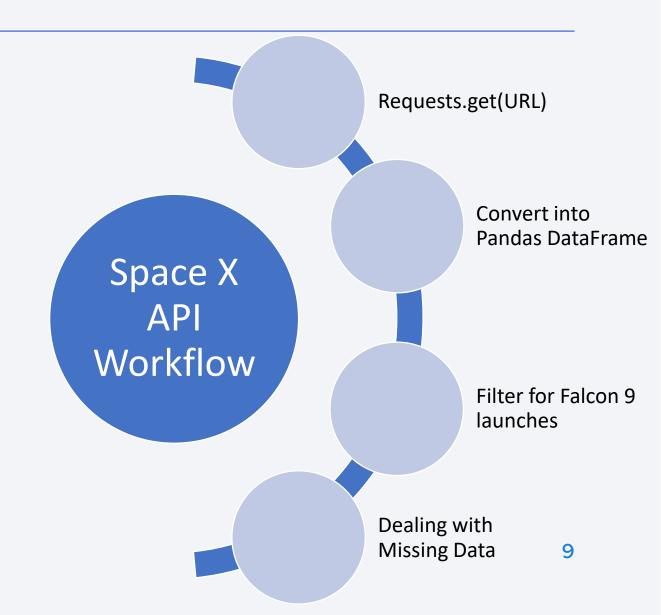
- Describe how data sets were collected.
- Two datasets were collected via:
 - API provided by SpaceX and
 - Websraping from Wikipedia

Data Collection - SpaceX API

- SpaceX offeres to access past launches via this URL:
 - https://api.spacexdata.com/v4/launche s/past

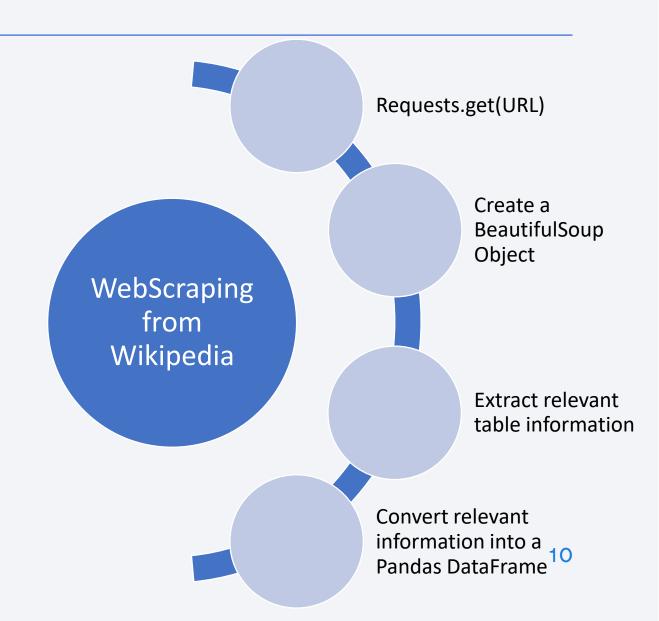
 The Jupyter Notebook can be found here:

https://github.com/KDennisL/AppliedDataScienceCapstoneProject/blob/main/J0101-jupyter-labs-spacex-data-collection-api.ipynb



Data Collection Scraping

- Wikipedia contains public available information about Falcon 9 launches from SpaceX
- This data was also parsed using Webscraping processes (see flow chart)
- The Jupyter Notebook can be found here:
 - WebScraping



Data Wrangling

- Previous data was used to:
 - Calculate the number of launches on each site
 - Calculate the number and occurrence of each orbit
 - Calculate the number and occurrence of mission outcome of the orbits
 - Create a landing outcome label from Outcome column
- The Jupyter notebook can be found here:
 - Data Wrangling

EDA with Data Visualization

- The following charts were plotted:
 - relationship between Flight Number and Launch Site as Catplot
 - relationship between Payload and Launch Site as Scatterplot
 - relationship between success rate of each orbit type as Bar Chart
 - relationship between FlightNumber and Orbit type as Scatterplot using Classes as hue
 - relationship between Payload and Orbit type using Scatterplot
 - launch success yearly trend using Lineplot
- The Jupyter Notebook can be found here:
 - EDA Data Visualization

