

# Winning Space Race with Data Science

<Name> 12.11.2023



#### **Outline**

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

#### **Executive Summary**

- Summary of methodologies
  - Exploratory Data Analysis (EDA) with SQL and data visualitaion
  - Data collection via the execution of API's and web scraping
  - Data wrangling
  - Building dashboard via Plotly Dash
  - Use of Machine Learning
- Summary of all results
  - Finding the best features to predict the success of launching via EDA and the use of machine learning algorithms

#### Introduction

- Project background and context
  - Working
- Problems you want to find answers
  - Main features that have an influence on a successful landing



# Methodology

#### **Executive Summary**

- Data collection methodology:
  - Space X API
  - Web scrapping
- Perform data wrangling
  - Using one hot encoding

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - How to build, tune, evaluate classification models

#### **Data Collection**

- Describe how data sets were collected.
  - The data sets were generated by running the SpaceX API and further information from Wikipedia through web scraping.
- You need to present your data collection process use key phrases and flowcharts

# Data Collection – SpaceX API

- Present your data collection with SpaceX REST calls using key phrases and flowcharts
- Add the GitHub URL of the completed SpaceX API calls notebook (must include completed code cell and outcome cell), as an external reference and peer-review purpose
- https://github.com/enes-13bot/coursera\_ibm\_applied\_data\_ science\_capstone\_enes/blob/8cf 6fd70b4978c110ad6192c1ac03ad 83e7f3f38/jupyter-labs-spacexdata-collection-api.ipynb

Request data via excectution of SpaceX API



Decode the response and the result in a dataframe



Filter the data in the dataframe which only include ,Falcon 9' launch



Replace missing values with the mean of the column ,PayLoadMess'

# **Data Collection - Scraping**

 Present your web scraping process using key phrases and flowcharts

- Add the GitHub URL of the completed web scraping notebook, as an external reference and peer-review purpose
- https://github.com/enes-13bot/coursera\_ibm\_applied\_data\_science\_capst one\_enes/blob/8cf6fd70b4978c110ad6192c 1ac03ad83e7f3f38/jupyter-labswebscraping.ipynb

Request the data from the Flacon9 Launch HTML page



Collect all relevant columns from the HTML table header



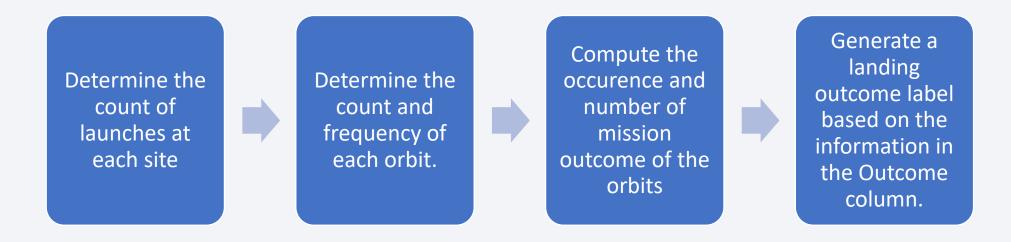
Filter the data in the dataframe which only include ,Falcon 9' launch



Generate a data frame by extracting information from the HTML tables related to launches.

# **Data Wrangling**

Describe how data were processed



- Add the GitHub URL of your completed data wrangling related notebooks, as an external reference and peer-review purpose
- https://github.com/enes-13bot/coursera\_ibm\_applied\_data\_science\_capstone\_enes/blob/8cf6fd70b4978c110ad 6192c1ac03ad83e7f3f38/labs-jupyter-spacex-Data%20wrangling.ipynb

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

- Summarize what charts were plotted and why you used those charts
  - Scatterplots and barplots were used to display the relationship between different pair of columns/feature
- Add the GitHub URL of your completed EDA with data visualization notebook, as an external reference and peer-review purpose
- https://github.com/enes-13bot/coursera\_ibm\_applied\_data\_science\_capstone\_enes/blob/ 8cf6fd70b4978c110ad6192c1ac03ad83e7f3f38/jupyter-labseda-dataviz.ipynb.jupyterlite.ipynb

