

# Winning Space Race with Data Science

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#### Outline

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

#### **Executive Summary**

- Data collection methodology was Using API's and Webscraping. Performed data wrangling using
  Python packages Pandas and NumPy. Performed exploratory data analysis (EDA) using visualization
  and SQL. Performed interactive visual analytics using Folium and Plotly Dash. Performed predictive
  analysis using classification models
- Flights gradually increased their success rate with time. Higher payload mass seems more successful.
   ES-L1, GEO, HEO and SSO orbits have highest success rates/ VLEO orbits have highest Payload mass.
   Success rate of launches increases year after year. Booster version category FT is the most successful for lower payloads but less successful for higher payload masses. KSC LC-39A has the highest percentage of successful launches. Logistic the best performing model as it has an accuracy of 0.9444

#### Introduction

- 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.
- Objectives
  - To determine the price of each launch.
  - To determine if SpaceX will reuse the first stage



# Methodology

#### **Executive Summary**

- Data collection methodology:
  - Using API's ans Webscraping
- Perform data wrangling
  - Using Python packages Pandas and NumPy
- 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**

- Data was collected via 2 methods.
  - 1. Requesting and cleaning data from the SpaceX API
  - 2. Extracting Falcon 9 HTML tables from Wikipedia and converting it to a Data Frame

### Data Collection – SpaceX API

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

• <a href="https://github.com/ChiranNala/Applied-Data-Science-">https://github.com/ChiranNala/Applied-Data-Science-</a>
<a href="Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/Lab1">Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/Lab1</a> Collect <a href="ing-the-data.ipynb">ing-the-data.ipynb</a>

### **Data Collection - Scraping**

- 1. Request the Falcon9 Launch Wiki page from its URL
- 2. Extract all column/variable names from the HTML table header
- 3. Create a data frame by parsing the launch HTML tables

 https://github.com/ChiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/Web scraping Falcon 9 and Falcon Heavy.ipynb

# **Data Wrangling**

- Data Wrangling was done with Python packages Pandas and NumPy
  - Calculate the number of launches on each site
  - Calculate the number and occurrence of each orbit
  - Create a landing outcome label from Outcome column
- <a href="https://github.com/ChiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/">https://github.com/ChiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/</a> <a href="https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/">https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/</a> <a href="https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/">https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/</a> <a href="https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/">https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/</a> <a href="https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/">https://chiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/</a>

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

- Three bar plots, one scatter plot and one line plot were plotted
  - Bar plot 1: Visualize the relationship between success rate of each orbit type
  - Bar plot 2: Visualize the relationship between FlightNumber and Orbit type
  - Bar plot 3: Visualize the relationship between Payload and Orbit type
  - Scatter plot: Visualize the relationship between Payload and Orbit type
  - Line plot: Visualize the launch success yearly trend
- https://github.com/ChiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/Assignment Exploring and Preparing Data.ipynb

#### **EDA** with SQL

- 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 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.
- https://github.com/ChiranNala/Applied-Data-Science-Capstone/blob/55cf767746cbc4f740e5bf8a0983473facd4f156/SQL\_Notebook\_for\_Peer\_Assignment.ipynb

### Build an Interactive Map with Folium

- Mark all launch sites on a map using Circles and Markers
- Mark the success/failed launches for each site on the map using MarkerCluster
- Calculate the distances between a launch site to its proximities

To better visualize data spatially on a map

• <a href="https://github.com/ChiranNala/Applied-Data-Science-">https://github.com/ChiranNala/Applied-Data-Science-</a>
<a href="Capstone/blob/299f443b4997fd040a2eeae73df48c79d4fa0e83/Launch Sites Locations Analysis with Folium.ipynb">https://github.com/ChiranNala/Applied-Data-Science-</a>
<a href="Capstone/blob/299f443b4997fd040a2eeae73df48c79d4fa0e83/Launch Sites Locations Analysis with Folium.ipynb">https://github.com/ChiranNala/Applied-Data-Science-</a>
<a href="Capstone/blob/299f443b4997fd040a2eeae73df48c79d4fa0e83/Launch Sites Locations Analysis with Folium.ipynb">https://github.com/ChiranNala/Applied-Data-Science-</a>
<a href="cations Analysis with Folium.ipynb">cations Analysis with Folium.ipynb</a>

