

# Winning Space Race with Data Science

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### Outline

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### Introduction

In this capstone, we will predict if the Falcon 9 first stage will land successfully. 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. Therefore if we can determine if the first stage will land, we can determine the cost of a launch. This information can be used if an alternate company wants to bid against SpaceX for a rocket launch. In this lab, you will collect and make sure the data is in the correct format from an API.



# Methodology

### **Executive Summary**

### Data collection methodology:

• Data collection involves gathering information from specified sources to address a given problem or achieve certain objectives. This could be done via surveys, web scraping, databases, APIs, or other means. Careful planning is needed to ensure data integrity and relevance.

### Perform data wrangling

• This stage focuses on converting raw data into a format suitable for analysis. It includes data normalization, transformation, and feature engineering. Effective processing ensures that subsequent analyses are efficient and reliable.

### **Exploratory Data Analysis (EDA) Using Visualization and SQL:**

• EDA involves analyzing datasets to summarize their main characteristics, often using statistical graphics and other data visualization methods. SQL can be used to query and aggregate data, which is then visualized to uncover patterns, spot anomalies, and test hypotheses.

### **Building, Tuning, and Evaluating Classification Models:**

Building: Models are constructed based on the problem and data type. Tuning: Model parameters are optimized (e.g., using grid search) to improve performance.

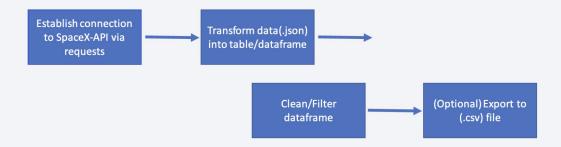
Evaluating: Models are evaluated using metrics like accuracy, precision, recall, and F1-score to determine their effectiveness.

### **Data Collection**

- How data sets were collected?
  - Gather and measure information on targeted variables
- Data collection process:



# Data Collection – SpaceX API



https://github.com/OlehLy/Applied-Data-Science-Capstone-Final-Exam/blob/main /Complete%20the%20EDA%20with%20 SQL.ipynb

### Data Collection - Scraping

https://github.com/OlehLy/Applied-Data-Science-Capstone-Final-Exam/blob/main/Complete%20the%20Data%20Collection%20with%20Web%20Scraping%20lab.ipynb



### **Data Wrangling**

### How data were processed?

· Cleaning and polishing possible messy/complex data sets for better handling

https://github.com/OlehLy/Applied-Data-Science-Capstone-Final-Exam/blob/main/Complete%20the%20Machine%20Learning%20Prediction%20lab.ipynb



### **EDA** with Data Visualization

### What needs to be done?

- Develop visuals and compile optical insights.
- Here's how:
- Generate scatterplots correlating Payload Mass/Flight Number with Launch Site/Orbit Type/Flight Number to illustrate dependencies.
- Construct bar charts comparing Success Rate with Orbit Type to evaluate its impact.
- Plot Launch Success against Year to observe trends over time.

https://github.com/OlehLy/Applied-Data-Science-Capstone-Final-Exam/blob/main/Complete%20the%20EDA%20with%20SQL.ipynb

### **EDA** with SQL

### What needs to be done?

- Employ heuristic methods to explore the database and make educated guesses about potential occurrences.
- Our inquiries are as follows:
- Provide a general overview of available landslides, particularly the first five entries starting with 'CCA'.
- Determine the number of successful and failed mission outcomes, listing the failed ones specifically in 2015.
- Identify the initial successful landing outcome on a drone ship.
- Specify, count, and rank outcomes between approximately 2010 and 2017.
- Calculate the average payload per booster for the 'F9v1.1' version and the total for boosters utilized by NASA (CRS).
- Determine the booster version with the maximum payload capacity.
- List the names of boosters that achieved successful ground pad landings with specific payloads.

https://github.com/OlehLy/Applied-Data-Science-Capstone-Final-Exam/blob/main/Complete%20the%20EDA%20with%20SQL. ipynb

### Build an Interactive Map with Folium

What has been accomplished?

• Implemented Map(Simple) with Icons, Circles markers, and Lines.

Why? To improve and consolidate visual representations of success and failure at each landing site.

https://github.com/OlehLy/Applied-Data-Science-Capstone-Final-Exam/blob/main/Interactive%20Visual%20Analytics%20with%20Folium%20lab.ipynb

### Build a Dashboard with Plotly Dash

What has been accomplished?