EDA with SQL

Using bullet point format, summarize the SQL queries you performed

- %sql SELECT DISTINCT("Launch_Site") FROM SPACEXTBL;
- %sql SELECT * FROM SPACEXTBL WHERE Launch_Site LIKE '%CCA%' LIMIT 5
- %sql PRAGMA table_info(SPACEXTBL);
- %sql SELECT SUM(PAYLOAD_MASS__KG_)/Count(PAYLOAD_MASS__KG_) FROM SPACEXTBL WHERE Booster_Version = 'F9 v1.1'
- %sql SELECT MIN(CURRENT_DATE Date), Date FROM SPACEXTBL WHERE Landing_Outcome = 'Success (ground pad)'
- %sql SELECT Booster_Version, Date,PAYLOAD_MASS__KG_ FROM SPACEXTBL WHERE Landing_Outcome = 'Success (drone ship)' AND PAYLOAD_MASS__KG_ > 4000 AND PAYLOAD_MASS__KG_ < 6000
- %sql SELECT Mission_Outcome, COUNT(*) FROM SPACEXTBL WHERE Mission_Outcome <> 'None' GROUP BY Mission_Outcome
- %sql SELECT DISTINCT(Booster_Version) FROM SPACEXTBL where PAYLOAD_MASS__KG_ = (select MAX(PAYLOAD_MASS__KG_) from SPACEXTBL)
- The Jupyter notebook can be found here:
 - EDA with SQL

Build an Interactive Map with Folium

- Markers, circles, lines and marker clusters added to the sitemap
- Launch sites were marked
- A highlighted circle area with a text label on a specific coordinates
- Marker clusters were used for labeling failed and successful launches
- Lines represent the distance between a launch site to the selected coastline point
- The Jupyter Notebook can be found here:
 - Folium

Build a Dashboard with Plotly Dash

- The Dashboard was created using Plotly Dash
- Interactive Pie Chars and Scatter Plots are shown for
 - Successful launches and
 - The Correlation between Payload and Success
- Interactive graphs easier data analysis than text or tables
- The Jupyter Notebook can be found here:
 - Dash

Predictive Analysis (Classification)

- Data from previous steps was loaded in
- Data was split into Train and Test Datasets
- Four different models were evaluated using GridSearchVG
- All models behaved very similar by accuracy and Confusion plot
- The Jupyter Notebook can be found here:
 - PredictiveAnalysis

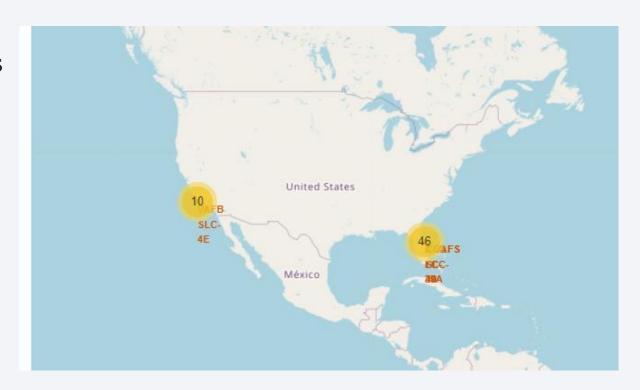
Results

- Exploratory data analysis results
 - Distinct Launch Sites: 4
 - The Total Payload was 45596.0 and its mean 2928.4
 - First landing was 2015
 - Four F9 were successful in drone ship and have a mass between 4 and 6k
 - The success of landing increased by year (except latest year)
- Interactive analytics demo in screenshots
- Predictive analysis results

Results

- Exploratory data analysis results
- Interactive analytics demo in screenshots
 - Most launches are close to a sea

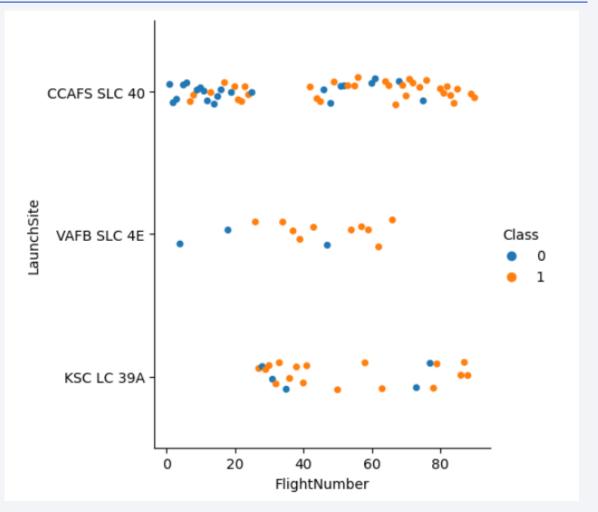
- Predictive analysis results
 - The accuracy of predicting the result of a landing is ~ 0.8





