

# Winning Space Race with Data Science

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## **Outline**

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

## **Executive Summary**

This presentation uses publicly available data to evaluate the success rate of the Space X program using the Falcon 9 launcher. Data science methods are used to collect and wrangle the data.

Afterwards the data is analysed and visualised. Further more predictive models were proposed and evaluated.

#### Introduction

- Project background and context
  - Space x is the cheapest space travel provider because it re-uses the first stage. Therefore if we determine that the first stage would land, then we can determine the cost of the launch. The aim of this project is to start a new company Space Y that can compite with Space X.
- Problems you want to find answers
  - Determine the price of each launch of Space X
  - Whether or not the first stage is re-used or not
  - Use machine learning and publicly available data to perform the tasks



## Methodology

## **Executive Summary**

The data are collected from Space X REST API and from Wiki pages using web scrabing methods. The collected data was wrangles and cleaned up. The cleaned data analysed and visualised using exploratory data analysis (EDA) and SQL approaxches. In addition Folium and Plotly Dash methods are used to interactively analyse the data. Finally predictive analysis was performed using classification models.

#### **Data Collection**

Space x data is gathered from Space X REST API. Information about the rocket used, payload delivered, launcher specification, landing specification and landing outcome are gathered. The goal is to predict whether space x will attempt to land a rocket or not. The data were collected in three steps:

- Step 1: Request and parse the SpaceX launch data using the GET request
- Step 2: Filter the dataframe to only include Falcon 9 launches
- Step 3: Data Wrangling Dealing with Missing Values

## Data Collection – SpaceX API

- Request and parse the Space X launch data using the GET request from Space X API and clean the data.
- Below the GitHub URL of the complete Space X API calls and data cleaning

https://github.com/Tadiwosz/Applied
-Data-ScienceCapstone/blob/main/jupyter-labsspacex-data-collection-api.ipynb

```
Task 1: Request and parse the SpaceX launch data using the GET request

To make the requested JSON results more consistent, we will use the following static response object for this project:

In [26]: M static_json_url='https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/datasets/API_c

We should see that the request was successfull with the 200 status response code

In [27]: M response.status_code

Out[27]: 200

Now we decode the response content as a Json using _json() and turn it into a Pandas dataframe using _json_normalize()

In [28]: M # Use json_normalize meethod to convert the json result into a dataframe data = pd.json_normalize(response.json())

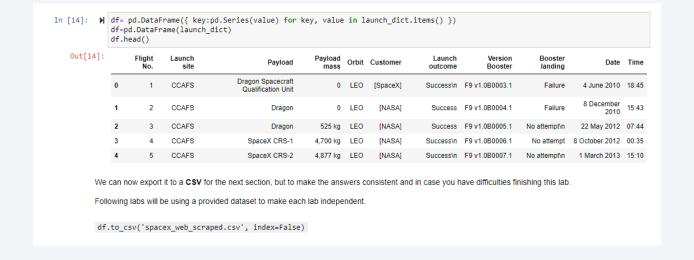
Using the dataframe data print the first 5 rows

In [29]: M # Get the head of the dataframe data.head()
```

## **Data Collection - Scraping**

- Webscrap Falcon 9 launch records using BeutifulSoup
  - Extract a Falcon 9 launch records HTML table from Wikipedia
  - Parse the table and convert it into a Pandas data frame
- Below is the GitHub URL of the completed web scraping notebook:

https://github.com/Tadiwosz/Appli ed-Data-Science-Capstone/blob/main/jupyter-labswebscraping.ipynb



## **Data Wrangling**

- Analyse the collected data,
  - Proportion of missing values in each attributes
  - Understand the data (number of launches per site, occurrence in each orbit, mission outcome)
  - Create landing outcome label (O or 1)
- perform some Exploratory Data Analysis (EDA) to find some patterns in the data and determine what would be the label for training supervised models.
- Below is the GitHub URL of my completed data wrangling related notebooks:

