

# Winning Space Race with Data Science

<Andre Hanna> <11/27/2024>



#### Outline

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

### **Executive Summary**

#### Summary of methodologies

- Data collection –
- Data wrangling –
- Exploratory Data Analysis with Data Visualization
- Exploratory Data Analysis with SQL -
- Building an interactive map with Folium
- Building a Dashboard with Plotly Dash
- Predictive analysis (Classification)

#### Summary of all results

- Exploratory Data Analysis results
- Interactive analytics demo in screenshots –
- Predictive analysis results

#### Introduction

#### Project background and context

SpaceX has emerged as the most successful entity within the commercial space sector, significantly reducing the costs associated with space travel. The company markets its Falcon 9 rocket launches at a price point of \$62 million, in stark contrast to other providers whose services typically exceed \$165 million per launch. A substantial portion of these cost savings can be attributed to SpaceX's capability to reuse the first stage of its rockets. Consequently, the ability to predict whether the first stage will successfully land is critical, as it directly influences the overall cost-effectiveness of a launch. To address this, we will utilize publicly available data and machine learning models to forecast the likelihood of first-stage reuse by SpaceX.

#### Problems you want to find answers

- How do variables such as payload mass, launch site, number of lights, and orbits affect the success of the first stage landing?
- Does the rate of successful landings increase over the years?
- What is the best algorithm that can be used for binary classification in this case?



### Methodology

#### **Executive Summary**

- Data collection methodology:
  - Using SpaceX Rest API
  - Using Web Scrapping from Wikipedia
- Perform data wrangling
  - Filtering the data –
  - Dealing with missing values –
  - Using One Hot Encoding to prepare the data to a binary classification
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Building, tuning and evaluation of classification models to ensure the best results

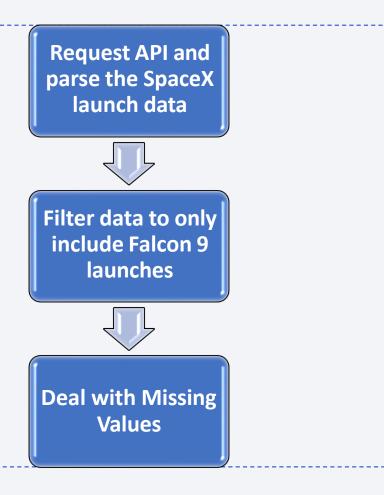
#### **Data Collection**

- The data collection process employed a combination of API requests from the SpaceX REST API and web scraping techniques to extract information from the relevant table in SpaceX's Wikipedia entry. This dual approach was necessary to gather comprehensive details about the launches, facilitating a more in-depth analysis.
- The following data columns were obtained through the SpaceX REST API: Flight Number, Date, Booster Version, Payload Mass, Orbit, Launch Site, Outcome, Flights, Grid Fins, Reused, Legs, Landing Pad, Block, Reused Count, Serial, Longitude, and Latitude.
- In contrast, the data columns sourced via Wikipedia web scraping include: Flight No., Launch Site, Payload, Payload Mass, Orbit, Customer, Launch Outcome, Booster Version, Booster Landing, Date, and Time.

### Data Collection – SpaceX API

- SpaceX offers a public API from where data can be obtained and then used
- This API was used according to the flowchart beside and then data is persisted.
- Github Link:

   https://github.com/andyh220/
   My First Test/blob/main/Data
   Science Capstone Project/ju
   pyter-labs-spacex-data-collection-api.ipynb



# Data Collection - Scraping

- Data from SpaceX launches can also be obtained from Wikipedia;
- Data are downloaded from Wikipedia according to the flowchart and then persisted.