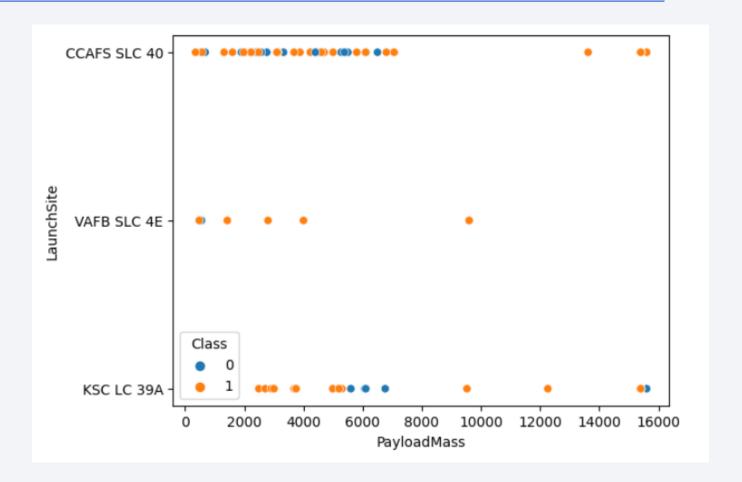
Flight Number vs. Launch Site

- Lowest Flight number is to be found at CCAFS SLC 40
- More Success for Higher Flight numbers



