



Space Race

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OUTLINE



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- Methodology
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 - Dashboard
- Discussion
 - Findings & Implications
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EXECUTIVE SUMMARY



- Summary of methodology
 - Complete the Data Collection API Lab and Data Wrangling
 - Exploratory Data Analysis and data visualization
 - EDA with SQL
 - Maps and Folium
 - Dash and Plotly
 - Prediction and analysis
- Summary of all results
 - Exploratory Data Analysis results
 - Maps and dashboards
 - Results

INTRODUCTION



- Project background
- 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 module, you will be provided with an overview of the problem and the tools you need to complete the course.
- Things to find
 - what are the main characteristics of a successful or failed landing?
 - how to allow SpaceX to achieve the best landing rate?

METHODOLOGY



- Data Collection
 - API and web Scrapping on Wikipedia
- Perform data wrangling
 - Remove unneeded columns
 - Encoding classification models
- Perform exploratory data analysis (EDA) using visualization and SQL
- Interactive Dashboards
- Perform predictive analysis

Data Collection with API

1. Getting Response from API

```
spacex_url="https://api.spacexdata.com/v4/launches/past"
response = requests.get(spacex_url)
```

2. Convert Response to JSON File

```
data = response.json()
data = pd.json_normalize(data)
```

3. Transform data

```
getLaunchSite(data)
getPayloadData(data)
getCoreData(data)
getBoosterVersion(data)
```

4. Create dictionary with data

```
launch_dict = {'FlightNumber': list(data['flight_number']),
               'Date': list(data['date']),
               'BoosterVersion':BoosterVersion,
               'PayloadMass':PayloadMass,
               'Orbit':Orbit,
               'LaunchSite':LaunchSite,
               'Outcome':Outcome,
               'Flights':Flights,
               'GridFins':GridFins,
               'Reused':Reused,
               'Legs':Legs,
               'LandingPad':LandingPad,
               'Block':Block,
               'ReuseCount':ReuseCount,
               'Serial':Serial,
               'Longitude': Longitude,
               'Latitude': Latitude}
```

5. Create dataframe

```
data = pd.DataFrame.from_dict(launch_dict)
```

6. Filter dataframe

```
data_falcon9 = data[data['BoosterVersion']!='Falcon 1']
```

7. Export to file

```
data_falcon9.to_csv('dataset_part_1.csv', index=False)
```

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/Applied%20Data%20Science%20Capstone.ipynb>

Scraping

1. Getting Response from HTML

```
response = requests.get(static_url)
```

2. Create BeautifulSoup Object

```
soup = BeautifulSoup(response.text, "html5lib")
```

3. Find all tables

```
html_tables = soup.findAll('table')
```

4. Get column names

```
for th in first_launch_table.findAll('th'):  
    name = extract_column_from_header(th)  
    if name is not None and len(name) > 0 :  
        column_names.append(name)
```

5. Create dictionary

```
launch_dict = dict.fromkeys(column_names)  
  
# Remove an irrelevant column  
del launch_dict['Date and time ( )']  
  
# Let's initial the launch_dict with each value to be an empty list  
launch_dict['Flight No.'] = []  
launch_dict['Launch site'] = []  
launch_dict['Payload'] = []  
launch_dict['Payload mass'] = []  
launch_dict['Orbit'] = []  
launch_dict['Customer'] = []  
launch_dict['Launch outcome'] = []  
  
# Added some new columns  
launch_dict['Version Booster'] = []  
launch_dict['Booster landing'] = []  
launch_dict['Date'] = []  
launch_dict['Time'] = []
```

6. Add data to keys

```
extracted_row = 0  
#Extract each table  
for table_number, table in enumerate(soup.findAll()  
    # get table row  
    for rows in table.findAll("tr"):  
        #check to see if first table heading is a  
        if rows.th:  
            if rows.th.string:  
                flight_number = rows.th.string.stri  
                flag = flight_number.isdigit()
```

See notebook for the rest of code

7. Create dataframe from dictionary

```
df = pd.DataFrame(launch_dict)
```

8. Export to file

```
df.to_csv('spacex_web_scraped.csv', index=False)
```

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/Applied%20Data%20Science%20Capstone.ipynb>

Data Wrangling

1. Calculate launches number for each site

```
df['LaunchSite'].value_counts()

CCAFS SLC 40    55
KSC LC 39A     22
VAFB SLC 4E     13
Name: LaunchSite, dtype: int64
```

