

Winning Space Race with Data Science

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Outline



Executive Summary



Introduction



Methodology



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1. Executive Summary



- SpaceX advertises Falcon 9 rocket with a cost of 62 million dollars; other providers cost upwards of 165 million.
- Much of the savings is because SpaceX can reuse the first stage.
- SpaceY wants to determine the price of each launch, training a machine learning model and using public information to **predict if SpaceX will reuse the first stage.**
- Public information indicates a first stage Falcon 9 Booster to cost upwards of 15 million.
- The machine learning model can predict, with an **83.3% level of accuracy**, the landing success of the first stage Falcon 9 Booster.

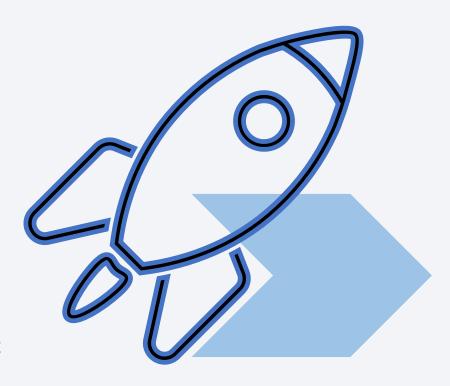
2. Introduction

- PROJECT SCENARIO: Capstone Course

In the Applied Data Science Capstone course, we will **predict if the Falcon 9 first stage will land successfully** as a real-world business problem to solve within our Data Scientist role.

- BUSINESS PROBLEM: Cost of a launch

Determine if the first stage will land in order to **determine the cost of a launch.** This information can be used to bid against SpaceX for a rocket launch.





3. Methodology: *Executive Summary*



- a) Data collection methodology: API & Web Scraping
- b) Perform data wrangling
- c) Perform exploratory data analysis (EDA) using visualization and SQL
- d) Perform interactive visual analytics using Folium and Plotly Dash
- e) Perform predictive analysis using classification models

Data Collection – SpaceX API

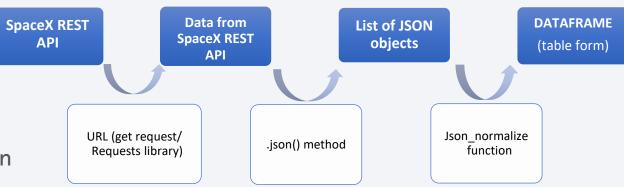


• Objective:

 Obtain a flat table with the structured json data.

• Additional process:

- Filtered the Dataframe to obtain Falcon 9 launches only.
- Replace missing values with the mean (payload mass).



GitHub URL:

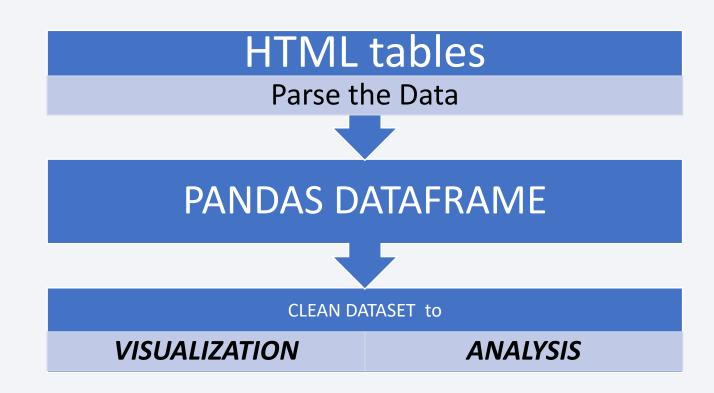
https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab%201 %20Collecting%20%20the%20data.ipynb





• Process:

- Use Python BeautifulSoup package to web scrape some HTML tables.
 - Extract columns and variables names from an ID Number.
 - Filter/sample the data to remove Falcon 1 launches.
 - Deal with NULLs.



Data Wrangling



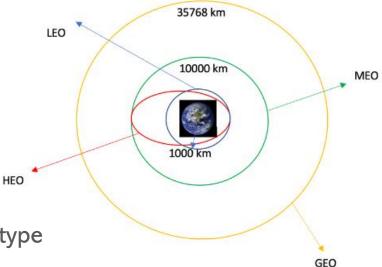
Objectives:

- Exploratory Data Analysis
- Determine Training Supervised Models Labels

• Process:

- Obtain the number of launches on each site
- Obtain the number and occurrence of each orbit
- Obtain the number and occurrence of mission outcome per orbit type
- Create a landing outcome label from "outcome" column

landing_class = 0 → bad_outcome (first stage did not land successfully)
landing_class = 1 → otherwise (first stage landed successfully)