#### **EDA** with SQL

- Using bullet point format, summarize the SQL queries you performed
  - Show 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 names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
  - 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.
- Add the GitHub URL of your completed EDA with SQL notebook, as an external reference and peer-review purpose
- https://github.com/enes-13-bot/coursera\_ibm\_applied\_data\_science\_capstone\_enes/blob/8cf6fd70b4978c110ad6192c1ac03ad83e7f3f38/jupyter4labs-eda-sql-coursera\_sqllite.ipynb

# Build an Interactive Map with Folium

- Summarize what map objects such as markers, circles, lines, etc. you created and added to a folium map
  - Folium maps annotate Launch Sites, both successful and unsuccessful landings, and illustrate proximity to key features such as Railway, Highway, Coast, and City. This visualization aids in comprehending the rationale behind the selection of launch sites and also depicts the spatial distribution of successful landings.

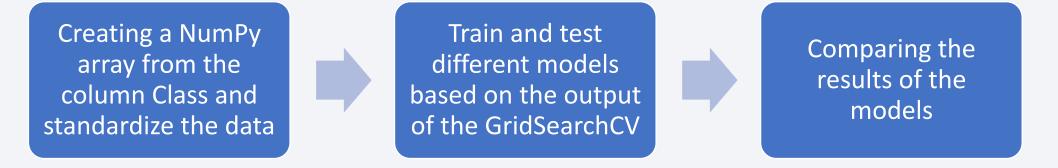
- Add the GitHub URL of your completed interactive map with Folium map, as an external reference and peer-review purpose
- https://github.com/enes-13bot/coursera\_ibm\_applied\_data\_science\_capstone\_enes/blob/8cf6fd70b4978c 110ad6192c1ac03ad83e7f3f38/lab\_jupyter\_launch\_site\_location.jupyterlite.ip ynb

#### 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

# Predictive Analysis (Classification)

• Summarize how you built, evaluated, improved, and found the best performing classification model



- Add the GitHub URL of your completed predictive analysis lab, as an external reference and peer-review purpose
- https://github.com/enes-13bot/coursera\_ibm\_applied\_data\_science\_capstone\_enes/blob/8cf6fd70b4978c110ad6192c1 acO3ad83e7f3f38/SpaceX\_Machine\_Learning\_Prediction\_Part\_5.jupyterlite.ipynb

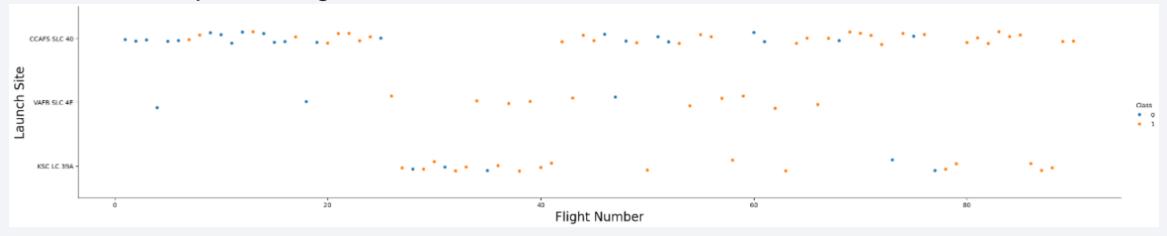
#### Results

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



# Flight Number vs. Launch Site

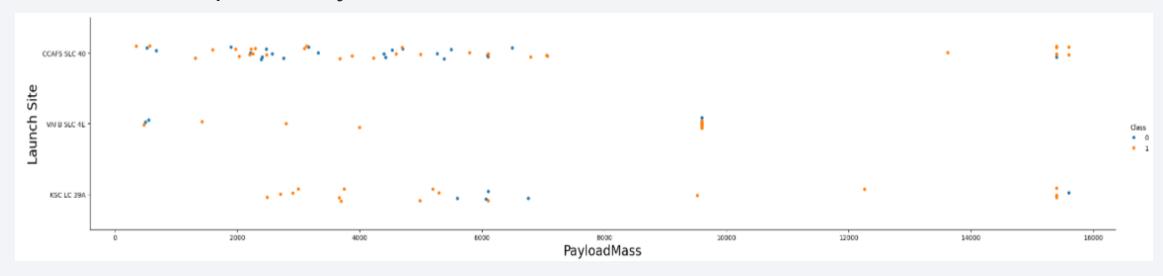
Show a scatter plot of Flight Number vs. Launch Site