#### **EDA** with Data Visualization

- Summary of chart types and the reason of the plots
  - Visualize the effect of flight number and pay load mass on launch outcome (scatter plot)
  - Visualise the relationship between flight no. and launch site (scatter plot)
  - Visualize the relationship between Payload and Launch Site (scatter plot)
  - Visualize the relationship between success rate of each orbit type (bar chart)
  - Visualize the relationship between FlightNumber and Orbit type (scatter plot)
  - Visualize the launch success yearly trend (line chart)
  - Visualize the relationship between Payload and Orbit type (scatter plot)
  - Visualize the launch success yearly trend (line chart)
- Berlow is the GitHub URL of my completed EDA with data visualization notebook, as an external reference and peer-review purpose
- https://github.com/Tadiwosz/Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-dataviz.ipynb.jupyterlite.ipynb

## **EDA** with SQL

#### • List of the SQL querie:

- Display the names of the unique launch sites in the space mission
- Display 5 records where launch sites begin with the string 'CCA'
- Display the total payload mass carried by boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1¶
- · List the date when the first succesful landing outcome in ground pad was acheived.
- List the total number of successful and failure mission outcomes
- List the names of the booster\_versions which have carried the maximum payload mass. Use a subquery
- 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.
- 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
- Here the GitHub URL of mycompleted EDA with SQL notebook,

https://github.com/Tadiwosz/Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-sql-coursera sqllite.ipynb

## Build an Interactive Map with Folium

Circles, markesr and polyline were created and added to the map and used for:

- · To mark all launch sites
- Mark the success/failed launches for each site on the map
- Calculate the distances between a launch site to its proximities
- Here the GitHub URL of your completed interactive map with Folium map

https://github.com/Tadiwosz/Applied-Data-Science-Capstone/blob/main/lab\_jupyter\_launch\_site\_location.jupyterlite.ipynb

## Build a Dashboard with Plotly Dash

- Summarize what plots/graphs and interactions you have added to a dashboard
- Explain why you added those plots and interactions
- Add the GitHub URL of your completed Plotly Dash lab, as an external reference and peer-review purpose

The plotly lab did not work for me!

## Predictive Analysis (Classification)

- · The following steps were followed to build, evaluate, improve, and find the best performing classification model
- · create a column for the class
- Standardize the data
- Split into training data and test data
- Find best Hyperparameter for SVM, Classification Trees and Logistic Regression
- Find the method performs best using test d

#### Detailed steps:

Split the data to training and test data  $\rightarrow$  standardize the training data  $\rightarrow$  choodse the model  $\rightarrow$  Set model parameters  $\rightarrow$  grid search  $\rightarrow$  fit the model using the training data  $\rightarrow$  predict using the model  $\rightarrow$  Evaluate the accuracy using confusion matrix and relevant scores

Here is the GitHub URL of my completed predictive analysis lab,

https://github.com/Tadiwosz/Applied-Data-Science-Capstone/blob/main/SpaceX Machine Learning Prediction Part 5.jupyterlite.ipynb