GitHub URL:

https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab%202 %20Data%20wrangling.ipynb

EDA with Data Visualization





Objectives:

Exploratory Data Analysis

Preparing Data Feature Engineering



Process:

SpaceX dataset → Pandas Dataframe

Using visualization libraries

➤ Matplotlib

➤ Seaborn



Plot out:

FlightNumber **vs** PayloadMass

Flight Number **vs** Launch Site

Payload **vs** Launch Site

Success rate **vs** Orbit type

FlightNumber vs Orbit type

Payload vs Orbit type

Launch success yearly trend

EDA with SQL



Objectives:

Execute SQL queries to answer assignment questions:

1. Names of the unique launch sites

SELECT DISTINCT LAUNCH_SITE FROM SPACEXTBL;

2. 5 records where launch sites begin with 'CCA'

SELECT LAUNCH_SITE FROM SPACEXTBL WHERE LAUNCH_SITE LIKE 'CCA%' LIMIT 5;

3. Total payload mass carried by boosters launched by NASA (CRS)

SELECT SUM(PAYLOAD_MASS__KG_) AS payload_mass_kg FROM SPACEXTBL WHERE Customer = 'NASA (CRS)';

4. Average payload mass carried by booster version F9 v1.1

SELECT AVG(PAYLOAD_MASS__KG_)
FROM SPACEXTBL
WHERE Booster_Version LIKE 'F9 v1.0%';

5. When the first successful landing outcome in ground pad was achieved.

SELECT MIN(DATE) AS Date
FROM SPACEXTBL
WHERE LANDING__OUTCOME = 'Success (ground pad)';

EDA with SQL



6. Names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000.

```
SELECT BOOSTER_VERSION
FROM SPACEXTBL
WHERE LANDING__OUTCOME = 'Success (drone ship)'
AND 4000 < PAYLOAD_MASS__KG_ < 6000;
```

- 7. Total number of successful and failure mission outcomes.
- SELECT MISSION_OUTCOME, COUNT(MISSION_OUTCOME) AS TOTAL FROM SPACEXTBL GROUP BY MISSION OUTCOME;

- 8. Names of the booster_versions which have carried the maximum payload mass.
- SELECT DISTINCT BOOSTER_VERSION
 FROM SPACEXTBL
 WHERE PAYLOAD_MASS__KG_ = (
 SELECT MAX(PAYLOAD_MASS__KG_)
 FROM SPACEXTBL);

9. Failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015.

SELECT LANDING__OUTCOME, BOOSTER_VERSION, LAUNCH_SITE FROM SPACEXTBL WHERE Landing__Outcome = 'Failure (drone ship)' AND YEAR(DATE) = 2015;

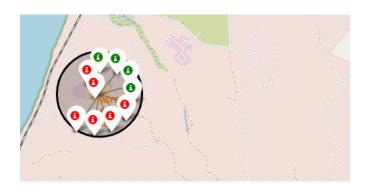
10. Rank the count of landing outcomes between the date 2010-06-04 and 2017-03-20, in descending order.

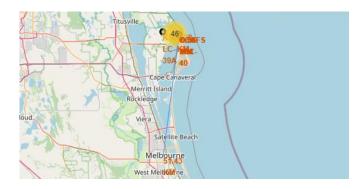
```
SELECT LANDING__OUTCOME, COUNT(LANDING__OUTCOME) AS TOTAL FROM SPACEXTBL
WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20'
GROUP BY LANDING__OUTCOME
ORDER BY TOTAL DESC
```











Build an Interactive Map with Folium

Objectives:

TASK 1: Mark all launch sites on a map.

- Using latitude and longitude coordinates (spacex_launch_geo.csv).
- Create a folium Map Object (NASA Johnson Space Center as a start location).
- Add a highlights circle areas and another markers.

TASK 2: Mark the success/failed launches for each site on the map.

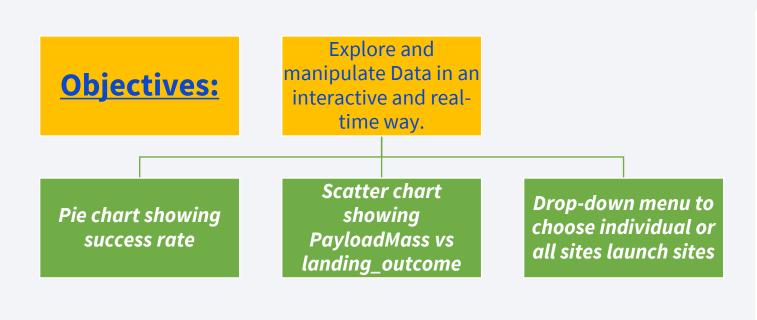
TASK 3: Calculate the distances between a launch site to its proximities.