- Show the screenshot of the scatter plot with explanations
  - Success rate grow over time

#### Payload vs. Launch Site

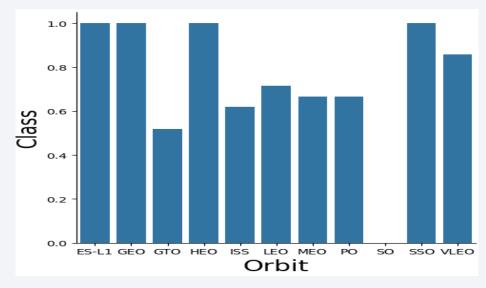
Show a scatter plot of Payload vs. Launch Site



- Show the screenshot of the scatter plot with explanations
  - Payload Mass over 8000 kg have a good success rate

# Success Rate vs. Orbit Type

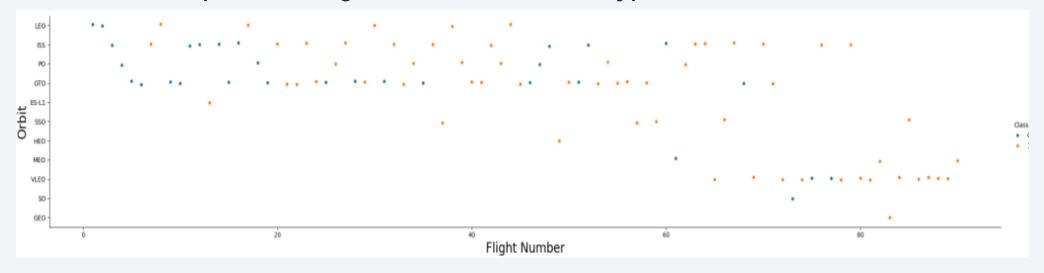
Show a bar chart for the success rate of each orbit type



- Show the screenshot of the scatter plot with explanations
  - ES-L1, GEO, HEO and SSO have the highest success rate

# Flight Number vs. Orbit Type

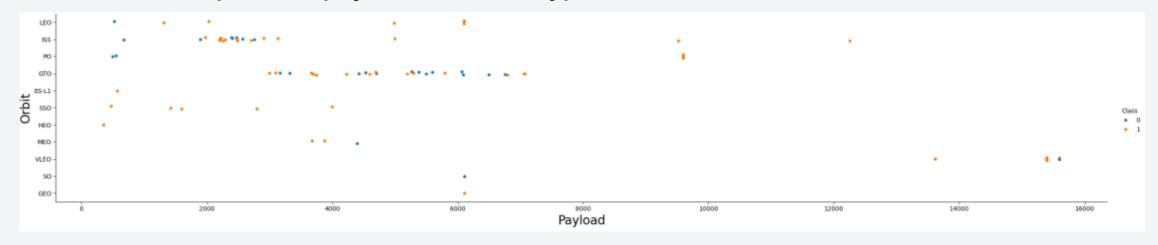
• Show a scatter point of Flight number vs. Orbit type



- Show the screenshot of the scatter plot with explanations
  - Success rate grow over time

# Payload vs. Orbit Type

• Show a scatter point of payload vs. orbit type



- Show the screenshot of the scatter plot with explanations
  - ISS orbit has a good success rate and the widest range of payload

# Launch Success Yearly Trend

• Show a line chart of yearly average success rate

Show the screenshot of the scatter plot with explanations

#### All Launch Site Names

- Find the names of the unique launch sites
- Present your query result with a short explanation here
  - Use the command 'distinct' to detect the unique launch sites

# Launch Site Names Begin with 'CCA'

Find 5 records where launch sites begin with `CCA`

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASSKG_	Orbit	Customer	Mission_Outcome	Landing_Outcome
2010-04-06	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
2010-08-12	15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0 1	.EO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
2012-05-22	07:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	.EO (ISS)	NASA (COTS)	Success	No attempt
2012-08-10	00:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500 I	.EO (ISS)	NASA (CRS)	Success	No attempt
2013-01-03	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	.EO (ISS)	NASA (CRS)	Success	No attempt

- Present your query result with a short explanation here
  - Use the command 'like' to filter only records which begin with 'CCA'

```
%sql Select * from SPACEXTABLE where Launch_Site like 'CCA%'
```

