

# Accessing the Success of Falcon 9's First-Stage Landing in Space X Rocket launches

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**IBM CAPSTONE PROJECT** 

# OUTLINE

- 1. EXECUTIVE SUMMARY
- 2. INTRODUCTION
- 3. METHODOLOGY
- 4. RESULTS
- 5. DISCUSSION
- 6. CONCLUSION

# **Executive Summary**

This report presents the result of a data science project that aimed to gain insights into Space X launches data, to build, evaluate and refine predictive models for discovering more insights.

The analysis was conducted on a dataset of Space X launches spanning over the years from 2010 to 2021



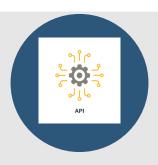
#### INTRODUCTION

SpaceX advertises Falcon 9 rocket launches on its website with a cost of 62 million dollars, while other providers cost upward of 165 million dollars each, much of its savings is because SpaceX can reuse the first stage. So, if we can determine if the first stage will land, we can determine the cost of a launch.

What we want to know is, what are the factors that affect the most how successful the first stage landing will be and with this have the capability to determine the cost of launch



#### **METHODOLOGY**



#### DATA COLLECTION SPACEX REST API

Using the public API that SpaceX offers, data was obtained by performing the following steps:

- Perform a request to the API
- Parse the response data to extract relevant information
- Store the data in a dataframe

Source: API Data Collection

- Filter the data for the Falcon 9 launches
- Replace null values

# DATA COLLECTION WEB SCARPING



Wikipedia contains information for SpaceX launches which can be obtained by performing the following:

- Make a request to Wikipedia
- Parse the HTML content using beautiful soup
- Save the extracted data and load into a dataframe

Source: Webscrapping Data Collection



#### **DATA WRANGLING**

The following steps were taken to clean, transform and prepare the data for analysis.

- Data Cleaning: The data was cleaned by looking for and handling missing values
- Data Transformation: The data was transformed by converting categorical variable to numerical.
   Then a landing outcome is created from the outcome column.

In this project, we aimed to analyze the factors that affect the landing outcome of the SpaceX falcon 9's first-stage. As part of the methodology, we used data visualization techniques such as **scatter/cat plots**, **line charts** and **bar chart** to explore the data and gained insights into the relationships between:

- Payload mass and flight number
- Launch site and flight number
- Orbit and flight number
- Launch site and Payload mass
- Payload and Orbit

#### **DATA VISUALIZATION**



Source: Data Wrangling

Source: Visualization



I also analyzed the dataset using SQL to extract and transform the data, in order to gain insights.

Analysis was performed using SQL queries to get the following information:

- Name of the unique launch sites in the space mission
- Five(5) records where launch sites begin with 'CCA'
- Total payload mass carried by booster launched by NASA (CRS)
- Average payload mass carried by booster version F9 v1.1
- Dates when the first successful landing outcome in group pad was achieved.

- Names of the boosters which have success in drone ship and have payload mass between 4000 and 6000
- Total number of successful and failure mission outcomes
- Names of the booster versions which have carried the maximum payload mass
- Records which will display month names, failure landing outcomes in drone ship, booster versions, launch site for the months in year 2015
- Count of successful landing outcomes between the date 04-06-2010 and 20-03-2017

EDA: SQL (cont'd)

Source: Analysis with SQL

# INTERACTIVE MAP WITH FOLIUM

To visualize the spatial distribution of the data. An interactive map was created using the python library Folium. The steps taken in creating these maps are as follows:

- Import the required libraries including Folium and pandas
- Loading the data into a pandas dataframe
- Creating a map object using folium.map()
- Adding markers or circles to the map
- Adding additional layer.