- Chosen Launch Site.
- Created Pie charts illustrating the percentage relation based on launch site.
- Developed Scatter graphs depicting the correlation between payload and success according to launch site.
- Why? To visually represent the efficacy of launch sites and payload mass.

https://github.com/OlehLy/Applied-Data-Science-Capstone-Final-Exam/blob/main/Interactive%20Visual%20Analytics%20with%20Folium%20lab.ipynb

# Predictive Analysis (Classification)



Standardizing/Transforming the data.

Dividing into test and training sets.

Initiating various Machine Learning algorithms using the training set. Evaluate accuracy through:

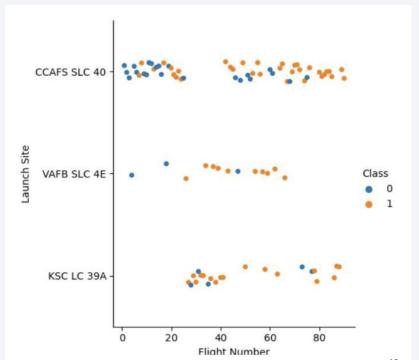
- R-score
- Confusion matrix (distinguishing true/false positives and negatives for landings)
- => Seek the optimal:
- Score
- -> Parameters for Machine Learning.

Based on accuracy:
Select the most suitable model.



# Flight Number vs. Launch Site

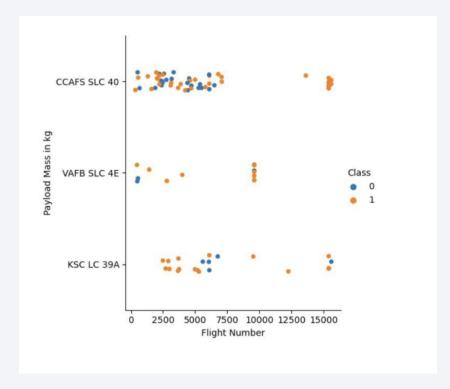
- First 25 Starts, in particular. at CCAFS SLC 40 were mostly failures
- Starts at KSC LC 39A gave them then information for success
- VAFB SLC 4E with most successful launches



# Payload vs. Launch Site

Following the successful launch of a payload weighing approximately 7500kg,

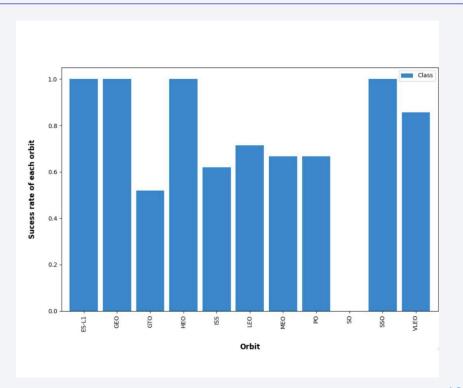
Launches carrying heavier payloads consistently achieved successful outcomes.



### Success Rate vs. Orbit Type

Launches to Sun-synchronous orbit (SSO) ended disastrously, with only one attempted launch.

However, launches to Earth-Sun Lagrange point 1 (ES-L1), Geostationary Earth Orbit (GEO), High Earth Orbit (HEO), and SSO were successful.

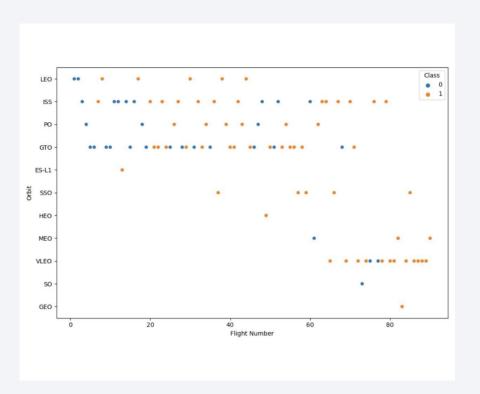


# Flight Number vs. Orbit Type

Initially, testing was conducted on four selected orbits.

Significant breakthroughs occurred after approximately the 20th flight, particularly with the International Space Station (ISS) and Geostationary Transfer Orbit (GTO).