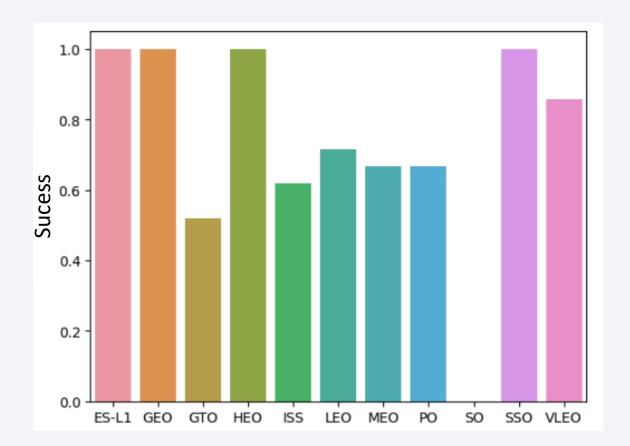
Payload vs. Launch Site

 Lower Payloads seem to have a negative impact on Success



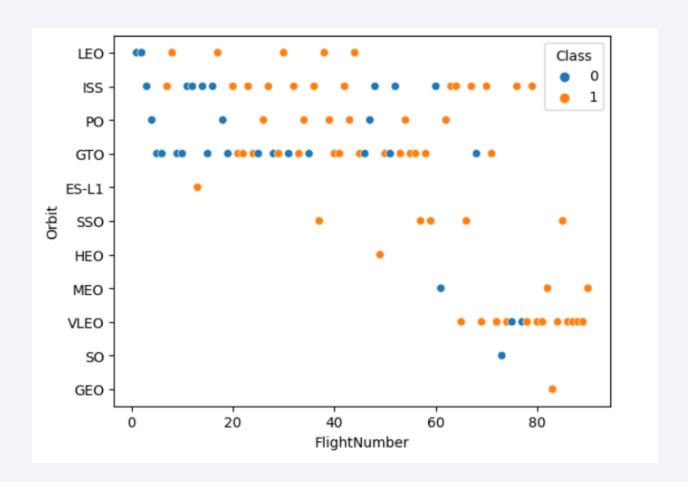
Success Rate vs. Orbit Type

- Highest Success for:
 - EL-L1
 - GEO
 - HEO and SSO



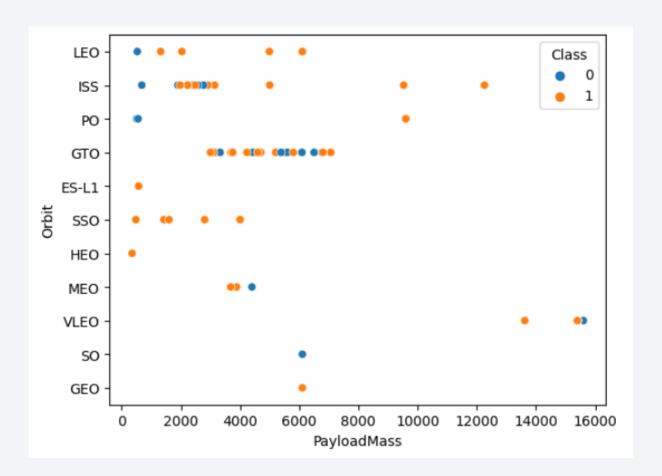
Flight Number vs. Orbit Type

 Highest number of sucess was achieved from VLEO



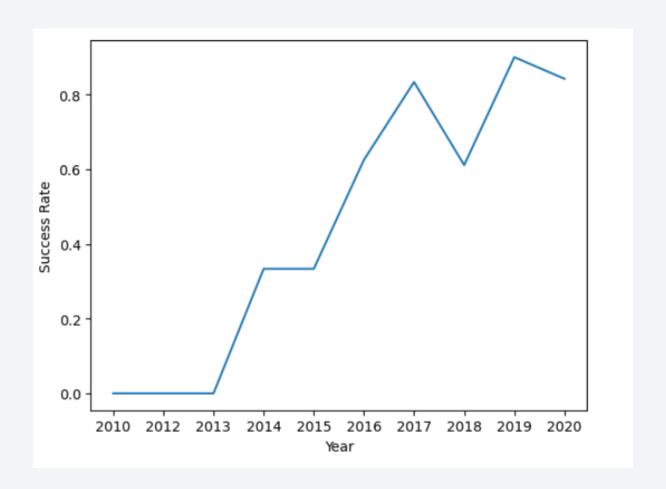
Payload vs. Orbit Type

 Most success in Landing with higher payload for LEO, ISS and Po



Launch Success Yearly Trend

- The success rate increased over the last years
- Potentially, there is a corona dip at 2020



All Launch Site Names

 Using %sql, we could obtain the unique items via DISTINCT from the table SPACEXTBL



Launch Site Names Begin with 'CCA'

In [8]:

- LIKE
 operator
 was used to
 find items
 that include
 CCA
- LIMIT was used to only show 5 items

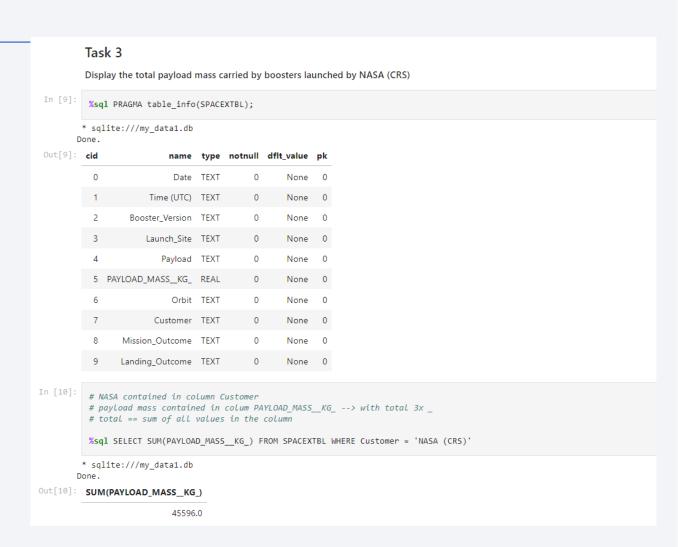
Task 2 Display 5 records where launch sites begin with the string 'CCA'