 Github Link: <a href="https://github.com/andyh220/My">https://github.com/andyh220/My</a> First Test/ <a href="blob/main/Data-Science Capstone Project/jupyter-labs-webscraping.ipynb">https://github.com/andyh220/My</a> First Test/ <a href="blob/main/Data-Science Capstone Project/jupyter-labs-webscraping.ipynb">blob/main/Data-Science Capstone Project/jupyter-labs-webscraping.ipynb</a> Request the Falcon9
Launch Wiki page



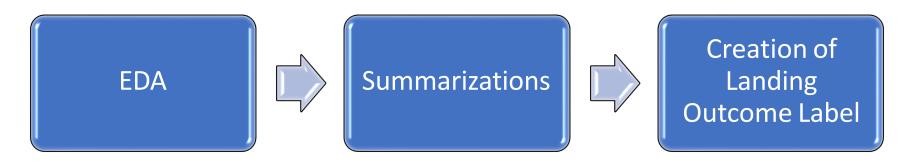
Extract all column/variable names from the HTML table header



Create a data frame by parsing the launch HTML tables

## Data Wrangling

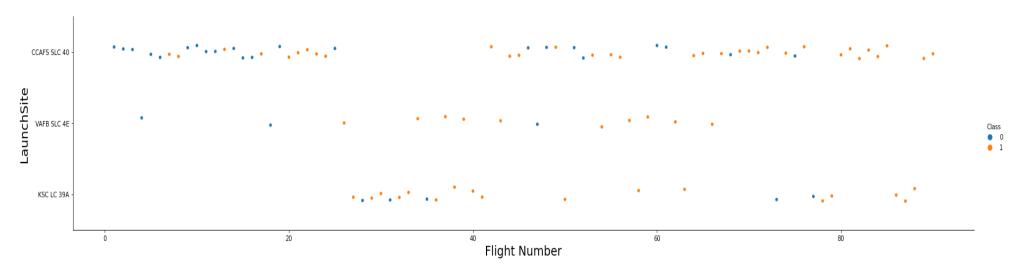
- Initially some Exploratory Data Analysis (EDA) was performed on the dataset.
- Then the summaries launches per site, occurrences of each orbit and occurrences of mission outcome per orbit type were calculated.
- Finally, the landing outcome label was created from Outcome column.



• Github: <a href="https://github.com/andyh220/My\_First\_Test/blob/main/Data\_Science\_Capstone\_Project/labs-jupyter-spacex-Data%20wrangling.ipynb">https://github.com/andyh220/My\_First\_Test/blob/main/Data\_Science\_Capstone\_Project/labs-jupyter-spacex-Data%20wrangling.ipynb</a>

#### EDA with Data Visualization

- To explore data, scatterplots and barplots were used to visualize the relationship between pair of features:
  - Payload Mass X Flight Number, Launch Site X Flight Number, Launch Site X Payload Mass,
     Orbit and Flight Number, Payload and Orbit



 Github: <u>https://github.com/andyh220/My First Test/blob/main/Data Science</u> Capstone Project/edadataviz.ipynb

### EDA with SQL

- The following SQL queries were performed:
  - Names of the unique launch sites in the space mission;
  - Top 5 launch sites whose name begin with the string 'CCA';
  - Total payload mass carried by boosters launched by NASA (CRS);
  - Average payload mass carried by booster version F9 v1.1;
  - Date when the first successful landing outcome in ground pad was achieved;
  - Names of the boosters which have success in drone ship and have payload mass between 4000 and 6000 kg;
  - Total number of successful and failure mission outcomes;
  - Names of the booster versions which have carried the maximum payload mass;
  - Failed landing outcomes in drone ship, their booster versions, and launch site names for in year 2015; and
  - Rank of the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20.
- Github:

# Build an Interactive Map with Folium

- Markers, circles, lines and marker clusters were used with Folium Maps
  - Markers indicate points like launch sites;
  - Circles indicate highlighted areas around specific coordinates, like NASA Johnson Space Center;
  - Marker clusters indicates groups of events in each coordinate, like launches in a launch site;
     and
  - Lines are used to indicate distances between two coordinates.