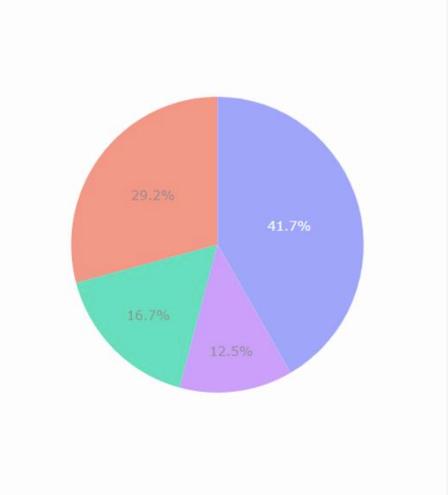
- Railways
- Highways
- Coastlines
- Cities

GitHub URL:

Build a Dashboard with Plotly Dash







Predictive Analysis

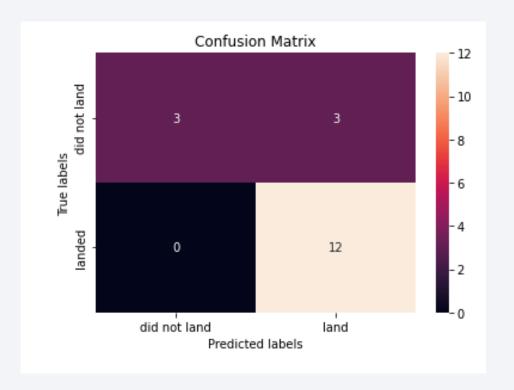


• Objectives:

• Preform exploratory Data Analysis and determine Training Labels.

Process:

- Import libraries
- Loaded the Dataframe created during data collection
- Create a column for the class
- Standardize the data
- Split into training data and test data
- Fit the training data to different model types:
 - LR
 - SVM
 - Decision Tree Classifier
 - K-nearest Neighbors Classifier
- Select the best one hyperparameters to each model,
- Evaluate the accuracy of each one.

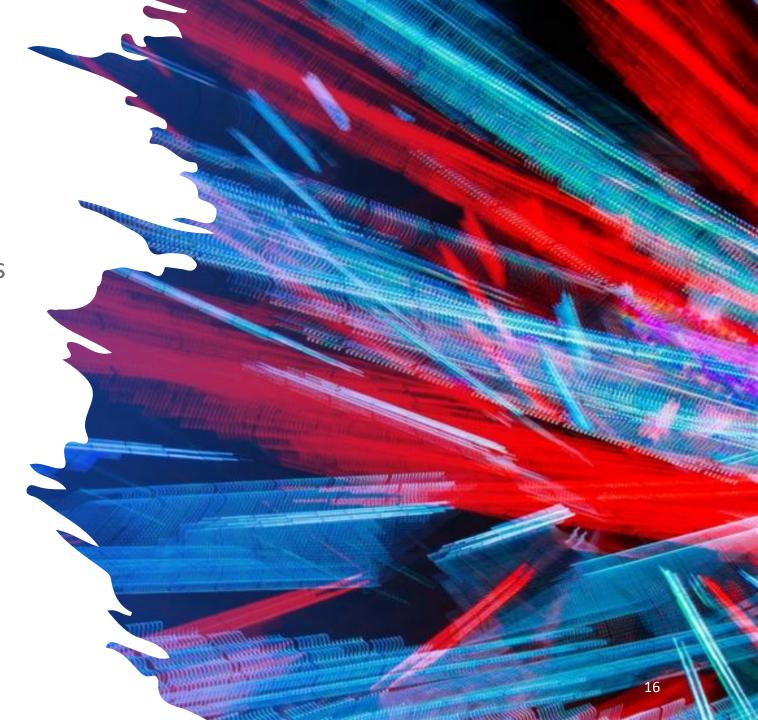


GitHub URL:

https://github.com/apnoguero/IBM-Data-Science-Capstone_SpaceX/blob/main/Lab7_%20Machine%20Learning%20Prediction.ipynb