%sql SELECT * FROM SPACEXTBL WHERE Launch_Site LIKE '%CCA%' LIMIT 5

* sqlite:///my_data1.db Out[8]: Date Booster_Version Launch_Site Payload PAYLOAD MASS_KG_ Orbit Customer Mission_Outcome Landing_Outc Dragon Spacecraft CCAFS LC-F9 v1.0 B0003 06/04/2010 18:45:00 0.0 LEO SpaceX Failure (paracl Qualification Unit Dragon demo flight C1, two NASA CCAFS LC-12/08/2010 15:43:00 F9 v1.0 B0004 CubeSats. (COTS) Failure (paracl Success barrel of NRO Brouere cheese Dragon CCAFS LC-LEO NASA 525.0 7:44:00 F9 v1.0 B0005 22/05/2012 demo flight Success No att€ (ISS) (COTS) LEO CCAFS LC-SpaceX NASA 10/08/2012 500.0 0:35:00 F9 v1.0 B0006 Success No atte CRS-1 (ISS) (CRS) LEO CCAFS LC-SpaceX NASA 03/01/2013 15:10:00 F9 v1.0 B0007 677.0 Success No atte CRS-2 (ISS) (CRS)

Total Payload Mass

- Total Payload was 455969
- Obtained via customer and the sum function

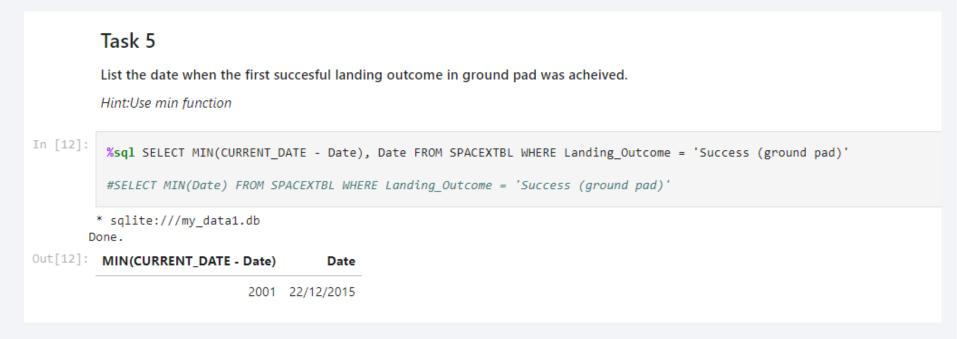


Average Payload Mass by F9 v1.1

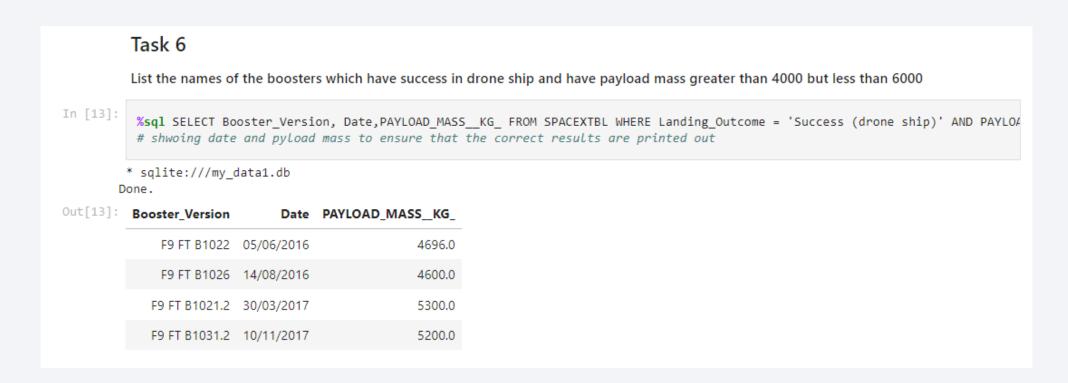
 Average payload was 2928 for F9 v.1.1.

