

SpaceX Launch Analysis

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https://github.com/kevinbenson8/IBM-data-science-classes/tree/main/SpaceX_Capstone

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EXECUTIVE SUMMARY



Data was obtained, cleaned, explored, then analyzed from SpaceX. This data was then used to determine if a launch will be successful or not.

Visualization with plotly and folium aided in getting a grasp of the relation between variables in this dataset, in addition to SQL commands.

Several machine learning models provided through Scikitlearn were tested to predict the outcome of a launch. Logistic regression, SVM, Decision Tree and KNN were all the best at predicting the launch outcome.

INTRODUCTION



It was set out to determine if the first stage of a SpaceX rocket will successfully land, and therefore be reused. This drastically lowers the cost of launching rockets.

SpaceX provides launch data on their website for free. This data was refined and analyzed using several methods detailed in this report.

If we can successfully predict a safe launch, then this is advantageous and will save the company lots of money.

METHODOLOGY: Data Collection/Wrangling



Data was collected from SpaceX using an API (https://api.spacexdata.com) in addition to web scraping with BeautifulSoup. Then, data was wrangled into a pandas dataframe with a 'Class' column indicating a launches result.

Core data was collected with functions doing API calls with the python requests module. Some of the core data column titles were: *BoosterVersion, PayloadMass, Orbit, LaunchSite, Outcome, Longitude, and Latitude.*

The webscraping portion of the data collection phase was done using an HTML table from wikipedia. Once the data had been collected, it was wrangled. Launch results were interpreted into '0' or '1' in the 'Class' column. This was then used as the independent variable in predictive analysis.

METHODOLOGY: EDA/Visual Analytics



Exploratory Data Analysis (EDA) was completed using SQL and visual analysis was done with folium, matplotlib, and seaborn. Using these graphical methods, the relationships between dependent variables and 'Class' was better understood.

For the SQL portion of EDA a table was uploaded to IBM's cloud server, then queries were performed with help from the sqalchemy toolkit for python. Queries were performed to extract information such as unique launch sites, average payload masses, different landing outcomes.

An interactive dashboard was made with plotly and dash to even further explore the data. The dashboard contains a dropdown menu to switch between landing sites and a slider to select a desired range of payload masses. Callback functions were used to build these interactive components.

METHODOLOGY: Predictive Analysis



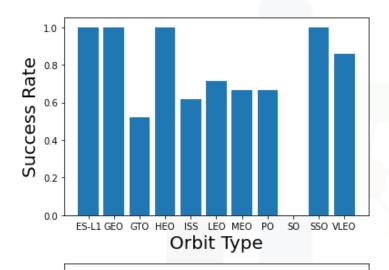
Lastly, several machine learning models were fit to training data extracted from the dataset and scored on their abilities to successfully predict the outcome of a launch from test data.

Using the sklearn python library, the data was pre-processed by standardizing with *StandardScaler*, then split 80/20 into training and testing sets with *train_test_split*.

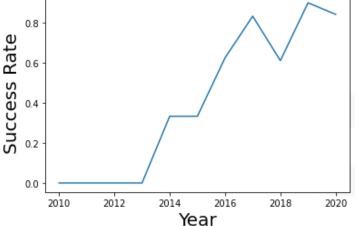
A *GridSearchCV* object was created for sever machine learning models, then fit to the training data. The models in question were: logistic regression, SVM, decision tree, and KNN.

The performance of these models was then analyzed via the R-squared score and by plotting a confusion matrix.

RESULTS: EDA with Visual Analysis



- Some orbit types are indicative of a higher success rate
- ES-L1, GEO, HEO, and SSO all has a success rate of 100%
- GTO was the least successful orbit type at 50%



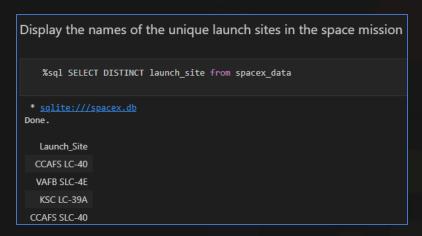
- There was not a success for the years 2010-2013
- From 2013-2017 the success rate increased from 0% to 80%.
- The success rate decreased from 2017 to 2018, and again from 2019 to 2020