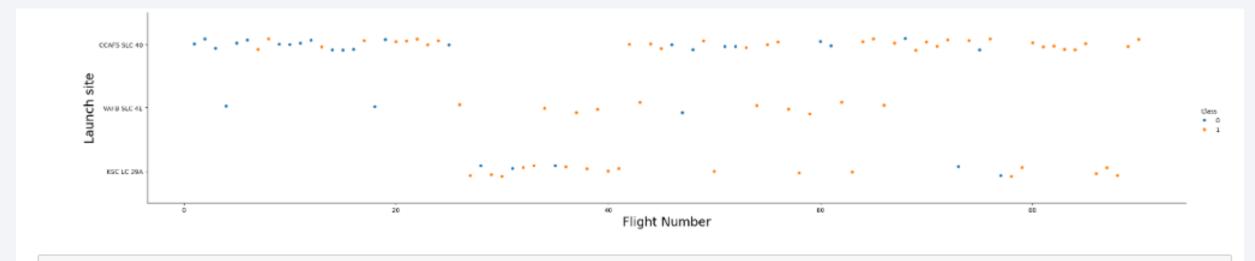
## Results

Logistic_Reg         0.833333           SVM         0.833333           Decision Tree         0.722222           KNN         0.833333	Method	Test Data Accuracy
Decision Tree 0.722222	Logistic_Reg	0.833333
	SVM	0.833333
KNN 0.833333	Decision Tree	0.722222
	KNN	0.833333

Best performers: Logistic reg, SVM and KNN

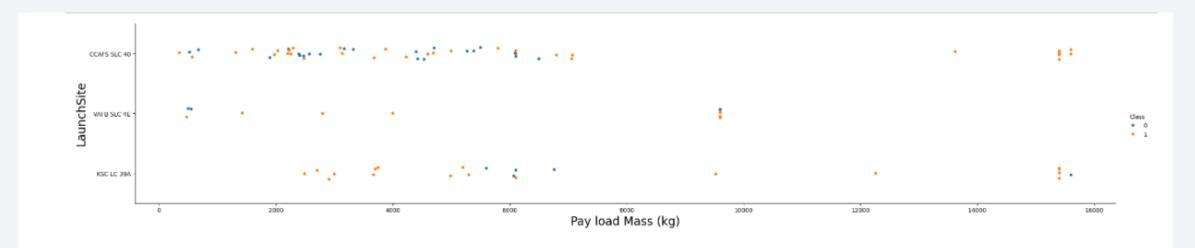


## Flight Number vs. Launch Site



Now try to explain the patterns you found in the Flight Number vs. Launch Site scatter point plots. In all the launch sites, the success increases as the flight number increases.

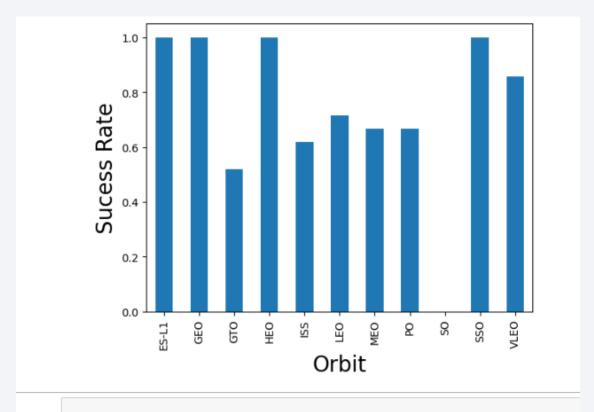
## Payload vs. Launch Site



Now if you observe Payload Vs. Launch Site scatter point chart you will find for the VAFB-SLC launchsite there are no rockets launched for heavypayload mass(greater than 10000).

As the payload increases, the success rate decrease in all the three launch sites.

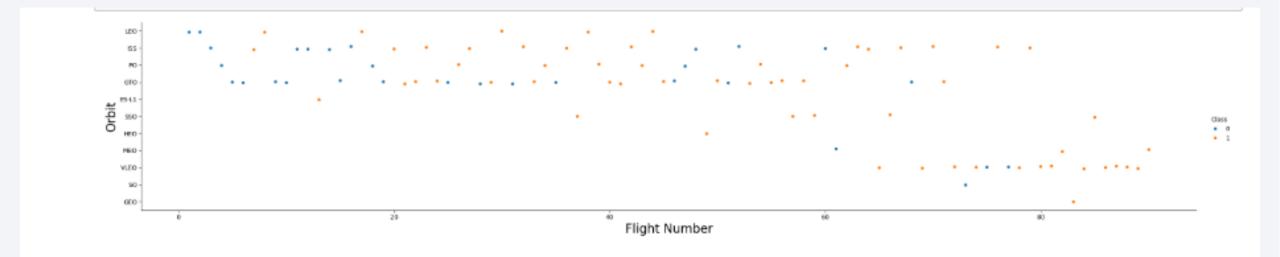
## Success Rate vs. Orbit Type



Analyze the ploted bar chart try to find which orbits have high sucess rate.