# **Total Payload Mass**

Calculate the total payload carried by boosters from NASA

Customer	sum(PAYLOAD_MASSKG_)
ABS Eutelsat	7759
AsiaSat	8963
Bulsatcom	3669
CONAE	3000
CONAE, PlanetIQ, SpaceX	3130

Present your query result with a short explanation here

```
%sql Select sum(PAYLOAD_MASS__KG_),customer from SPACEXTABLE group by customer
```

• Use a aggregation to calculate the total payload

# Average Payload Mass by F9 v1.1

Calculate the average payload mass carried by booster version F9 v1.1

```
avg(PAYLOAD_MASS__KG_)
2928.4
```

Present your query result with a short explanation here

```
%sql Select avg(PAYLOAD_MASS__KG_) from SPACEXTABLE where Booster_Version = 'F9 v1.1'
```

• Use a aggregation(avg) to calculate the average payload mass carried by booster version F9 v1.1

# First Successful Ground Landing Date

Find the dates of the first successful landing outcome on ground pad

```
min(Date)
2018-03-12
```

Present your query result with a short explanation here

```
%sql Select min(Date) from SPACEXTABLE where Landing_Outcome = 'Success'
* sqlite:///my_data1.db
```

• Use a aggregation(min) to find the dates of the first successful landing outcome on ground pad

#### 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

# F9 FT B1022 F9 FT B1026 F9 FT B1021.2 F9 FT B1031.2

Present your query result with a short explanation here

```
%sql Select Booster_Version from SPACEXTABLE where Landing_Outcome = 'Success (drone ship)' and PAYLOAD_MASS__KG_ between 4000 and 6000
```

• Use the commands 'filter' and 'between' to list the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

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

Calculate the total number of successful and failure mission outcomes

count(Mission_Outcome)	Mission_Outcome
1	Failure (in flight)
98	Success
1	Success
1	Success (payload status unclear)

• Present your query result with a short explanation here

%sql Select count(Mission\_Outcome), Mission\_Outcome from SPACEXTABLE group by Mission\_Outcome

# **Boosters Carried Maximum Payload**

List the names of the booster which have carried the maximum payload mass

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 B1060.2
F9 B5 B1058.3
F9 B5 B1060.3
F9 B5 B1060.3

• 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

Booster_Version	Launch_Site	month	Date	Landing_Outcome
F9 v1.1 B1012	CCAFS LC-40	10	2015-10-01	Failure (drone ship)
F9 v1.1 B1015	CCAFS LC-40	04	2015-04-14	Failure (drone ship)

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

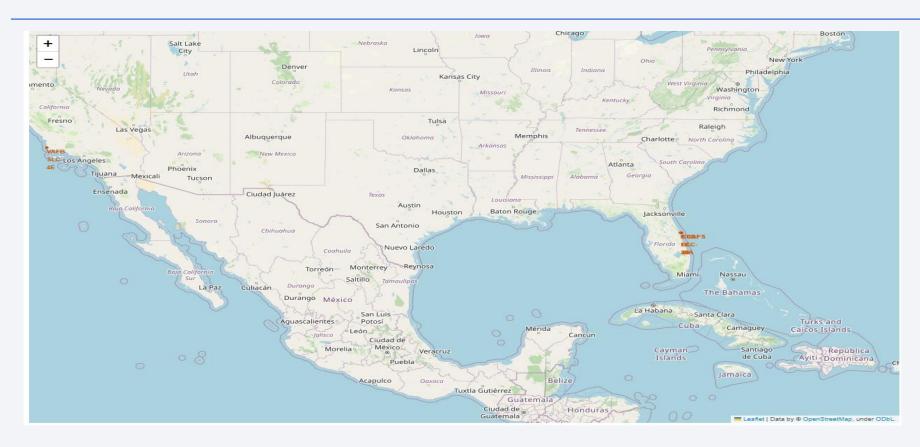
Landing_Outcome	م م الم يه count(landing_outcome)
No attempt	10
Success (ground pad)	5
Success (drone ship)	5
Failure (drone ship)	5
Controlled (ocean)	3
Uncontrolled (ocean)	2
Precluded (drone ship)	1
Failure (parachute)	1