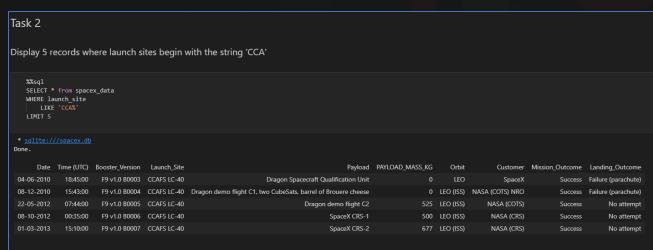






Results: EDA with SQL (Tasks 1-4)





```
Display the total payload mass carried by boosters launched by NASA (CRS)

***sq1
SELECT customer, sum(payload_mass_kg) as 'Total Payload Mass' from spacex_data
WHERE customer
LIKE 'NASA (CRS)'

* sqlite://spacex.db
Done.

Customer Total Payload Mass
NASA (CRS) 45596
```

```
Display average payload mass carried by booster version F9 v1.1

%%sql
SELECT booster_version, AVG(payload_mass_kg) as 'Average Payload Mass' from spacex_data
WHERE booster_version
LIKE 'F9 v1.1%'

* sqlite:///spacex.db
Done.

Booster_Version Average Payload Mass
F9 v1.1 B1003 2534.666666666665
```

Results: EDA with SQL (Tasks 6-8)

```
List the date when the first successful landing outcome in ground pad was acheived.

Hint:Use min function

**sql
SELECT MIN(date) as 'First Success' from spacex_data
WHERE landing_outcome
LIKE 'Success (ground pad)'

* sqlite:///spacex.db
Done.

First Success
01-05-2017
```

```
List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
    %%sal
    SELECT booster_version, payload from spacex_data
    WHERE landing outcome
       LIKE 'Success (drone ship)'
        AND payload mass kg > 4000
       AND payload mass kg < 6000
  Booster Version
                          Payload
    F9 FT B1022
                          JCSAT-14
    F9 FT B1026
                          JCSAT-16
   F9 FT B1021.2
                           SES-10
   F9 FT B1031.2 SES-11 / EchoStar 105
```

IBM **Developer**

```
%%sql
SELECT count(*) as 'Number of Successful' from spacex_data
WHERE mission_outcome LIKE 'Success%'

* sqlite:///spacex.db
Done.
Number of Successful
100
```

```
List the names of the booster_versions which have carried the maximum payload mass. Use a subquery
    %%sql
    SELECT booster version from spacex data
    WHERE payload_mass_kg =
         (SELECT max(payload mass kg) from spacex data)
 Done.
  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
```

Results: EDA with SQL (Tasks 9-10)

```
List the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015

%%sql
SELECT landing_outcome, booster_version, launch_site from spacex_data
WHERE landing_outcome
LIKE 'Failure (drone ship)'
AND date
LIKE '%2015%'

* sqlite:///spacex.db
Done.

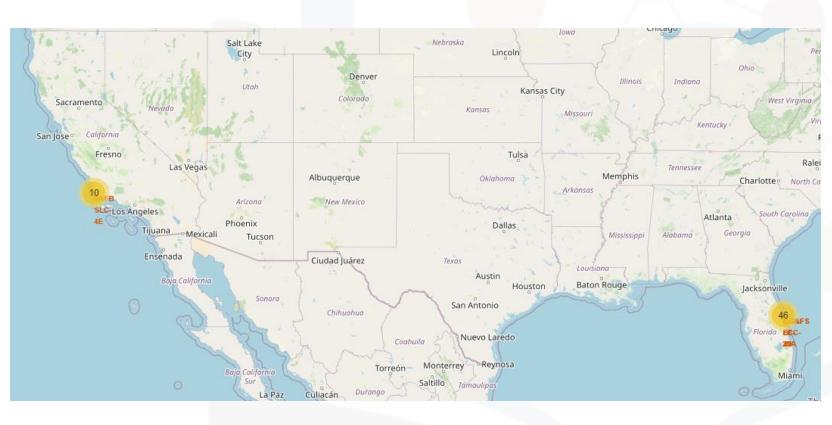
Landing_Outcome Booster_Version Launch_Site
Failure (drone ship) F9 v1.1 B1012 CCAFS LC-40

Failure (drone ship) F9 v1.1 B1015 CCAFS LC-40
```

```
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
    %%sql
    SELECT landing_outcome, count(landing_outcome) from spacex_data
    WHERE date
       BETWEEN '06-04-2010' AND '20-03-2017'
   GROUP BY landing_outcome
   ORDER BY count(landing_outcome) desc
Done.
    Landing_Outcome count(landing_outcome)
            Success
  Success (drone ship)
  Success (ground pad)
         No attempt
   Failure (drone ship)
   Controlled (ocean)
             Failure
         No attempt
    Failure (parachute)
```