2. Calculate the number and occurrence of each orbit

```
df['Orbit'].value_counts()

GTO    27
ISS    21
VLEO   14
PO      9
LEO      7
SSO      5
MEO      3
SO       1
ES-L1    1
HEO       1
GEO       1
Name: Orbit, dtype: int64
```

3. Calculate number and occurrence of mission outcome per orbit type

```
landing_outcomes = df['Outcome'].value_counts()
landing_outcomes

True ASDS    41
None None    19
True RTLS    14
False ASDS    6
True Ocean    5
None ASDS     2
False Ocean   2
False RTLS    1
Name: Outcome, dtype: int64
```

4. Create landing outcome label from Outcome column

```
landing_class = []
for key,value in df["Outcome"].items():
    if value in bad_outcomes:
        landing_class.append(0)
    else:
        landing_class.append(1)
df['Class']=landing_class
```

5. Export to file

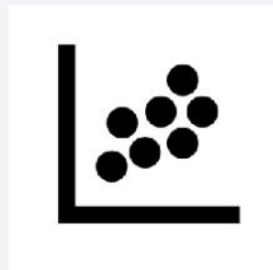
```
df.to_csv("dataset_part_2.csv", index=False)
```

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/labs-jupyter-spacex-Data-wrangling.ipynb>

EDA with Data Visualization

- Scatter Graphs

- Flight Number vs. Payload Mass
- Flight Number vs. Launch Site
- Payload vs. Launch Site
- Orbit vs. Flight Number
- Payload vs. Orbit Type
- Orbit vs. Payload Mass



Scatter plots show relationship between variables. This relationship is called the correlation.

- Bar Graph

- Success rate vs. Orbit

Bar graphs show the relationship between numeric and categoric variables.



- Line Graph

- Success rate vs. Year

Line graphs show data variables and their trends. Line graphs can help to show global behavior and make prediction for unseen data.



<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/Complete-the-EDA-with-Visualization-lab.ipynb>

EDA SQL

- We performed SQL queries to gather and understand data from dataset:
 - Displaying 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 successful landing outcome in ground pad was achieved.
 - 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.
 - 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 successful landing_outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/Complete-the-EDA-with-SQL.ipynb>

Maps

- Folium map object is a map centered on NASA Johnson Space Center at Houston, Texas
 - Red circle at NASA Johnson Space Center's coordinate with label showing its name (*folium.Circle*, *folium.map.Marker*).
 - Red circles at each launch site coordinates with label showing launch site name (*folium.Circle*, *folium.map.Marker*, *folium.features.DivIcon*).
 - The grouping of points in a cluster to display multiple and different information for the same coordinates (*folium.plugins.MarkerCluster*).
 - Markers to show successful and unsuccessful landings. **Green** for successful landing and **Red** for unsuccessful landing. (*folium.map.Marker*, *folium.Icon*).
 - Markers to show distance between launch site to key locations (railway, highway, coastway, city) and plot a line between them. (*folium.map.Marker*, *folium.PolyLine*, *folium.features.DivIcon*)
- These objects are created in order to understand better the problem and the data. We can show easily all launch sites, their surroundings and the number of successful and unsuccessful landings.

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/Interactive-Visual-Analytics-with-Folium-lab.ipynb>

Dashboard

- Dashboard has dropdown, pie chart, rangeslider and scatter plot components
 - Dropdown allows a user to choose the launch site or all launch sites (*dash_core_components.Dropdown*).
 - Pie chart shows the total success and the total failure for the launch site chosen with the dropdown component (*plotly.express.pie*).
 - Rangeslider allows a user to select a payload mass in a fixed range (*dash_core_components.RangeSlider*).
 - Scatter chart shows the relationship between two variables, in particular Success vs Payload Mass (*plotly.express.scatter*).

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/Interactive-Visual-Analytics-with-Folium-lab.ipynb>

Analysis

- **Data preparation**
 - Load dataset
 - Normalize data
 - Split data into training and test sets.
- **Model preparation**
 - Selection of machine learning algorithms
 - Set parameters for each algorithm to GridSearchCV
 - Training GridSearchModel models with training dataset
- **Model evaluation**
 - Get best hyperparameters for each type of model
 - Compute accuracy for each model with test dataset
 - Plot Confusion Matrix
- **Model comparison**
 - Comparison of models according to their accuracy
 - The model with the best accuracy will be chosen (see Notebook for result)

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project/blob/master/Complete-the-EDA-with-Visualization-lab.ipynb>

Results

<https://github.com/aelbadrawi/Python-Basics-for-Data-Science-Project>