#### Build a Dashboard with Plotly Dash

- Pie charts and a scatter plot were created to answer the following questions
- Which site has the largest successful launches?
- Which site has the highest launch success rate?
- Which payload range(s) has the highest launch success rate?
- Which payload range(s) has the lowest launch success rate?
- Which F9 Booster version (v1.0, v1.1, FT, B4, B5, etc.) has the highest launch success rate?
- https://github.com/ChiranNala/Applied-Data-Science-Capstone/blob/a0029be32fd9edeb4a4a07c4972567f7559f02e7/spacex\_dash\_app%20(1).py

### Predictive Analysis (Classification)

- Performed exploratory Data Analysis and determine Training Labels
- Created a column for the class
- Standardized the data
- Split into training data and test data
- Found best Hyperparameter for SVM, Classification Trees and Logistic Regression
- Found the method performs best using test data
- https://github.com/ChiranNala/Applied-Data-Science-Capstone/blob/a0029be32fd9edeb4a4a07c4972567f7559f02e7/Assignment Machine Learning Prediction.ipynb

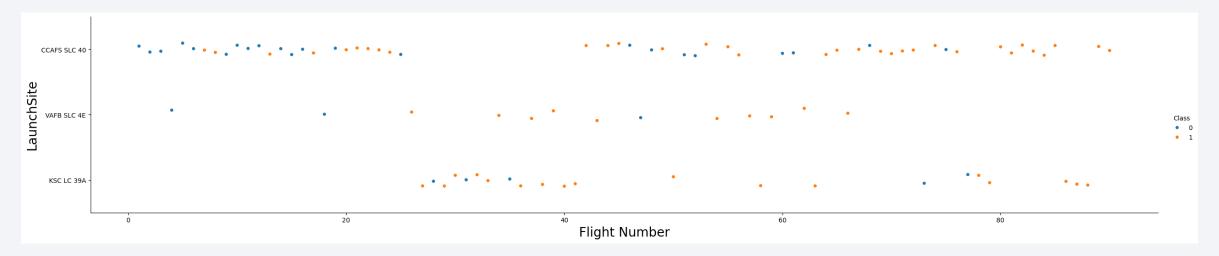
#### Results

- Exploratory data analysis results
  - Flights gradually increased their success rate with time
  - Higher payload mass seems more successful
  - ES-L1, GEO, HEO and SSO orbits have highest success rates
  - VLEO orbits have highest Payload mass
  - Success rates increase with time
- Predictive analysis results
  - Logistic regression, SVM and KNN show the same highest accuracy of 0.944
  - Decision tree shows lowest accuracy with accuracy 0.833