Interactive Folium Map



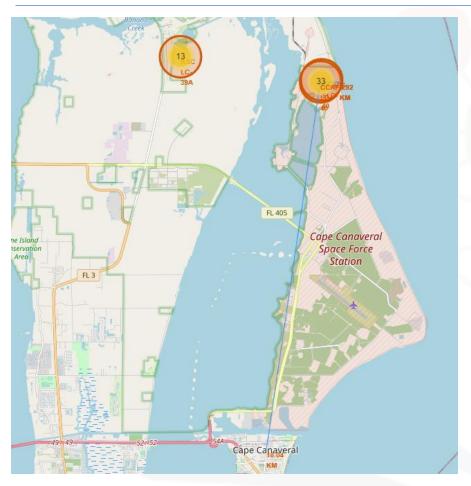
- VAFB SLC-4E is the only launch site on the west coast. All three other launch sites are in Florida
- 10 out of 56 launches have occurred at VABFC SLC-4E

Interactive Folium Map



- Three of the launch sites are on the coast of Florida, near Cape
 Canaveral
- Two of them are right on top of each other
- Three out of thirteen launches at KSC LC-39A were unsuccessful

Interactive Folium Map

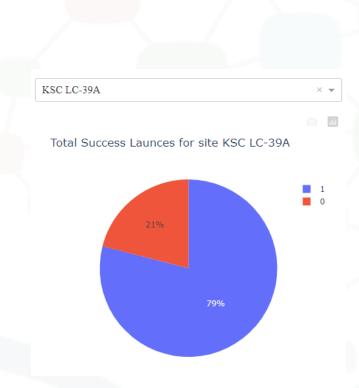


- The distance from CCAFS LC-40 to Cape Canaveral is 18.04 km
- The distance from CCAFS-LC-40 to the Atlantic Ocean is 0.93 km

Interactive Dashboard

SpaceX Launch Records Dashboard





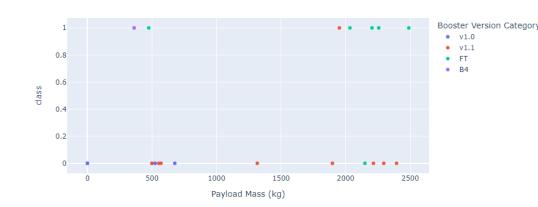
- Of the 4 launch sites, KSC LC-39A
 was used the most, 41.7% of the
 time
- CCAFS SLC-40 was used the least,
 12.5% of the time
- KSC LC-39A had a 79% success rate
 on launches, which was the highest
 among the launch sites

Interactive Dashboard



- For all payload weights, booster versions 1.0 and 1.1
 were not very successful
- The largest payloads have only been carried by boosters v1.0 and FT

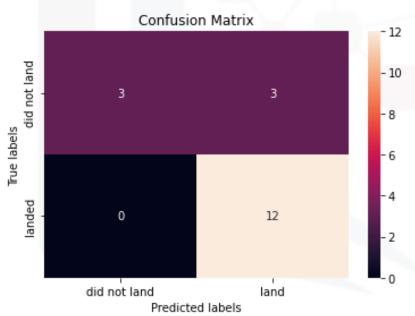
Correlation Between Payload Success for All Sites



- For payloads below 2500 kg, v1.0 had no successes,
 while BT v1.1 only had a single success each
- Booster version FT has only one failure with a payload below 2500 kg

RESULTS: Predictive Analysis

ML Algorithm	R-squared score
Logistic Regression	0.8333
SVC	0.8333
Decision Tree	0.8333
KNN	0.8333



- LR, SVC, and Decision Tree and KNN all performed identically
- This identical performance is likely due to such a small testing set
- The confusion matrix looked identical for all of the algorithms
- The algorithms each predicted 3 false
 negatives, while they never correctly predicted
 a false positive

CONCLUSION



To generalize what was shown in this report, visual analysis showed us that success rate generally increased over time.

We also saw how the interactive dashboard explained that launch site KSC LC-39A was the most used, and most successful

Folium was used to explore the geospatial data relating to the launch sites, and SQL was used for preliminary EDA.

Lastly, we fit the training data to four different machine learning algorithms and saw identical performance from all of them.

APPENDIX