Source: Folium Map

To provide an interactive and customizable interface for exploring the data, a dashboard was created using Plotly Dash library. The steps involved are as follow:

- Importing the necessary libraries including pandas, dash, dash components (html and core, dash dependencies
- Defining the layout and structure of the dashboard such as the dropdown menus, sliders and graphs
- Defining the callbacks for each component of the app
- Finally, the app was ran using app.run\_server()

#### **Dashboard with Plotly Dash**

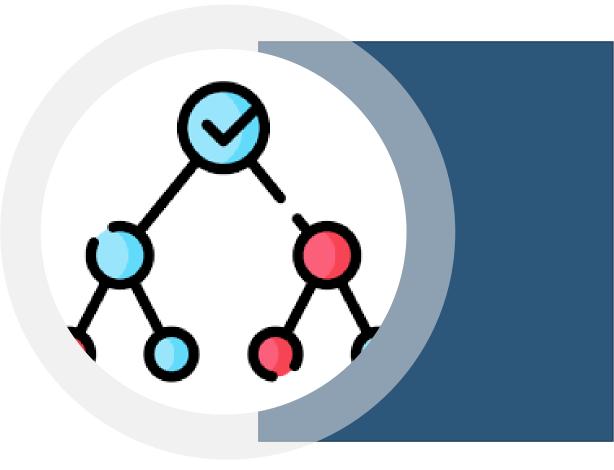
Source: Dashboard

# Predictive Analysis: Classification

Predictive analysis was performed on the dataset using machine learning techniques. Four classification models were used;
Logistic Regression, Support Vector Machine, Decision Tree and K Nearest Neighbors.

With the following steps, predictive analysis was performed with these models:

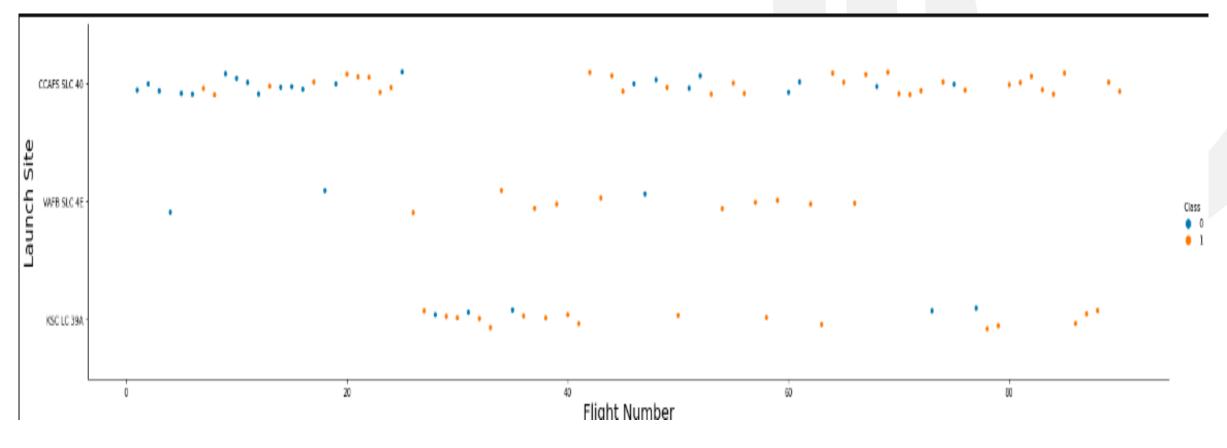
- Data preparation and standardization
- Split data into training and test sets
- Evaluate the models using GridSearch
- Obtain accuracy of the model and compared the result



Source: <u>Predictive Analysis</u>

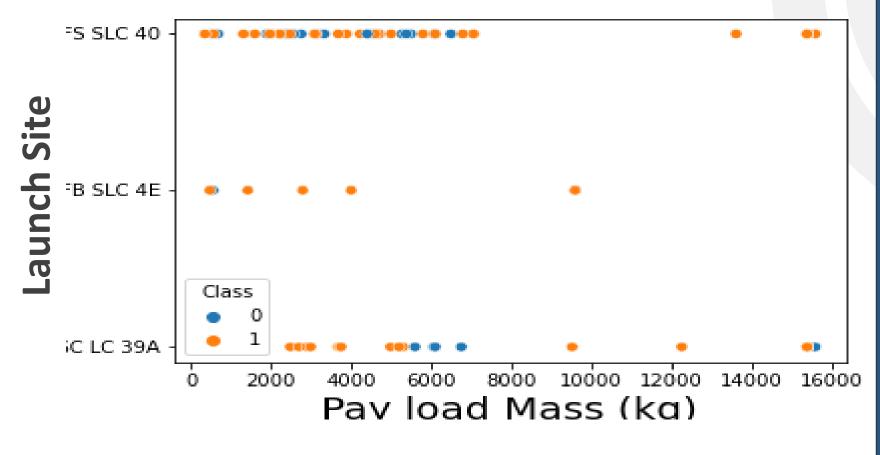
# **RESULTS**

- > Exploratory Data Analysis results
  - > Visualization
  - > SQL
  - Folium Map
  - Dashboard
- Predictive Analysis results
  - > Classification