• Present your query result with a short explanation here

```
%%sql
SELECT landing_outcome , count(landing_outcome) | FROM SPACEXTABLE
WHERE Date > '2010-06-04' AND Date < '2017-03-20'
GROUP BY landing_outcome
ORDER BY COUNT(landing_outcome) DESC;</pre>
```

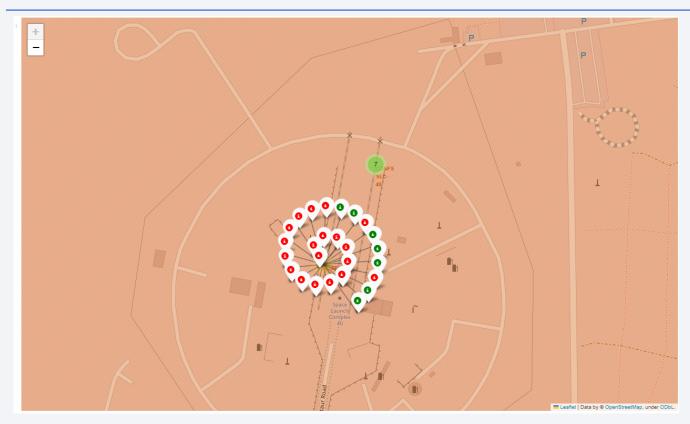


#### Launch Sites on the map



- Explain the important elements and findings on the screenshot
  - All displayed launch sites are close to the sea

#### Launch Markers



- Explain the important elements and findings on the screenshot
  - Marker which indicates if a launch was successful (green marker) or if the launch failed (red marker)

# <Folium Map Screenshot 3>

• Replace <Folium map screenshot 3> title with an appropriate title

• Explore the generated folium map and show the screenshot of a selected launch site to its proximities such as railway, highway, coastline, with distance calculated and displayed

• Explain the important elements and findings on the screenshot



#### < Dashboard Screenshot 1>

Replace < Dashboard screenshot 1> title with an appropriate title

• Show the screenshot of launch success count for all sites, in a piechart

• Explain the important elements and findings on the screenshot

#### < Dashboard Screenshot 2>

• Replace < Dashboard screenshot 2> title with an appropriate title

• Show the screenshot of the piechart for the launch site with highest launch success ratio

• Explain the important elements and findings on the screenshot

#### < Dashboard Screenshot 3>

Replace <Dashboard screenshot 3> title with an appropriate title

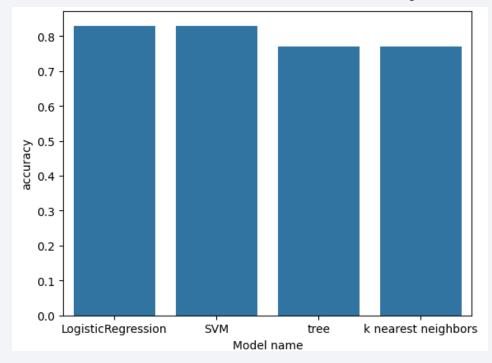
• Show screenshots of Payload vs. Launch Outcome scatter plot for all sites, with different payload selected in the range slider

• Explain the important elements and findings on the screenshot, such as which payload range or booster version have the largest success rate, etc.



# Classification Accuracy

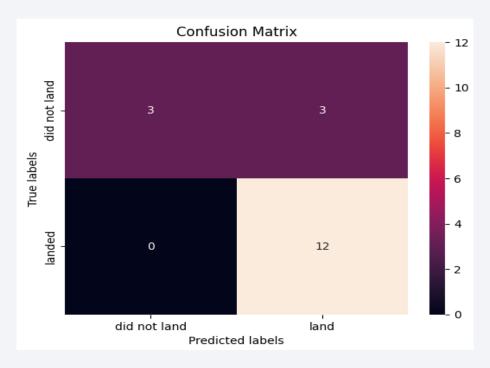
• Visualize the built model accuracy for all built classification models, in a bar chart



- Find which model has the highest classification accuracy
  - LogisticRegression 43
  - SVM

#### **Confusion Matrix**

- Show the confusion matrix of the best performing model with an explanation
- SVM confusion matrix



• The models erroneously forecasted 3 successful landings when the actual outcome was unsuccessful landings, indicating false positives.

#### Conclusions

- Launch success increase over time
- The orbits SSO, GEO, HEO and ES-L1 have a very high success rate

# **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