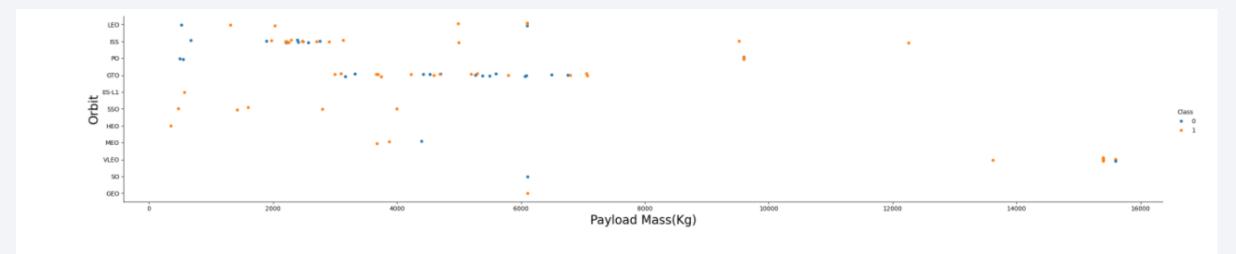
Orbits ES-L1, GEO, HEO, SSO have the highest sucess rate of 1.0.

## Flight Number vs. Orbit Type



You should see that in the LEO orbit the Success appears related to the number of flights; on the other hand, there seems to be no relationship between flight number when in GTO orbit.

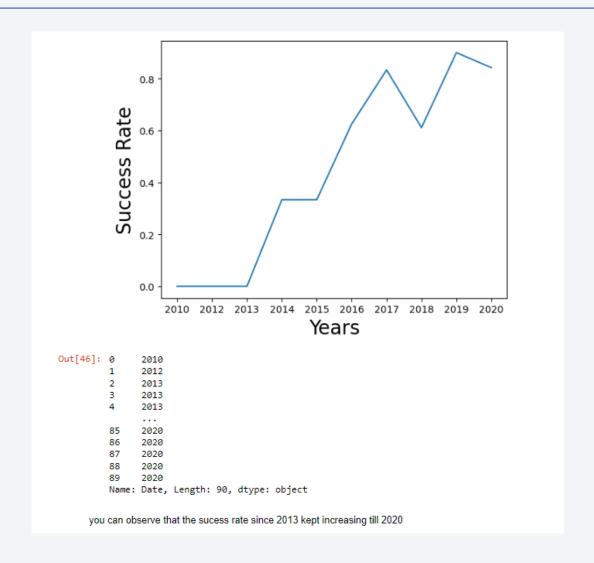
## Payload vs. Orbit Type



With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS.

However for GTO we cannot distinguish this well as both positive landing rate and negative landing(unsuccessful mission) are both there here.

## Launch Success Yearly Trend



#### All Launch Site Names

#### Task 1 Display the names of the unique launch sites in the space mission [27]: %sql SELECT DISTINCT LAUNCH\_SITE as "Launch\_Sites" FROM SPACEXTBL; \* sqlite:///my\_data1.db Done. [27]: Launch\_Sites CCAFS LC-40 VAFB SLC-4E KSC LC-39A CCAFS SLC-40

## Launch Site Names Begin with 'CCA'

	Task 2 Display 5 re	cords where	launch sites begir	n with the stri	ng 'CCA'					
[28]:	%sql SELECT	* FROM 'SP	ACEXTBL' WHERE	Launch_Site	LIKE 'CCA%' LIMIT 5;					
	* sqlite:/ Done.	///my_data1.	db							
[28]:	Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG_	Orbit	Customer	Mission_Outcome	Landing_Outcome
	2010-06-04	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
	2010-12-08	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	Failure (parachute)
	2012-05-22	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
	2012-10-08	0:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
	2013-03-01	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