#### **INSIGHTS**

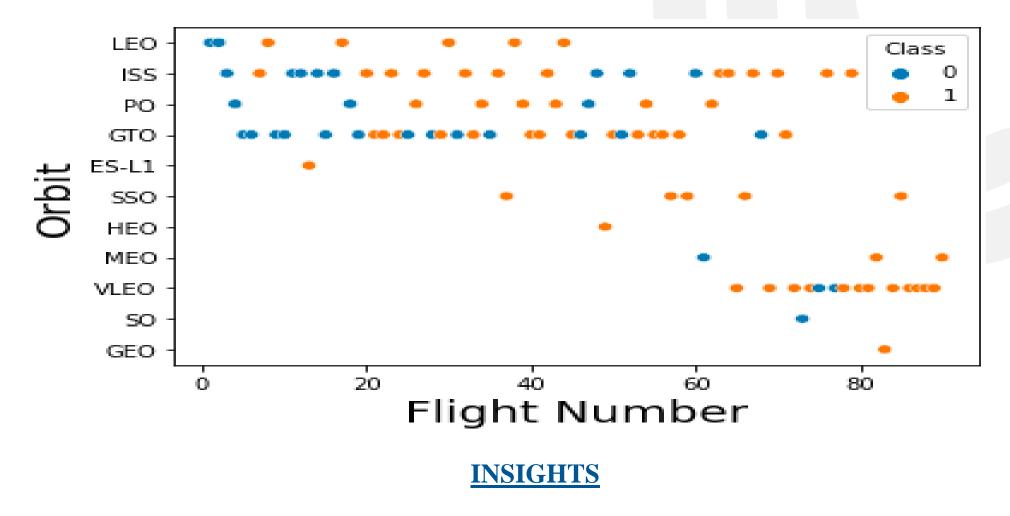
- The success rate increased over time in each launch-site
- The launch-site with the highest number of falcon 9 launches is CCAFS LC-40
- The launch site with the highest success rate is KSC LC- 39A



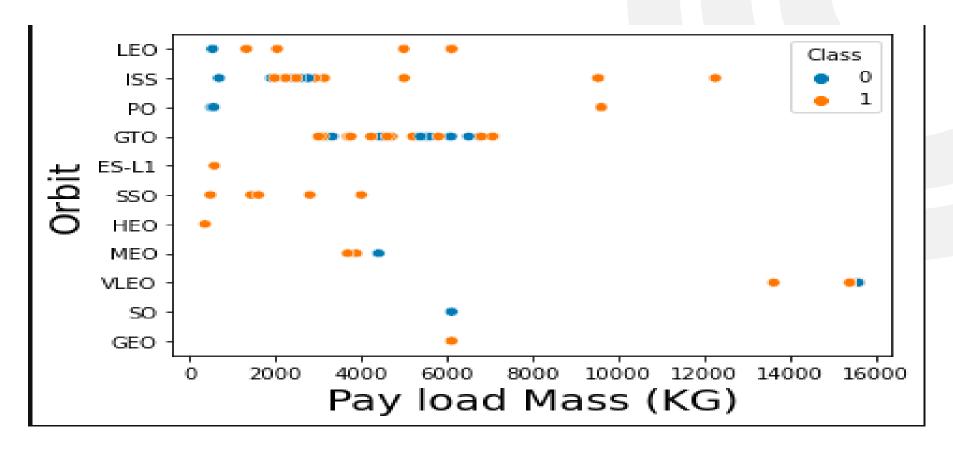
- Failure
- Success

# **Insights**

- Payloads of over 8000 kg have a high success rate
- VAFB-SLC launch-site there are no rockets launched for heavy payload mass(greater than 10000).
- KSC LC-39A has a high success rate for payloads under 6000 kg