Github:
 <a href="https://github.com/andyh220/My">https://github.com/andyh220/My</a> First Test/blob/main/Data Science Capstone Project/lab jupyter launch site location.ipynb

# Build a Dashboard with Plotly Dash

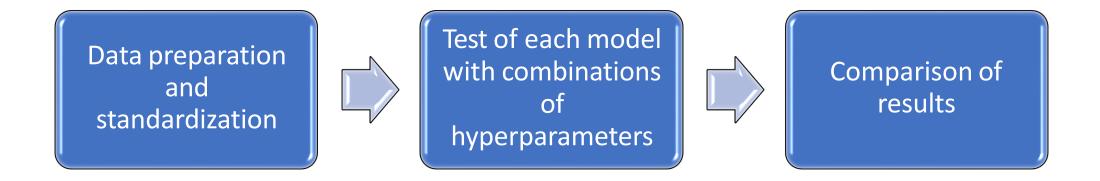
- The following graphs and plots were used to visualize data
  - Percentage of launches by site
  - Payload range
- This combination allowed to quickly analyze the relation between payloads and launch sites, helping to identify where is best place to launch according to payloads.

#### • Github:

https://github.com/andyh220/My First Test/blob/main/Data Science Capstone Project /spacex dash app.py

# Predictive Analysis (Classification)

 Four classification models were compared: logistic regression, support vector machine, decision tree and k nearest neighbors.



Github:

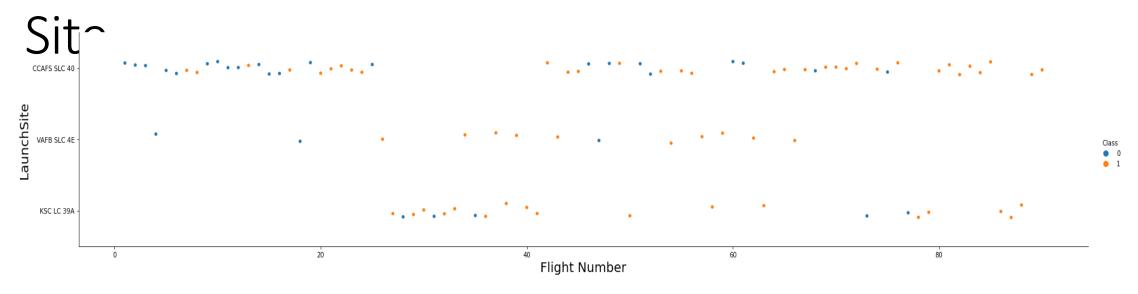
https://github.com/andyh220/My First Test/blob/main/Data Science Capstone Project/SpaceX Machine%2 OLearning%20Prediction Part 5.ipynb

#### Results

- Exploratory data analysis results:
  - Space X uses 4 different launch sites;
  - The first launches were done to Space X itself and NASA;
  - The average payload of F9 v1.1 booster is 2,928 kg;
  - The first success landing outcome happened in 2015 fiver year after the first launch;
  - Many Falcon 9 booster versions were successful at landing in drone ships having payload above the average;
  - Almost 100% of mission outcomes were successful;
  - Two booster versions failed at landing in drone ships in 2015: F9 v1.1 B1012 and F9 v1.1 B1015;
  - The number of landing outcomes became as better as years passed.

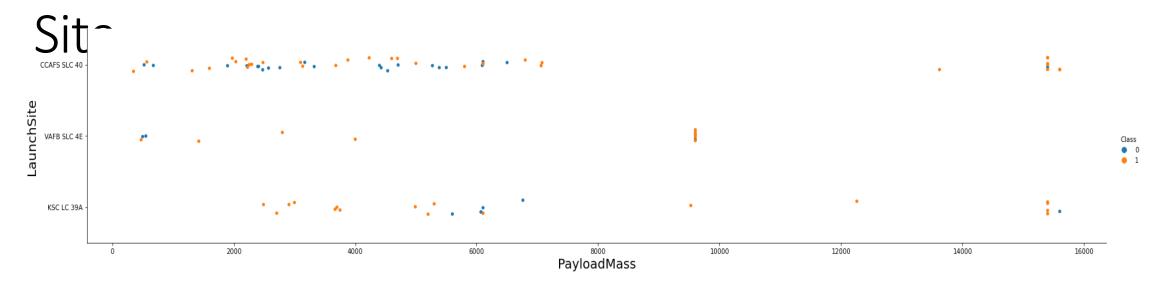


## Flight Number vs. Launch



- According to the plot above, it's possible to verify that the best launch site nowadays is CCAF5 SLC 40, where most of recent launches were successful;
- In second place VAFB SLC 4E and third place KSC LC 39A;
- It's also possible to see that the general success rate improved over time.