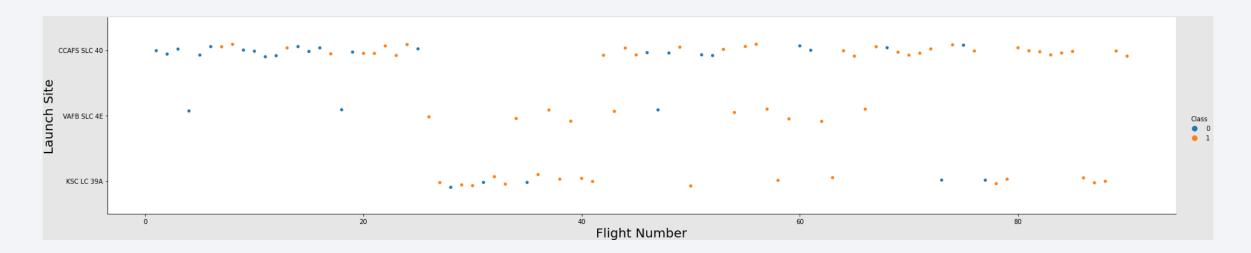
Results

- Exploratory data analysis results
- Interactive analytics demo in screenshots
- Predictive analysis results



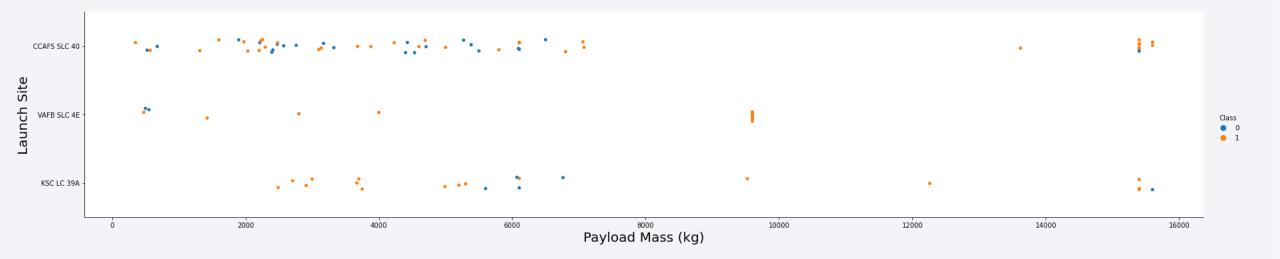


Flight Number vs. Launch Site



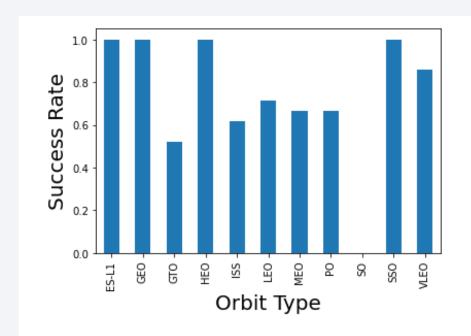
- Launch Site CCAFS SLC 40 & Flight Number < 25, appears to have most landing failures (colour blue).
- Launch Site CCAFS SLC 40 report more launches than the others sites.
- As the flight number increases, the first stage is more likely to land successfully (colour orange).

Payload vs. Launch Site



- Launch Site *CCAFS SLC 40* appears to concentrate launches when the *Payload Mass is less than 7500 kg.*
- The rockets with heavy payload mass (greater than 10000 kg) are launched from CCAFS SLC 40 and KSC LC 39A sites.

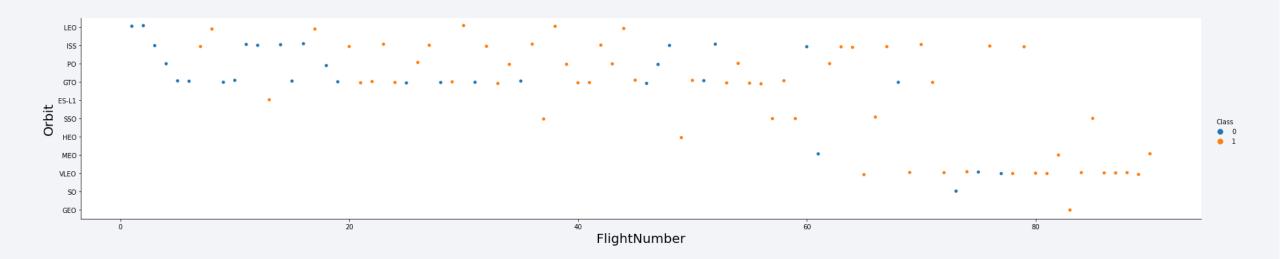
Success Rate vs. Orbit Type



Analyze the ploted bar chart try to find which orbits have high success rate. High success rate: ES-L1, GEO, HEO and SSO. Lowest success rate: SO and GTO.