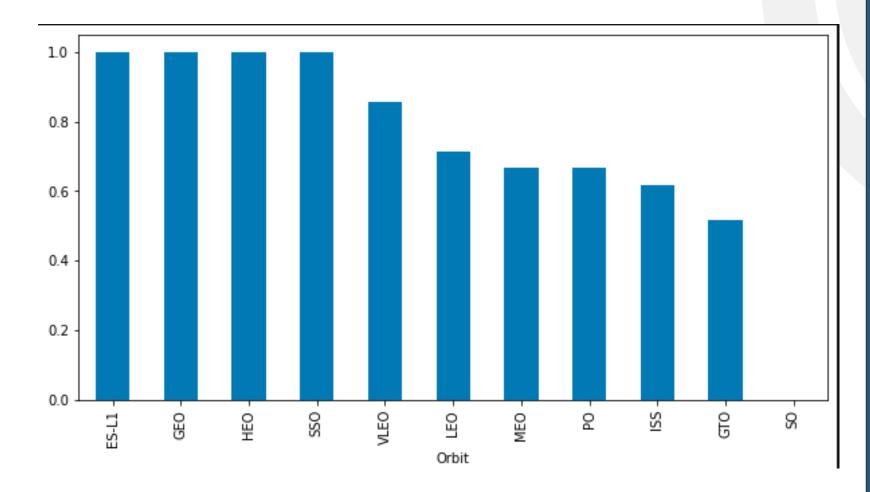
- In the LEO orbit the Success appears related to the number of flights
- There seems to be no relationship between flight number when in GTO orbit.



#### **INSIGHTS**

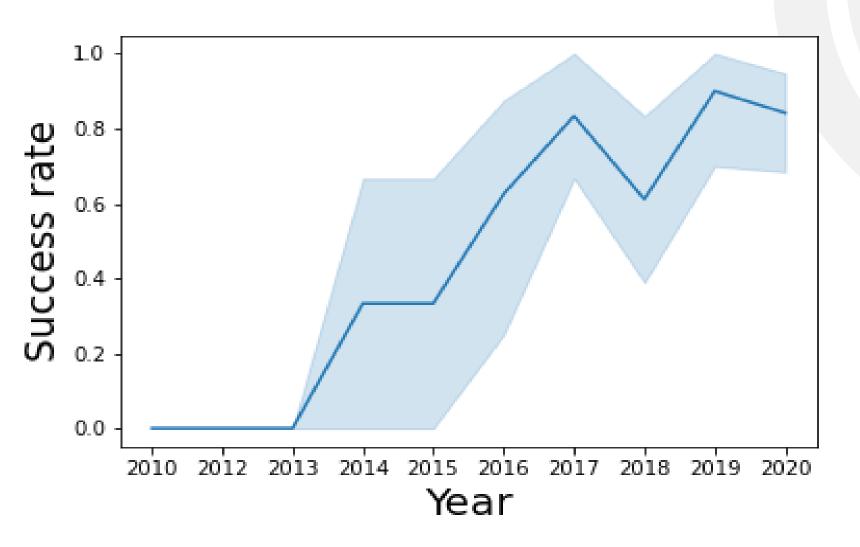
- With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS
- In GTO we cannot distinguish well between the successful landing and unsuccessful ones.

# **Visualization**



#### **INSIGHTS**

- ES-L1, GEO, HEO and SSO have a 100% success rate
- SO has 0% success rate

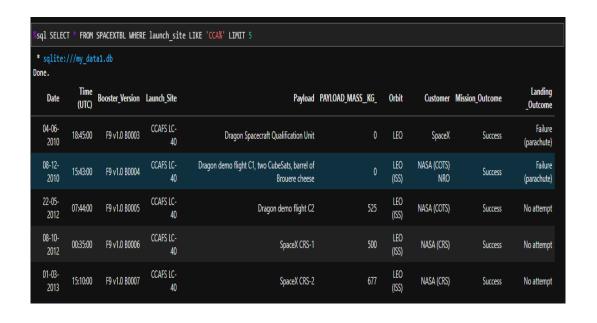


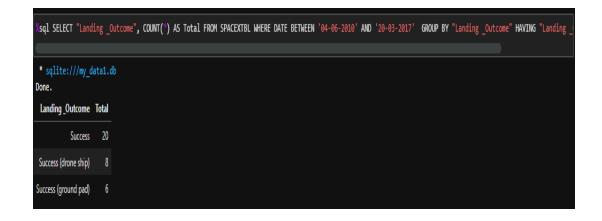
# **Insights**

- Success rate kept on increasing since 2013 to 2020
- There was a dip in the success rate in the year 2018