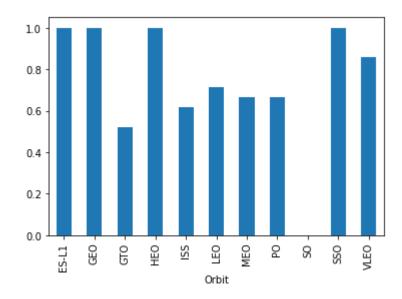
## Payload vs. Launch



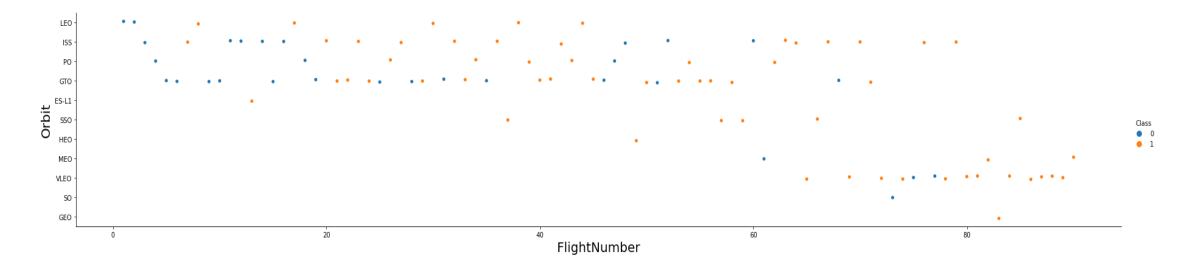
- Payloads over 9,000kg (about the weight of a school bus) have excellent success rate;
- Payloads over 12,000kg seems to be possible only on CCAFS SLC 40 and KSC LC 39A launch sites.

# Success Rate vs. Orbit Type

- The biggest success rates happens to orbits:
  - ES-L1;
  - GEO;
  - HEO; and
  - SSO.
- Followed by:
  - VLEO (above 80%); and
  - LFO (above 70%).

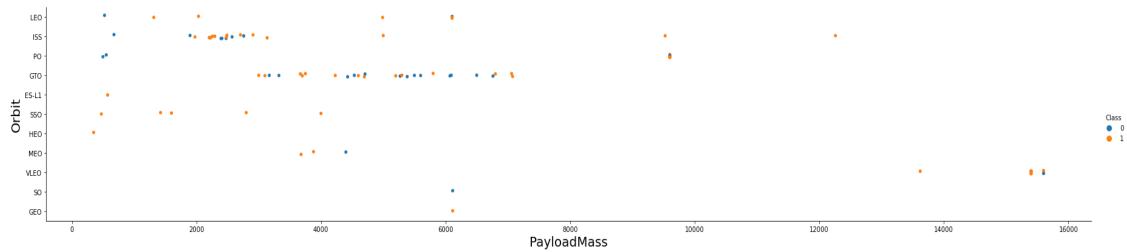


# Flight Number vs. Orbit Type



- Apparently, success rate improved over time to all orbits;
- VLEO orbit seems a new business opportunity, due to recent increase of its frequency.

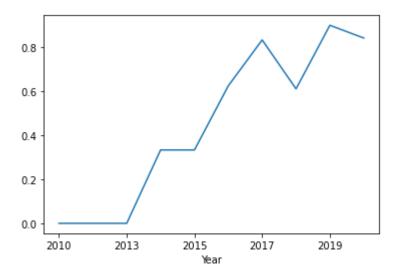
## Payload vs. Orbit Type



- Apparently, there is no relation between payload and success rate to orbit GTO;
- ISS orbit has the widest range of payload and a good rate of success;
- There are few launches to the orbits SO and GEO.

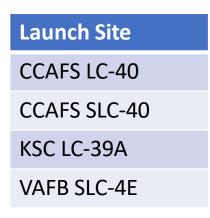
# Launch Success Yearly Trend

- Success rate started increasing in 2013 and kept until 2020;
- It seems that the first three years were a period of adjusts and improvement of technology.