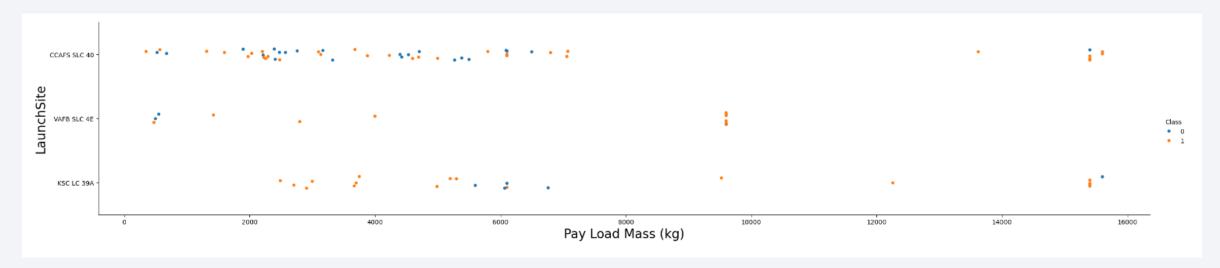
### Flight Number vs. Launch Site

- Scatter plot of Flight Number vs. Launch Site
- We can see that initial flights were unsuccessful and were from launch site CCAFS SLS
   40 and VAFB SLS 4E and none from KSC IC 39A
- In later flights we can see they are majorly successful and from Launch site and are from CCAFS SLS 40 and KSC IC 39A only



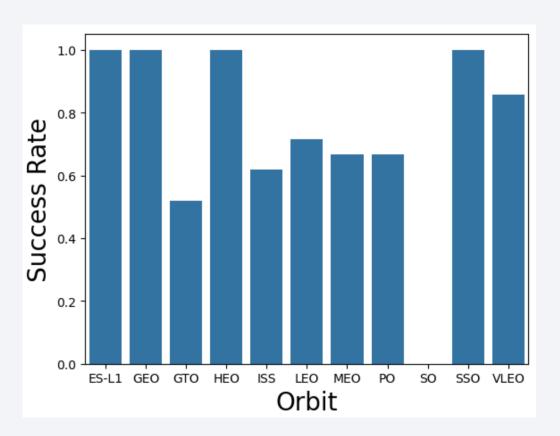
#### Payload vs. Launch Site

- Scatter plot of Payload vs. Launch Site
- Larger payloads are more successful and are launched from CCAFS SLS 40 and KSC IC 39A
- Smaller payloads are have a lower success rate and are launched from CCAFS SLS 40 and VAFB SLS 4E



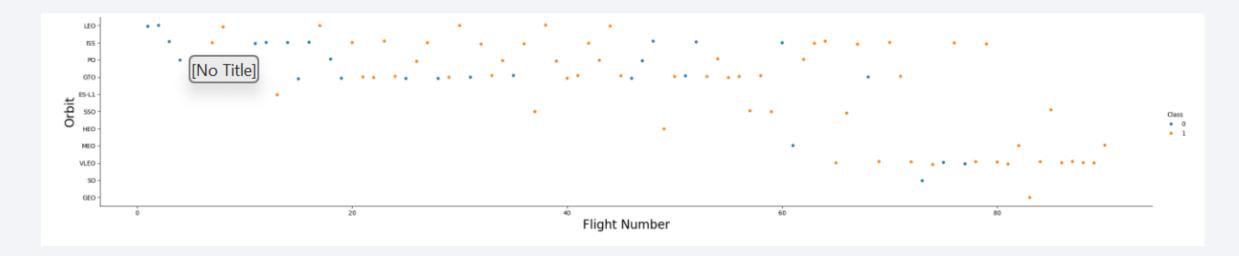
# Success Rate vs. Orbit Type

- Bar chart for the success rate of each orbit type
- ES-L1, GEO, HEO and SSO orbits have highest success rates
- SO, GTO and ISS have the lowest



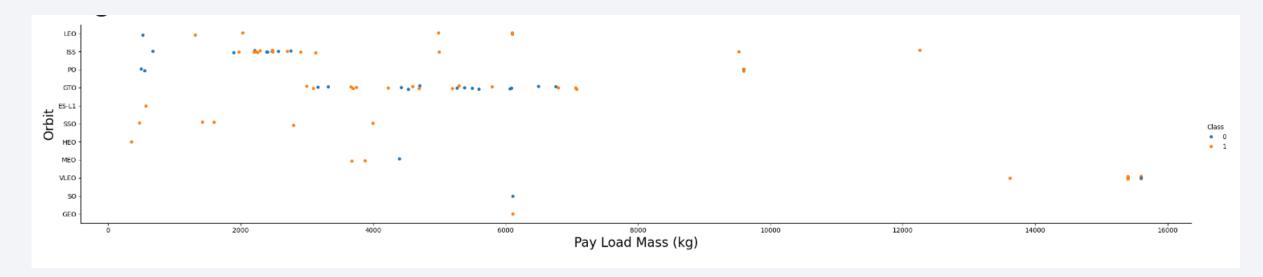
# Flight Number vs. Orbit Type

- Show a scatter point of Flight number vs. Orbit type
- Higher flight numbers are more successful than lower ones.
- Large variation of success rate for lower flight numbers