First Successful Ground Landing Date

- The minimum of current date to the date in the table was obtained
- This was 22/12/2015



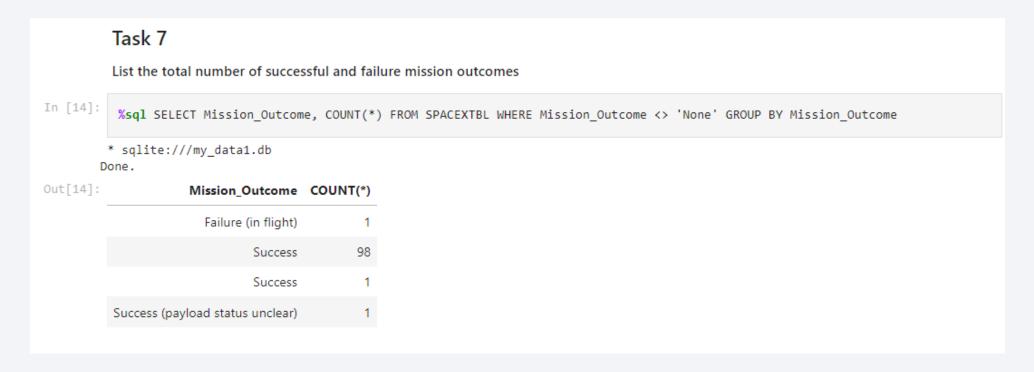
Successful Drone Ship Landing with Payload between 4000 and 6000



WHERE and Payload mass range was used to get the results

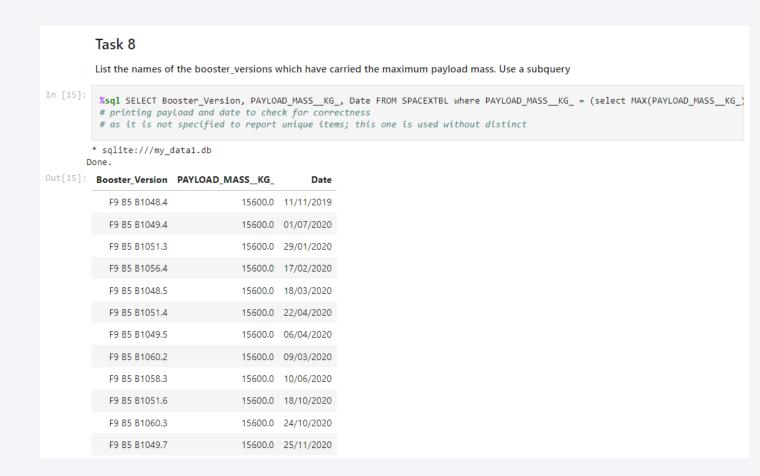
Total Number of Successful and Failure Mission Outcomes

- WHERE clause was used with unequal to non and
- Grouped by Mission outcome

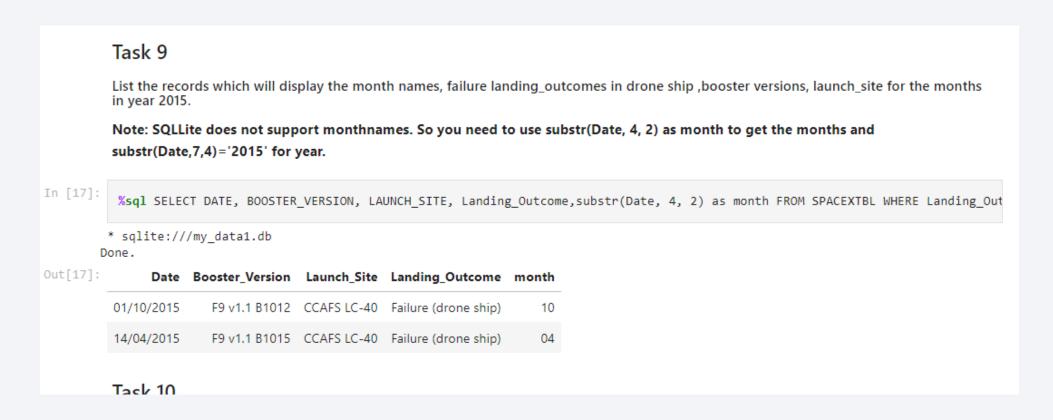


Boosters Carried Maximum Payload

- Subquerey was used to get the maximum of Payload mass
- Maximum value was used as a filter to get the highest values