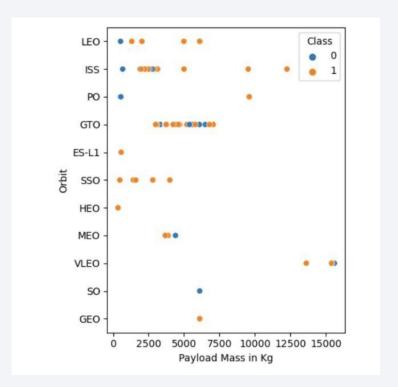
Subsequently, after approximately the 60th flight, attention was also directed towards other orbits, notably Very Low Earth Orbit (VLEO).



# Payload vs. Orbit Type

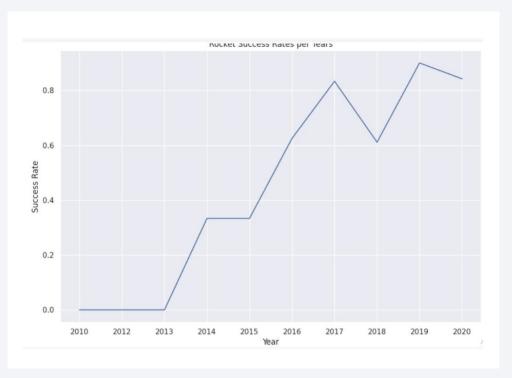
Started mostly with 2500-7500kg

Launches with higher payload > 7500kg rather successful



# Launch Success Yearly Trend

Since 2013 mostly improving
Since 2017 mostly >80% except for
Dip in Success in 2018



### All Launch Site Names

- Find the names of the unique launch sites
- Present your query result with a short explanation here

```
%sql SELECT DISTINCT launch_site FROM SPACEX
 * ibm db sa://vzv80836:***@764264db-9824-4b7c
   sqlite:///my_data1.db
Done.
 launch_site
 CCAFS LC-40
CCAFS SLC-40
 KSC LC-39A
 VAFB SLC-4E
```

### Launch Site Names Begin with 'CCA'

- Find 5 records where launch sites begin with `CCA`
- Present your query result with a short explanation here

| * ibm_db_sa:://vzv80836:***@764264db-9824-4b7c-82df-40d1b13897c2.bs2io90108kqb1od8lcg.databases.appdomain.cloud:32536/BLUDB sqlite:///my_data1.db Done. |           |                 |                 |   |                 |              |                    |                 |                     |
|---|-----------|-----------------|-----------------|---|-----------------|--------------|--------------------|-----------------|---------------------|
| DATE  | time_utc_ | booster_version | launch_site     | payload   | payload_masskg_ | orbit        | customer           | mission_outcome | landing_outcome     |
| 2010-06-<br>04  | 18:45:00  | F9 v1.0 B0003   | CCAFS LC-<br>40 | Dragon Spacecraft Qualification Unit                          | 0               | LEO          | SpaceX             | Success         | Failure (parachute) |
| 2010-12-  | 15:43:00  | F9 v1.0 B0004   | CCAFS LC-<br>40 | Dragon demo flight C1, two CubeSats, barrel of Brouere cheese | 0               | LEO<br>(ISS) | NASA (COTS)<br>NRO | Success         | Failure (parachute) |
| 2012-05-  | 07:44:00  | F9 v1.0 B0005   | CCAFS LC-<br>40 | Dragon demo flight C2   | 525             | LEO<br>(ISS) | NASA (COTS)        | Success         | No attempt          |
| 2012-10-<br>08  | 00:35:00  | F9 v1.0 B0006   | CCAFS LC-<br>40 | SpaceX CRS-1  | 500             | LEO<br>(ISS) | NASA (CRS)         | Success         | No attempt          |
| 2013-03-  | 15:10:00  | F9 v1.0 B0007   | CCAFS LC-       | SpaceX CRS-2  | 677             | LEO<br>(ISS) | NASA (CRS)         | Success         | No attempt          |

### **Total Payload Mass**

- Calculate the total payload carried by boosters from NASA
- Present your query result with a short explanation here