- All orbit types except SO have reported successfully 1st stage landings.
- The highest success rates were reported for the ES-L1, GEO, HEO, and SSO orbits.

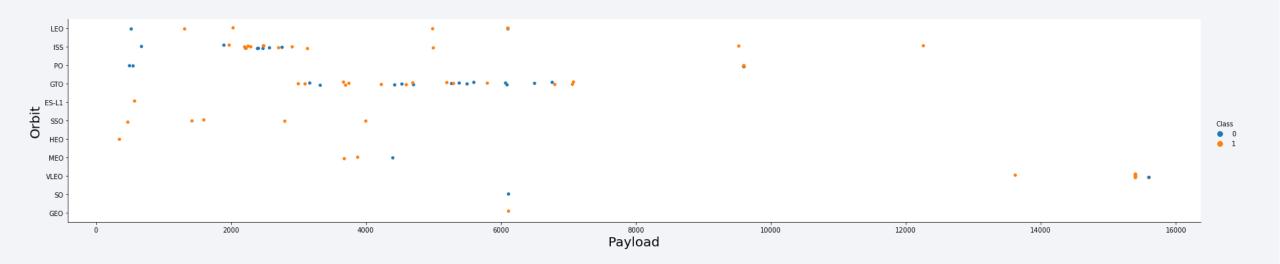
Flight Number vs. Orbit Type



Observations:

Visually, no significant relationship is reported between the Flight
 Number vs Orbit type, although in the LEO orbit, the success
 appears related to the number of flights (higher number reports
 higher success).

Payload vs. Orbit Type



- Heavy payloads (higher than 9000) report positive landing rates for LEO, ISS and PO orbit.
- For GTO, we cannot distinguish a conclusive result of succession rates.
- We can observe a correlation between ISS and Payload when Payload values are around 2000.

Launch Success Yearly Trend

```
In [10]: # Plot a line chart with x axis to be the extracted year and y axis to be the su
          ccess rate
          df2 = df.copy()
          df2['Year'] = year
          df2.groupby('Year')['Class'].mean().plot()
Out[10]: <AxesSubplot:xlabel='Year'>
           0.8
           0.6
           0.4
           0.2
           0.0
                       2013
                                 2015
                                          2017
                                                   2019
              2010
                                    Year
```

Observations:

 Visually, we can observe that launch success rate is increasing significantly since 2013, although in 2018 it decrease a bit.

All Launch Site Names

Display the names of the unique launch sites in the space mission

```
%%sql
SELECT DISTINCT LAUNCH_SITE
FROM SPACEXTBL;
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1 od8lcg.databases.appdomain.cloud:31498/bludb Done.

launch_site

CCAFS LC-40

CCAFS SLC-40

KSC LC-39A

VAFB SLC-4E

Launch Site Names Begin with 'CCA'

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

```
%%sql
SELECT LAUNCH_SITE
FROM SPACEXTBL
WHERE LAUNCH_SITE LIKE 'CCA%'
LIMIT 5;
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1 od8lcg.databases.appdomain.cloud:31498/bludb Done.

launch_site

CCAFS LC-40

CCAFS LC-40

CCAFS LC-40

CCAFS LC-40

CCAFS LC-40

Total Payload Mass

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

```
%%sql
SELECT SUM(PAYLOAD_MASS__KG_) AS payload_mass_kg
FROM SPACEXTBL
WHERE Customer = 'NASA (CRS)';

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1
od8lcg.databases.appdomain.cloud:31498/bludb
Done.

payload_mass_kg

45596
```

Average Payload Mass by F9 v1.1

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

```
%%sql
SELECT AVG(PAYLOAD_MASS__KG_)
FROM SPACEXTBL
WHERE Booster_Version LIKE 'F9 v1.0%';
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1 od8lcg.databases.appdomain.cloud:31498/bludb Done.

1

340

First Successful Ground Landing Date

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

Hint:Use min function

```
%%sql
SELECT MIN(DATE) AS Date
FROM SPACEXTBL
WHERE LANDING__OUTCOME = 'Success (ground pad)';
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1 od8lcg.databases.appdomain.cloud:31498/bludb Done.