#### All Launch Site Names

According to data, there are four launch sites:



 They are obtained by selecting unique occurrences of "launch\_site" values from the dataset.

# Launch Site Names Begin with 'CCA'

• 5 records where launch sites begin with `CCA`:

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	07:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
2012-10-08	00:35:00	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 attemp

Here we can see five samples of Cape Canaveral launches.

# Total Payload Mass

Total payload carried by boosters from NASA:

Total Payload (kg) 111.268

 Total payload calculated above, by summing all payloads whose codes contain 'CRS', which corresponds to NASA.

## Average Payload Mass by F9 v1.1

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

Avg Payload (kg)
2.928

• Filtering data by the booster version above and calculating the average payload mass we obtained the value of 2,928 kg.

# First Successful Ground Landing Date

First successful landing outcome on ground pad:

Min Date 2015-12-22

• By filtering data by successful landing outcome on ground pad and getting the minimum value for date it's possible to identify the first occurrence, that happened on 12/22/2015.

Successful Drone Ship Landing with Payload between 4000 and 6000

 Boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

<b>Booster Version</b>
F9 FT B1021.2
F9 FT B1031.2
F9 FT B1022
F9 FT B1026

 Selecting distinct booster versions according to the filters above, these 4 are the result.

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

Number of successful and failure mission outcomes:

Mission Outcome	Occurrences
Success	99
Success (payload status unclear)	1
Failure (in flight)	1

 Grouping mission outcomes and counting records for each group led us to the summary above.

# Boosters Carried Maximum Payload

Boosters which have carried the maximum payload mass

Booster Version ()	<b>Booster Version</b>
F9 B5 B1048.4	F9 B5 B1051.4
F9 B5 B1048.5	F9 B5 B1051.6
F9 B5 B1049.4	F9 B5 B1056.4
F9 B5 B1049.5	F9 B5 B1058.3
F9 B5 B1049.7	F9 B5 B1060.2
F9 B5 B1051.3	F9 B5 B1060.3

 These are the boosters which have carried the maximum payload mass registered in the dataset.

### 2015 Launch Records

 Failed landing outcomes in drone ship, their booster versions, and launch site names for in year 2015

<b>Booster Version</b>	Launch Site
F9 v1.1 B1012	CCAFS LC-40
F9 v1.1 B1015	CCAFS LC-40

The list above has the only two occurrences.

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

Ranking of all landing outcomes between the date 2010-06-04 and 2017-

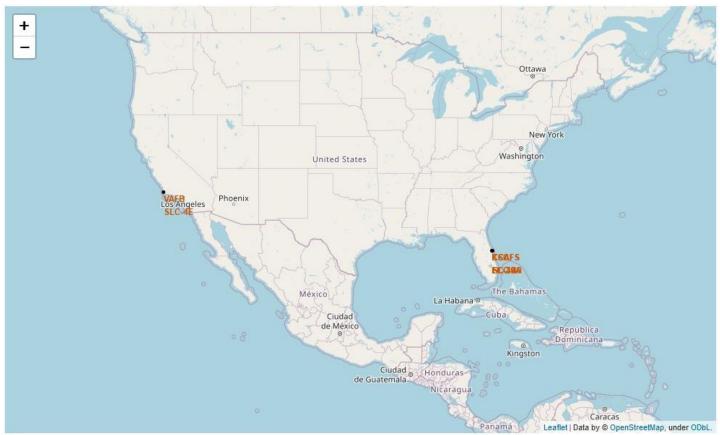
03-20:		
	Landing Outcome	Occurrences
	No attempt	10
	Failure (drone ship)	5
	Success (drone ship)	5
	Controlled (ocean)	3
	Success (ground pad)	3
	Failure (parachute)	2
	Uncontrolled (ocean)	2
	Precluded (drone ship)	1

This view of data alerts us that "No attempt" must be taken in account.



### All launch

sites



 Launch sites are near sea, probably by safety, but not too far from roads and railroads.