## **Total Payload Mass**

## Average Payload Mass by F9 v1.1

## First Successful Ground Landing Date

```
%sql SELECT MIN(Date) FROM 'SPACEXTBL' WHERE 'Landing_Outcome' = "Success (groundpad)";

* sqlite://my_data1.db
Done.

MIN(Date)

None
```

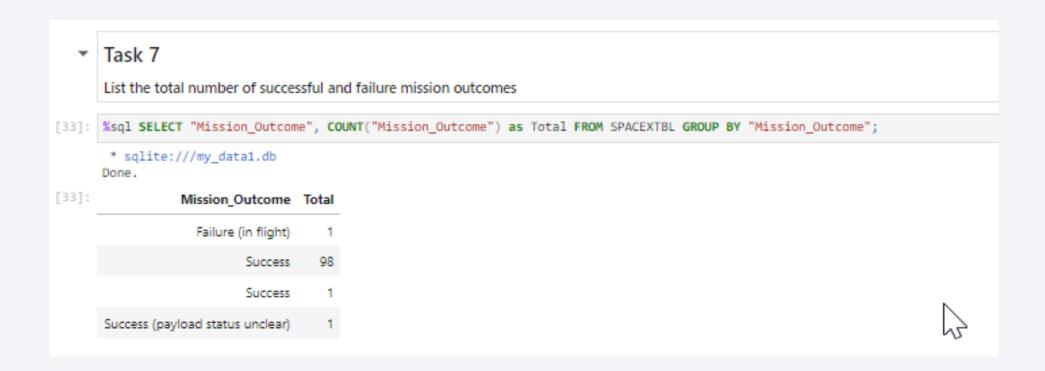
The sql command seems ok, but the results are not displayed

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

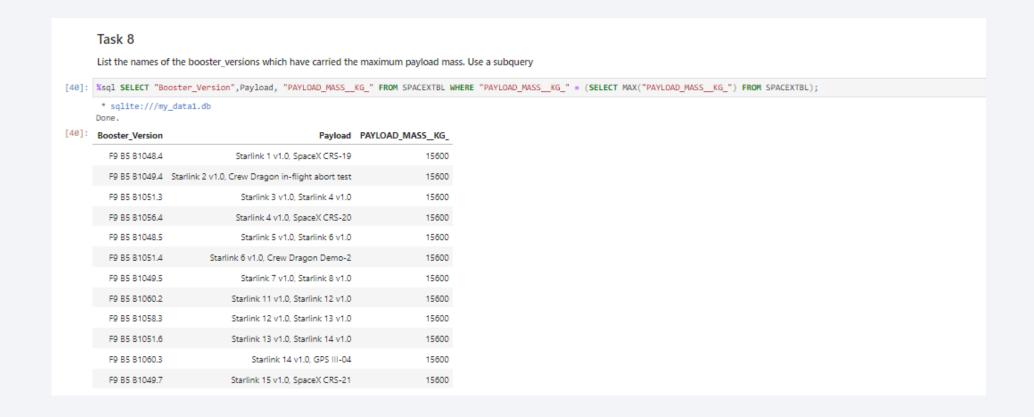
# Task 6 List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000 [52]: %sql SELECT DISTINCT Booster\_Version, Payload FROM SPACEXTBL WHERE "Landing \_Outcome" = "Success (drone ship)" AND PAYLOAD\_MASS\_\_KG\_ > 4000 AND PAYLOAD\_MASS\_\_KG\_ < 6000; \* sqlite://my\_datal.db Done.