DATE

2015-12-22

Successful Drone Ship Landing with Payload between 4000 and 6000

List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

```
%%sql
SELECT BOOSTER_VERSION
FROM SPACEXTBL
WHERE LANDING__OUTCOME = 'Success (drone ship)'
   AND 4000 < PAYLOAD_MASS__KG_ < 6000;
* ibm db sat//ivl47663:***039930704 19ff 4afo boos fa21c41761d2 bs2io00109kgb1</pre>
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1 od8lcg.databases.appdomain.cloud:31498/bludb Done.

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

List the total number of successful and failure mission outcomes

```
%%sql
SELECT MISSION_OUTCOME, COUNT(MISSION_OUTCOME) AS TOTAL
FROM SPACEXTBL
GROUP BY MISSION_OUTCOME;
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1 od8lcg.databases.appdomain.cloud:31498/bludb Done.

mission_outcome	total
Failure (in flight)	1
Success	99
Success (payload status unclear)	1

Boosters Carried Maximum Payload

List the names of the booster_versions which have carried the maximum payload mass. Use a subquery

```
%%sql
SELECT DISTINCT BOOSTER VERSION
FROM SPACEXTBL
WHERE PAYLOAD_MASS__KG_ = (
    SELECT MAX(PAYLOAD_MASS__KG_)
    FROM SPACEXTBL);
 * ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1
od8lcg.databases.appdomain.cloud:31498/bludb
 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

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 SPACEXTBL
WHERE Landing__Outcome = 'Failure (drone ship)'
   AND YEAR(DATE) = 2015;

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1
od8lcg.databases.appdomain.cloud:31498/bludb
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 Landing Outcomes Between 2010-06-04 and 2017-03-20

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

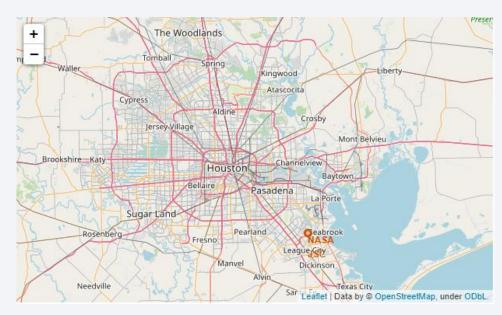
```
%%sql
SELECT LANDING__OUTCOME, COUNT(LANDING__OUTCOME) AS TOTAL
FROM SPACEXTBL
WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20'