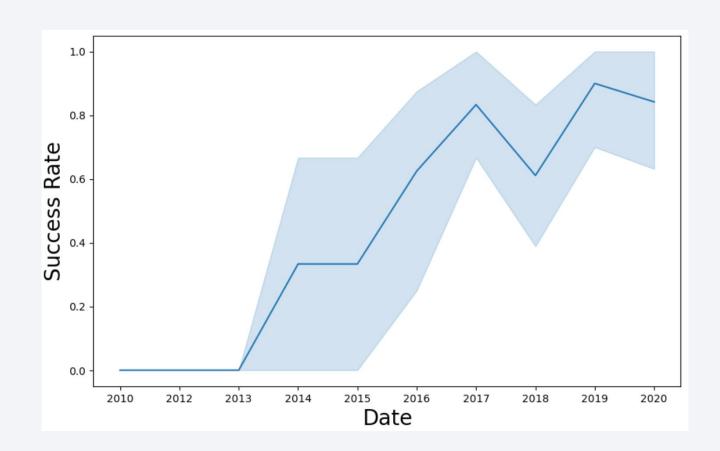
# Payload vs. Orbit Type

- Scatter point of payload vs. orbit type
- Larger payloads are sent into orbit VLEO and PO
- Larger payloads are more successful especially above 800kg



# Launch Success Yearly Trend

- Line chart of yearly average success rate
- Success rate increases year after year
- Larger variation in success rate in early years but becomes less in later years
- A dip of 20% percent seen from 2017 to 2019



#### All Launch Site Names

Used DISTINCT command

```
%sql SELECT DISTINCT "Launch_Site" FROM SPACEXTABLE
 * sqlite:///my_data1.db
Done.
   Launch_Site
  CCAFS LC-40
  VAFB SLC-4E
   KSC LC-39A
 CCAFS SLC-40
```

# Launch Site Names Begin with 'CCA'

#### Used LIKE command

%sql SELECT * FROM SPACEXTABLE WHERE "Launch_Site" LIKE 'CCA%' LIMIT 5									
* sqlite:///my_data1.db Done.									
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**

Used SUM and WHERE command

```
* sqlite://my_data1.db
Done.

* SUM(PAYLOAD_MASS_KG_)

45596
* FROM SPACEXTABLE WHERE Customer = 'NASA (CRS)'
```

# Average Payload Mass by F9 v1.1

Used AVG and WHERE command

# First Successful Ground Landing Date

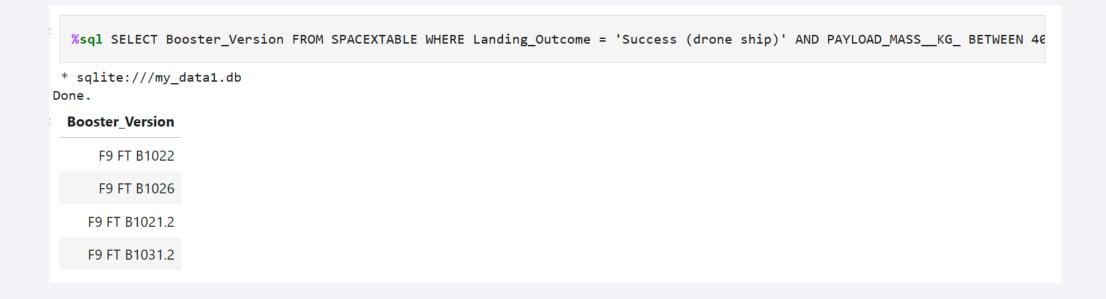
Used MIN and WHERE command

```
* *sqlite://my_data1.db
Done.
# MIN(DATE)

2015-12-22
```