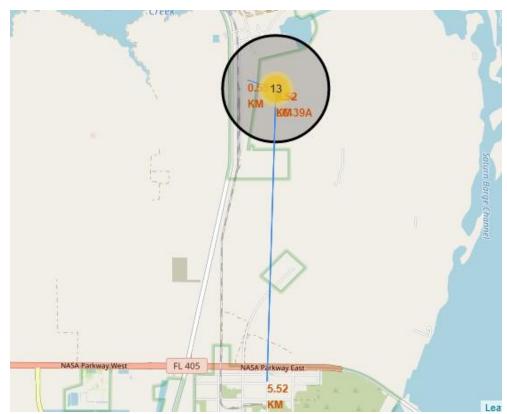
# Launch Outcomes by Site

• Example of KSC LC-39A launch site launch outcomes



· Green markers indicate successful and red ones indicate failure.

Logistics and Safety

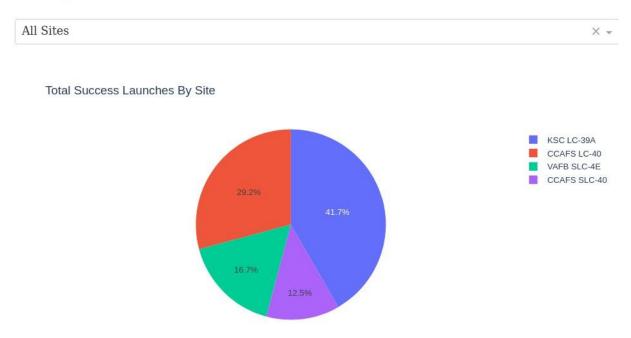


 Launch site KSC LC-39A has good logistics aspects, being near railroad and road and relatively far from inhabited areas.



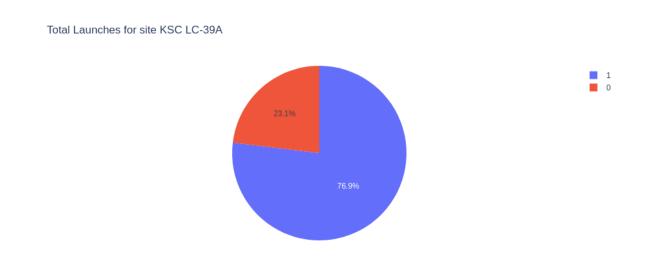
# Successful Launches by Site

**SpaceX Launch Records Dashboard** 



 The place from where launches are done seems to be a very important factor of success of missions.

### Launch Success Ratio for KSC LC-39A



• 76.9% of launches are successful in this site.

## Payload vs. Launch Outcome



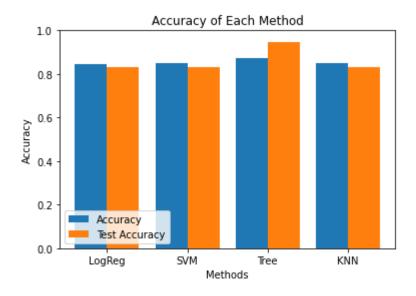
• Payloads under 6,000kg and FT boosters are the most successful combination.



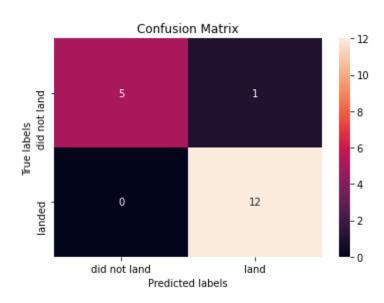
# Classification Accuracy

 Four classification models were tested, and their accuracies are plotted beside;

• The model with the highest classification accuracy is Decision Tree Classifier, which has accuracies over than 87%.



### Confusion Matrix of Decision Tree Classifier



 Confusion matrix of Decision Tree Classifier proves its accuracy by showing the big numbers of true positive and true negative compared to the false ones.

#### Conclusions

- Different data sources were analyzed, refining conclusions along the process;
- The best launch site is KSC LC-39A;
- Launches above 7,000kg are less risky;
- Although most of mission outcomes are successful, successful landing outcomes seem to improve over time, according the evolution of processes and rockets;
- Decision Tree Classifier can be used to predict successful landings and increase profits.

## Appendix

- As an improvement for model tests, it's important to set a value to np.random.seed variable;
- Folium didn't show maps on Github, so I took screenshots.