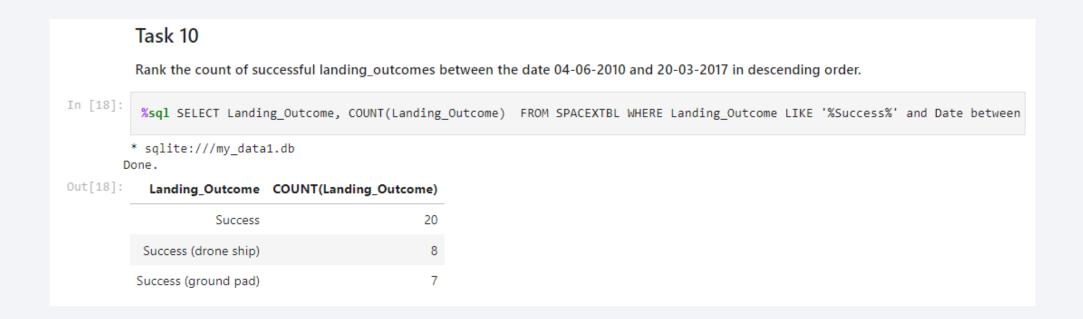
2015 Launch Records



WHERE Landing_Outcome = 'Failure (drone ship)' AND substr(Date,7,4)='2015' was used to filter the data

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

- Using LIKE operator to get a list of successfully landed items
- Date Between to only contain specific values





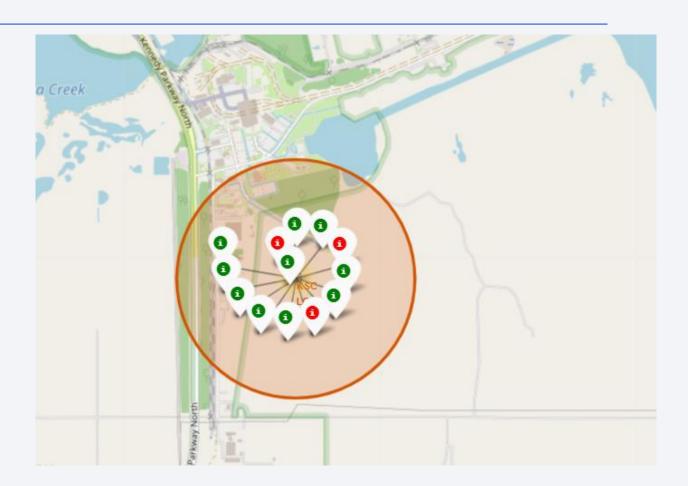
Launch Sites

- Launch Sites were only found in the US
- Either on the east of west cost



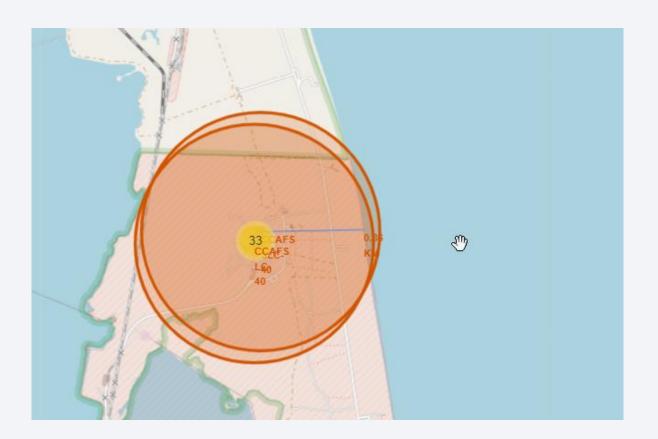
Landing Success

 RSC LC-39A has mostly successful landing (green marks)



Distance to Sea

• To eliminate any injury to humans, it is best to have a close location to the sea