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

Used WHERE and BETWEEN command

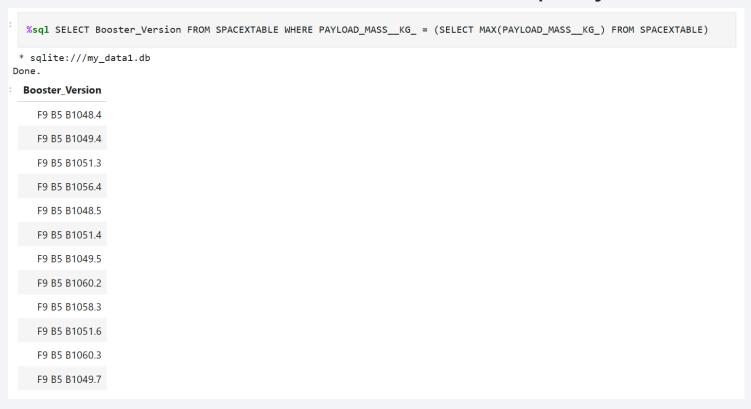


#### Total Number of Successful and Failure Mission Outcomes

Used COUNT and sub-querry

# **Boosters Carried Maximum Payload**

Used SELECT and MAX command with sub-querry



#### 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

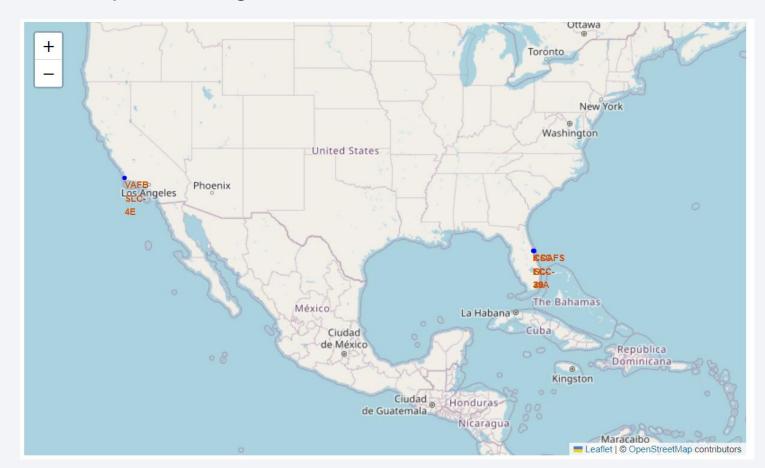
 Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order

Present your query result with a short explanation here



#### **Launch Site Locations**

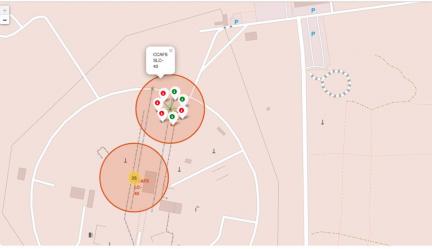
Launch Site of Space X Flights



#### Success/failed launches for each site on the map

• In depth of Launch Site with Markers indicating successful and failed flights





#### Calculate the distances between a launch site to its proximities

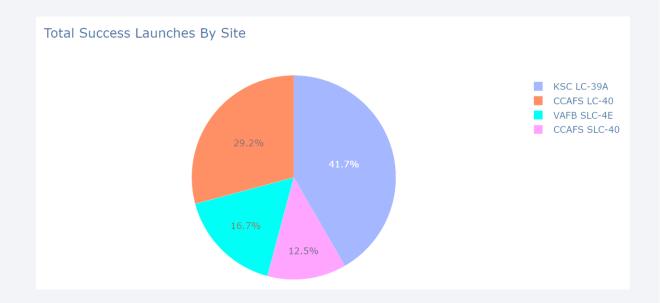
 Interactive map to calculate the distances between a launch site to its proximities