# Average Payload Mass by F9 v1.1

- Calculate the average payload mass carried by booster version F9 v1.1
- Present your query result with a short explanation here

```
%sql SELECT AVG(payload_mass__kg_) as "Total payload mass" FROM SPACEX WHERE booster_version LIKE 'F9 v1.1'

* ibm_db_sa://vzv80836:***@764264db-9824-4b7c-82df-40d1b13897c2.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:32536/BLUDB sqlite:///my_data1.db
Done.

Total payload mass

2928
```

# First Successful Ground Landing Date

- Find the dates of the first successful landing outcome on ground pad
- Present your query result with a short explanation here



### Successful Drone Ship Landing with Payload between 4000 and 6000

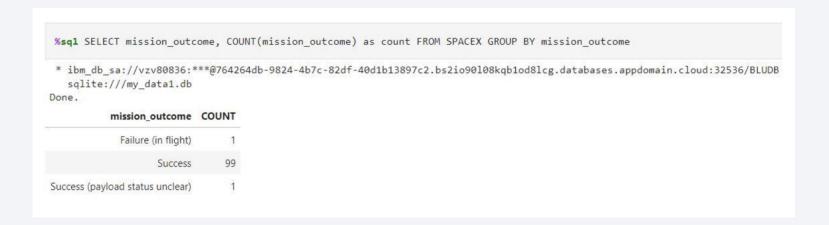
 List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

Present your query result with a short explanation here

| * ibm_db_sa://vzv80836:***@764264db-9824-4b7c-82df-40d1b13897c2.bs2io90108kqb1od8lcg.databases.appdomain.cloud:32536/BLUDB sqlite:///my_data1.db Done. |           |                 |             |                       |                 |       |                        |                 |                      |
|--|-----------|-----------------|-------------|-----------------------|-----------------|-------|------------------------|-----------------|----------------------|
| DATE   | time_utc_ | booster_version | launch_site | payload               | payload_masskg_ | orbit | customer               | mission_outcome | landing_outcome      |
| 2016-05-06   | 05:21:00  | F9 FT B1022     | CCAFS LC-40 | JCSAT-14              | 4696            | GTO   | SKY Perfect JSAT Group | Success         | Success (drone ship) |
| 2016-08-14   | 05:26:00  | F9 FT B1026     | CCAFS LC-40 | JCSAT-16              | 4600            | GTO   | SKY Perfect JSAT Group | Success         | Success (drone ship) |
| 2017-03-30   | 22:27:00  | F9 FT B1021.2   | KSC LC-39A  | SES-10                | 5300            | GTO   | SES                    | Success         | Success (drone ship) |
| 2017-10-11   | 22:53:00  | F9 FT B1031.2   | KSC LC-39A  | SES-11 / EchoStar 105 | 5200            | GTO   | SES EchoStar           | Success         | Success (drone ship) |

### Total Number of Successful and Failure Mission Outcomes

- Calculate the total number of successful and failure mission outcomes
- Present your query result with a short explanation here



### **Boosters Carried Maximum Payload**

- List the names of the booster which have carried the maximum payload mass
- Present your query result with a short explanation here



### 2015 Launch Records

• List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015

Present your query result with a short explanation here

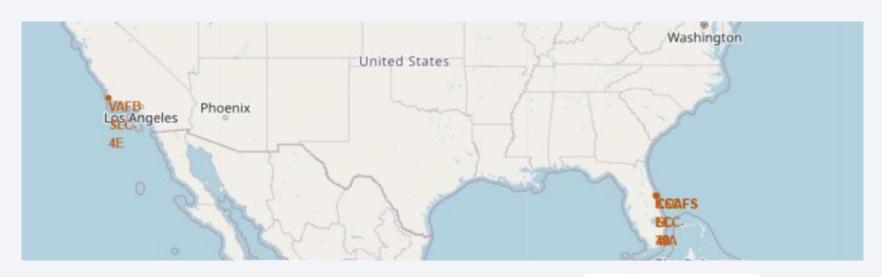


### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

- 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
- Present your query result with a short explanation here

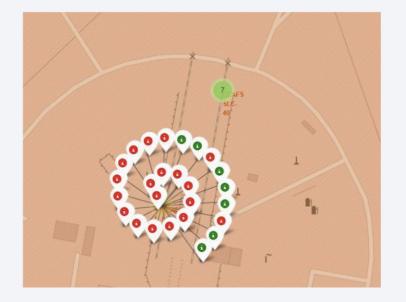






Launch Sites are situated on both the east and west coasts of the United States
Specifically in California and Florida.