GROUP BY LANDING__OUTCOME
ORDER BY TOTAL DESC
```

* ibm_db_sa://jyl47663:***@3883e7e4-18f5-4afe-be8c-fa31c41761d2.bs2io90l08kqb1 od8lcg.databases.appdomain.cloud:31498/bludb Done.

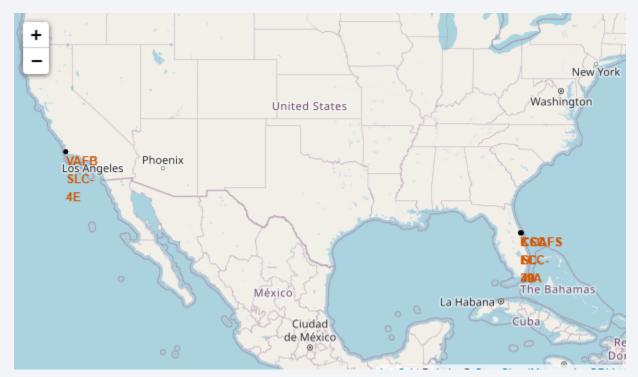
landing_outcome	total
No attempt	10
Failure (drone ship)	5
Success (drone ship)	5
Controlled (ocean)	3
Success (ground pad)	3
Failure (parachute)	2
Uncontrolled (ocean)	2
Precluded (drone ship)	1
Precluded (drone ship)	1



All launch sites on a map

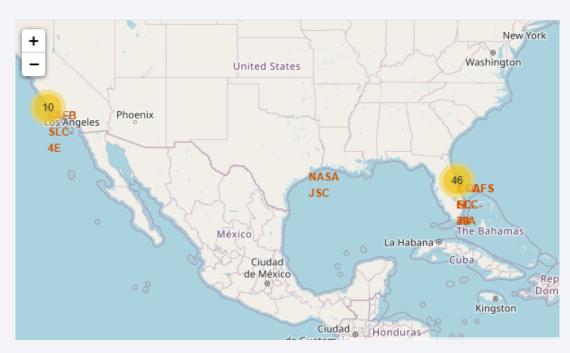


NASA Johnson Space Center's

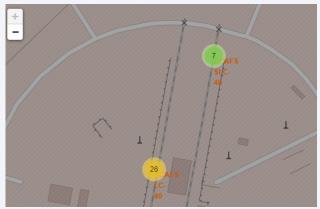


All launch sites are close to the Equator lane and to the cost. It's congruent with the axis of rotation of the earth (less fuel) and with security reasons (areas with less population).

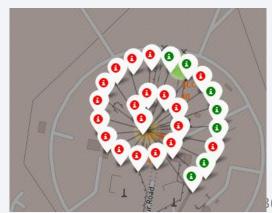
Success/Failed launches for each site



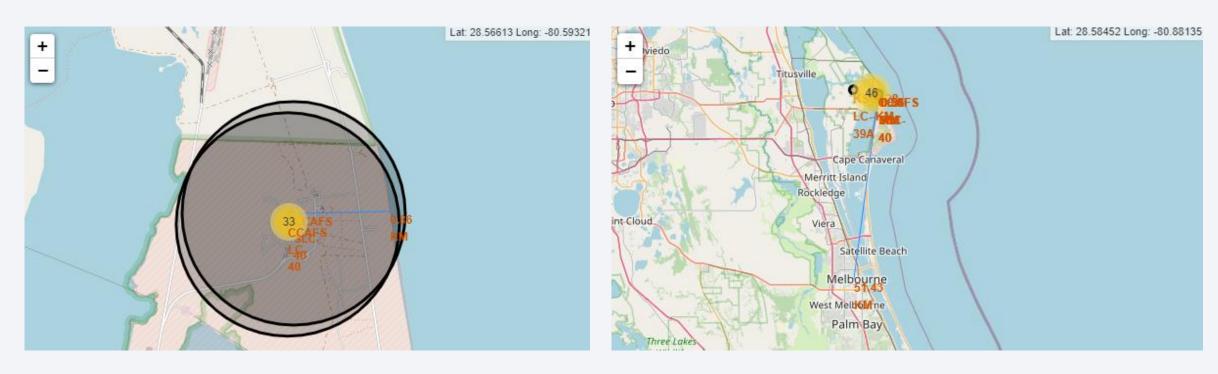








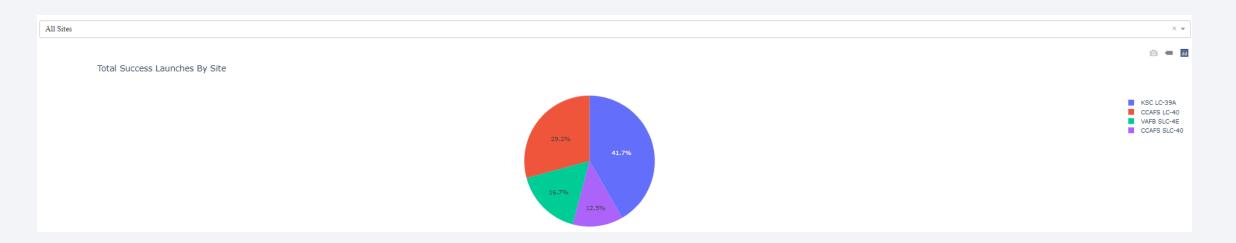
Distances between a launch site to its proximities



- Launch Sites are close to railways and highways, which allows easy transportation.
- They are not in close to proximity to cities, for security reasons (less population less probability of damage).

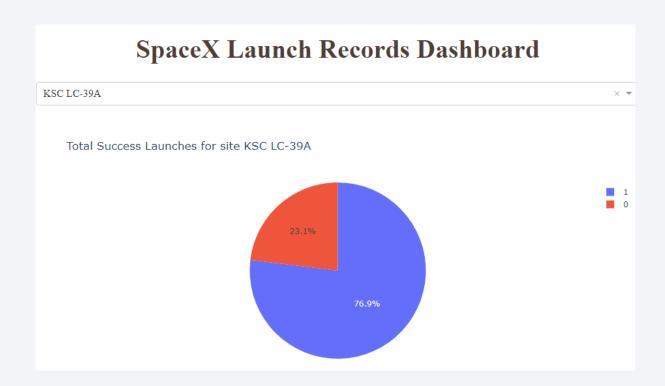


Success launches by all sites



- KSC LC-39A presents the most successful launches from all the sites (41,7%), followed by CCAFS LC-40.
- Lowest ratio of successful launches reported by CCAFS SLC-40.

Highest launch success ratio

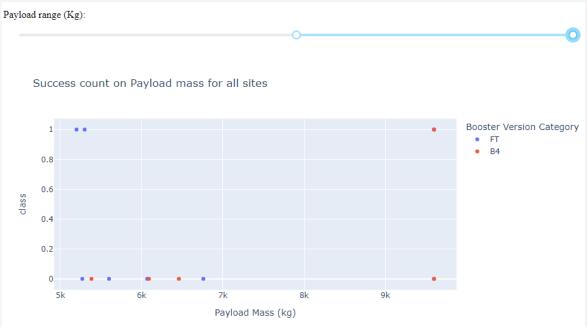


Observations:

- KSC LC-39A reported a 76,9% success and 23,1% failure rate.

Payload vs. Launch Outcome

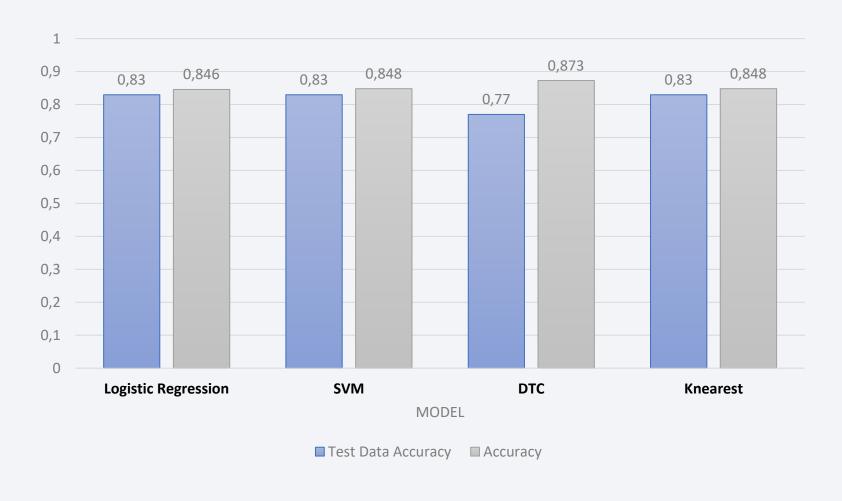




- Payload < 5000 kg → Highest booster landing success rate
- Payload > 5000 kg → Lowest booster landing success rate

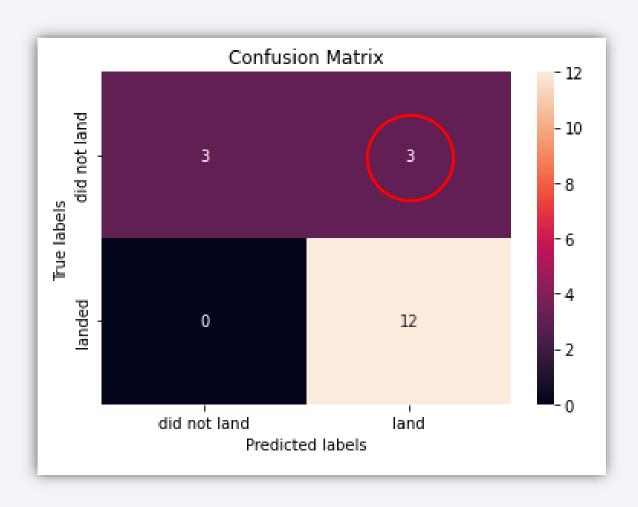


Classification Accuracy