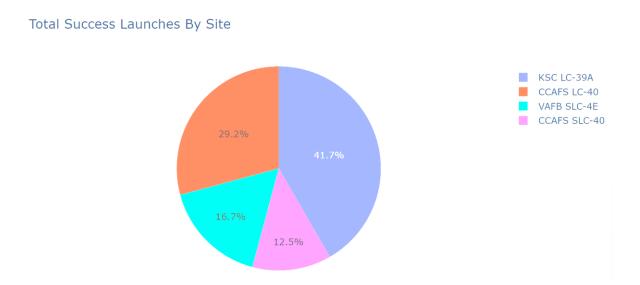
#### Success count for Launch sites

- KSC LC-39A has the highest count of successful launches
- CCAFS SLC-40 has the lowest



#### Success percentage for Launch sites

- KSC LC-39A has the highest percentage of successful launches
- CCAFS SLC-40 has the lowest



# Payload mass vs Booster Version category

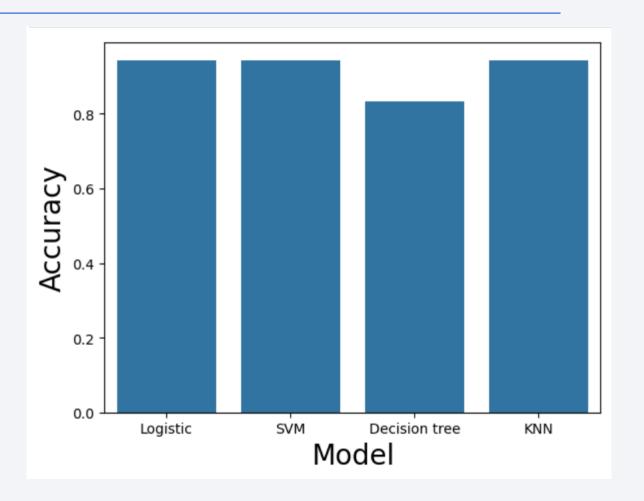
- Booster version category FT is the most successful for lower payloads but less successful for higher payload masses
- Booster version category v1.0 is the least successful





# Classification Accuracy

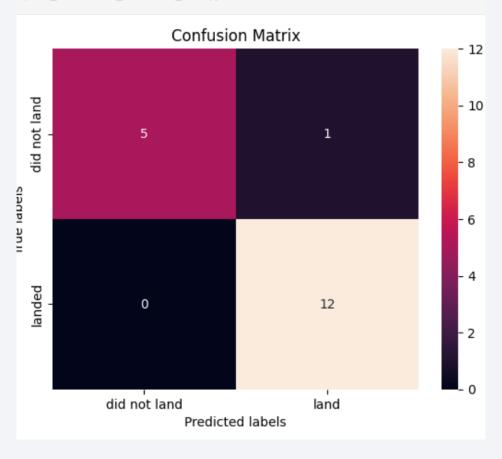
 Logistic, SVM and KNN model has the highest classification accuracy



#### **Confusion Matrix**

- Logistic the best performing model as it has an accuracy of 0.9444
- It is a simpler model

yhat=logreg\_cv.predict(X\_test)
plot\_confusion\_matrix(Y\_test,yhat)



#### **Conclusions**

- Flights gradually increased their success rate with time
- Higher payload mass seems more successful
- ES-L1, GEO, HEO and SSO orbits have highest success rates
- VLEO orbits have highest Payload mass
- Success rate of launches increases year after year
- Booster version category FT is the most successful for lower payloads but less successful for higher payload masses
- KSC LC-39A has the highest percentage of successful launches
- Logistic the best performing model as it has an accuracy of 0.9444