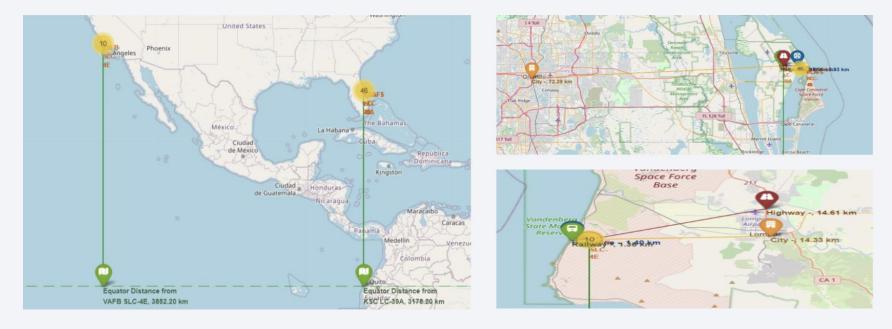
# CCAFS LC-40 CCAFS SLC-40 KSC LC-39A VAFB SLC-4E





Markers have been added to denote success and failure at each launch site.

To accommodate their size, they are consolidated in larger zooming views.



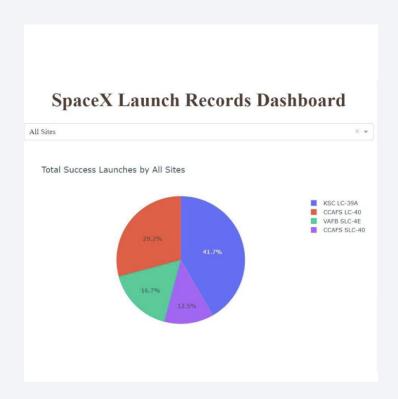
Launch Sites are directly at the coasts but ~ 10-15km away from the next big city and next highway, safe enough in case of an accident Roughly 3000-4000km from the equator where the highest earth rotational speed of the earth is and thus, better launch conditions



### Launch success count for all sites

KSC-LC-39A has highest share of Success

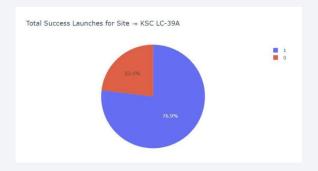
CCAFS.SLC-40 has lowest share of success



### The launch site with highest launch success ratio

Mixed success for FT boosters while overall best

B4 and B5 version are well





### Payload vs. Launch Outcome

In general, Booster category FT performs well for lower payload masses.

However, for payloads exceeding 5000kg, it tends to result in failure.

Booster version v1.1 performs poorly for mid-size payloads ranging from 2000kg to 7000kg.

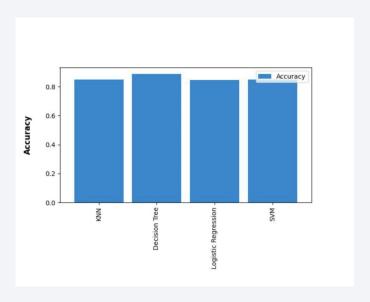






# **Classification Accuracy**

Decision Tree Model is the best with an accuracy with arround 91%
Others perform only minorly worse with accuracy rates > 84%

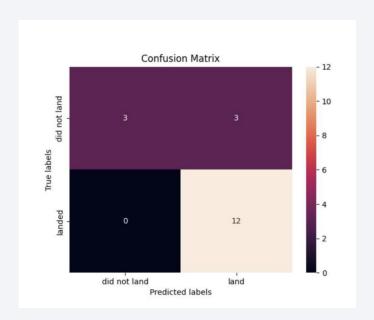


### **Confusion Matrix**

True/false versus Land/Not Land Matrix: Predicted landing outcomes for the test data (a subset of the original data) were assessed.

Regrettably, there were instances of True/Not-Land outcomes.

Nevertheless, the model achieved an overall accuracy of 15 out of 18 correct predictions.



### Conclusions

Orbits ES-L1, GEO, HEO, SSO have highest success rates

Success rate for launches increased over time

KSC LC-39A had the most successful launches

For higher payloads (> 5000kg) rather failure

Decision Tree is the best predictive Model