- 3 models report the same accuracy score on the test set (0,83).
- The highest test data
 accuracy is reported by the
 Decision Tree (0,873),
 although obtain the lowest
 accuracy on the test set.

Confusion Matrix



- All the models report the same confusion matrix.
- We can detect the problem of False Positives cases (the models incorrectly predict the positive class in 3 occasions of the 18 samples).

Conclusions

- The model from this report can predict with an 83,3% of accuracy when SpaceX will successfully land the 1st stage booster.
- This results can help SpaceX to reduce costs, sacrificing \$15+ million predicting and reusing the 1st stage booster.

- Technicities:

- As the flight number increases, the first stage is more likely to land successfully.
- The highest success rates were reported for the ES-L1, GEO, HEO, and SSO orbits.
- Heavy payloads (higher than 9000) report positive landing rates for LEO, ISS and PO orbit.
- Launch success rate is increasing significantly since 2013.
- Launch sites are close to the Equator lane and to the cost (logistic and security issues?).
- KSC LC-39A presents the most successful launches from all the sites (41,7%).
- Highest booster landing success rate obtained when Payload < 5000 kg.

APPENDIX

Notebooks to recreate dataset, analysis and models:

https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab%201 %20Collecting%20%20the%20data.ipynb
https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab%202 %20Data%20wrangling.ipynb
https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab%204 %20Exploring%20and%20Preparing%20Data.ipynb
https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab%203 %20SQL.ipynb
https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab5 %20Launch%20Sites%20Analysis%20with%20Folim.ipynb
https://github.com/apnoguero/IBM-Data-Science-Capstone SpaceX/blob/main/Lab7 %20Machine%20Learning%20Prediction.ipynb