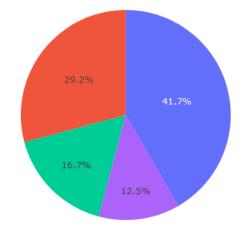
Successful launches at each site

KSC LC-39A had the most successful launches

SpaceX Launch Records Dashboard

Success Count for all launch sites

All Sites







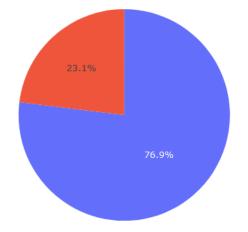
Successful launches for KSC LC-39A

• In 77%, the launches were successful

SpaceX Launch Records Dashboard

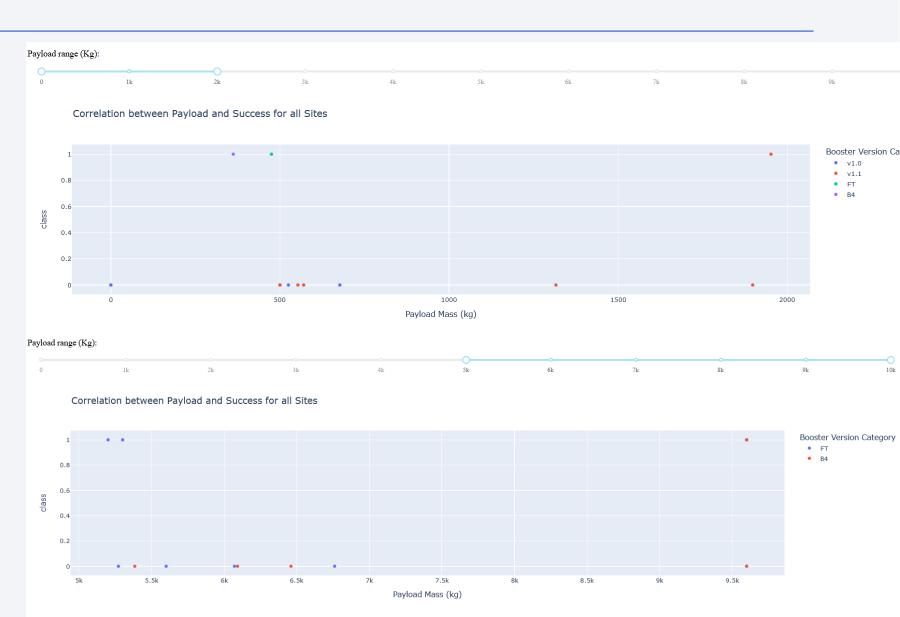
Total Success Launches for site KSC LC-39A

KSC LC-39A



Payloads as a Filter for Success

 The Success is low for extrem high and low payloards





Classification Accuracy

- I am not going to create another plot here....
- Best model was SM with a score of 0.85

TASK 6

Create a support vector machine object then create a GridSearchCV object svm_cv with cv - 10. Fit the object to find the best parameters from the dictionary parameters.

```
parameters = {'kernel':('linear', 'rbf', 'poly', 'rbf', 'sigmoid'),
                        'C': np.logspace(-3, 3, 5),
                        'gamma':np.logspace(-3, 3, 5)}
          svm = SVC()
          svm cv = GridSearchCV(svm,parameters,cv=10)
          # as -10 failed, I continued with cd =10
          svm_cv.fit(X_train, Y_train)
Out[76]: GridSearchCV(cv=10, estimator=SVC(),
                      param grid={'C': array([1.00000000e-03, 3.16227766e-02, 1.00000000e+00, 3.16227766e+01,
                                    'gamma': array([1.00000000e-03, 3.16227766e-02, 1.00000000e+00, 3.16227766e+01,
                1.00000000e+03]),
                                   'kernel': ('linear', 'rbf', 'poly', 'rbf', 'sigmoid')})
        In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
        On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
          print("tuned hpyerparameters :(best parameters) ",svm cv.best params )
          print("accuracy :",svm_cv.best_score_)
       tuned hpyerparameters :(best parameters) {'C': 1.0, 'gamma': 0.03162277660168379, 'kernel': 'sigmoid'}
       accuracy : 0.8482142857142856
```

Confusion Matrix

- There were not type II errors
- The number of type I error was similar to the number of correctly predicted not landed

TASK 7 Calculate the accuracy on the test data using the method score: n [78]: print("Accurace of test data set is :",svm_cv.score(X_test, Y_test)) Accurace of test data set is: 0.83333333333333334 We can plot the confusion matrix n [79]: yhat=svm_cv.predict(X_test) plot_confusion_matrix(Y_test,yhat) **Confusion Matrix** did not land - 10 3 **True labels** 12 did not land land Predicted labels

Conclusions

- Launch success increased from year to year starting with year 2015
- Starting points are close to the sea
- KSC LC-39A had the most successful launches
- Extreme values for Payload had a negative impact
- SVM performed best to predict the outcome

Appendix

• Include any relevant assets like Python code snippets, SQL queries, charts, Notebook outputs, or data sets that you may have created during this project