The sql command seems ok, but the results are not displayed

#### Total Number of Successful and Failure Mission Outcomes



## **Boosters Carried Maximum Payload**



#### 2015 Launch Records

```
%sql SELECT substr(Date,6,2), substr(Date, 6, 2), "Booster_Version", "Launch_Site", Payload, "PAYLOAD_MASS__KG_", "Mission_Outcome", "Landing _Outcome" FROM SPACEXTBL WHERE substr(Date,0,5)='2015' AND "Landing _Outcome" = 'Failure (drone ship)';
```

The sql command seems ok, but the results are not displayed

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

# Task 10 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. [53]: \*\* sql SELECT \* FROM SPACEXTBL WHERE "Landing \_Outcome" LIKE 'Success%' AND (Date BETWEEN '04-06-2010' AND '20-03-2017') ORDER BY Date DESC; \* sqlite:///my\_datal.db

The sql command seems ok, but the results are not displayed

Done.



## Launch site location



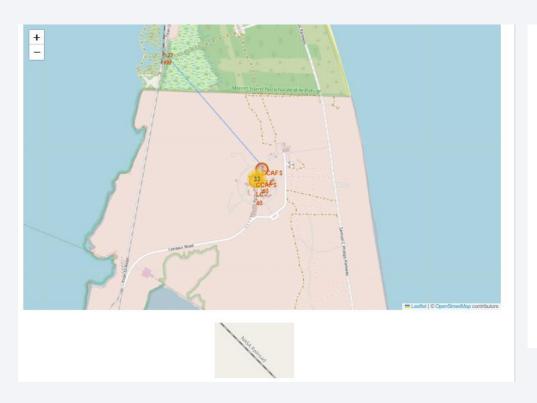
The four launch sites

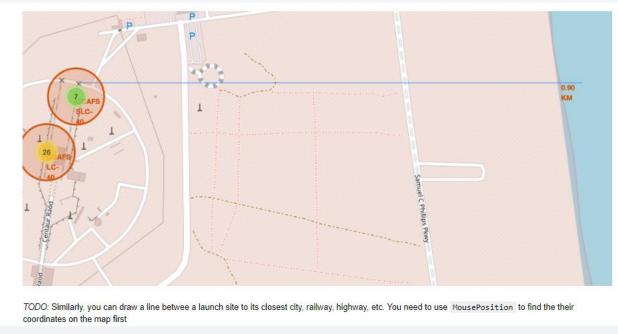






## Launch site proximity to important locations





The proximity of the lanch site to important locations such as railways and city centers can be determined using folium

## Use of colour markers to identify launch properties





#### < Dashboard Screenshot 1>

The plotly lab did not work for me!



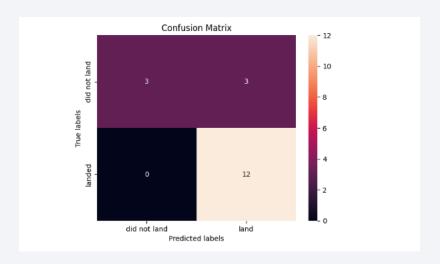
## **Classification Accuracy**

	0
Method	Test Data Accuracy
Logistic_Reg	0.833333
SVM	0.833333
Decision Tree	0.722222
KNN	0.833333

Logistic regression, SVM and KNN have equal accuracy of 0.833 while Decision tree has accuracy of 0.72

#### **Confusion Matrix**

• All the 3 best performing models have the same confusion matrix, The main problem is of the models is the false landing prediction.



#### **Conclusions**

- Real business case can be studied using publicly available data sets
- The success rate of Falcon 9 laches in space x program is proportional to the number of flights, the more flights the more the success.
- Increased pay load is linked to launch failures
- Some launch sites have better success rate than other sites. The reason for the difference need to be studied.
- Similarly there are differences in the success rate of launches to different orbits. Here also the reason need to be investigated.
- The success rate since 2012 is increasing till 2020. Is there a reason for it? This is something to be looked at
- Classification models could predict the mission outcome with high accuracy and these models can be used for similar tasks in the fuiture

## **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