# **SQL**

# 5 Records where Launch Sites begin with the string 'CCA'



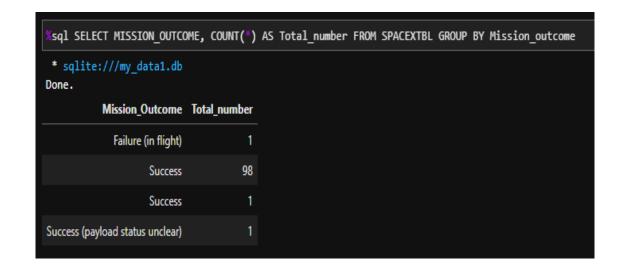


Ranking the count of successful landing outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

# **SQL**

# Names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

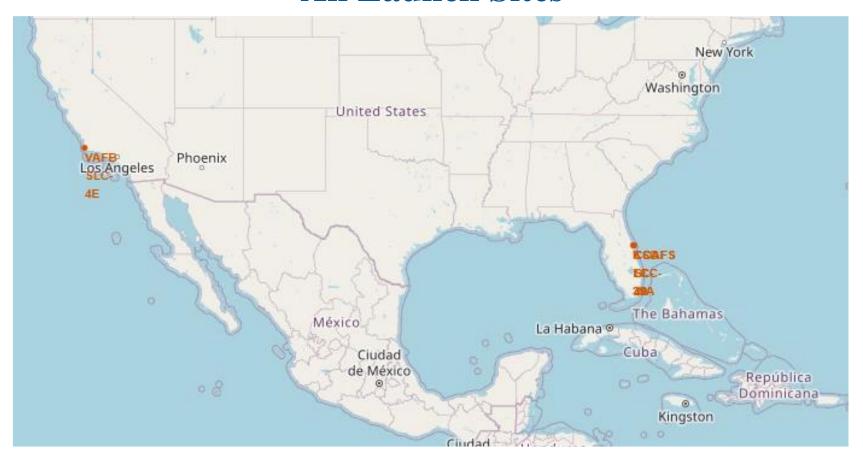
%sql SELECT DI	ISTINCT(booster_	version)	FROM SPACEXTBL	WHERE	"Landing _	_Outcome" =	(drone shi	ip)" AND	PAYLOAD_	MASS_KG	BETWEEN	4000 AND	6000
* sqlite:///m Done.	ny_data1.db												
Booster_Version													
F9 FT B1022													
F9 FT B1026													
F9 FT B1021.2													
F9 FT B1031.2													



Total number of successful and failure mission outcomes

# **FOLIUM MAP**

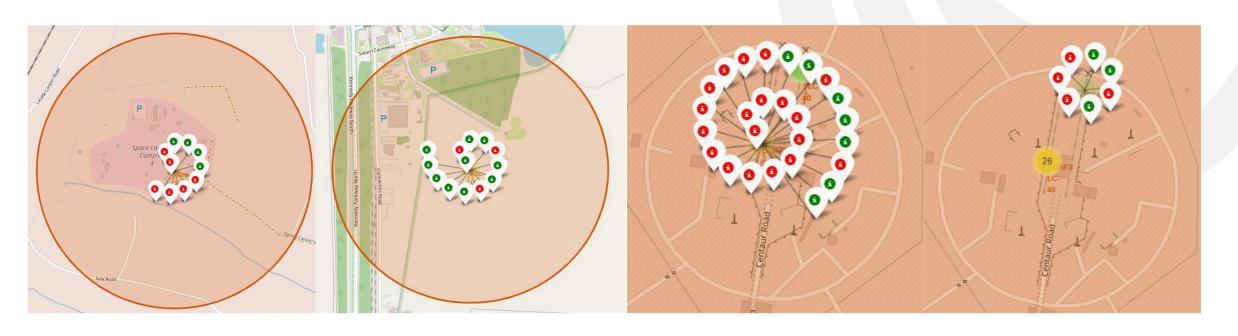
#### **All Launch Sites**



# **Insights**

The map shows the location of each launch site. It can be seen from the map that the launch sites are located near the sea to avoid populated areas

#### **Launch Outcome in each Launch Site**



**VAFB SLC-4E** 

**KSC LC-39** 

CCAFS SLC-40

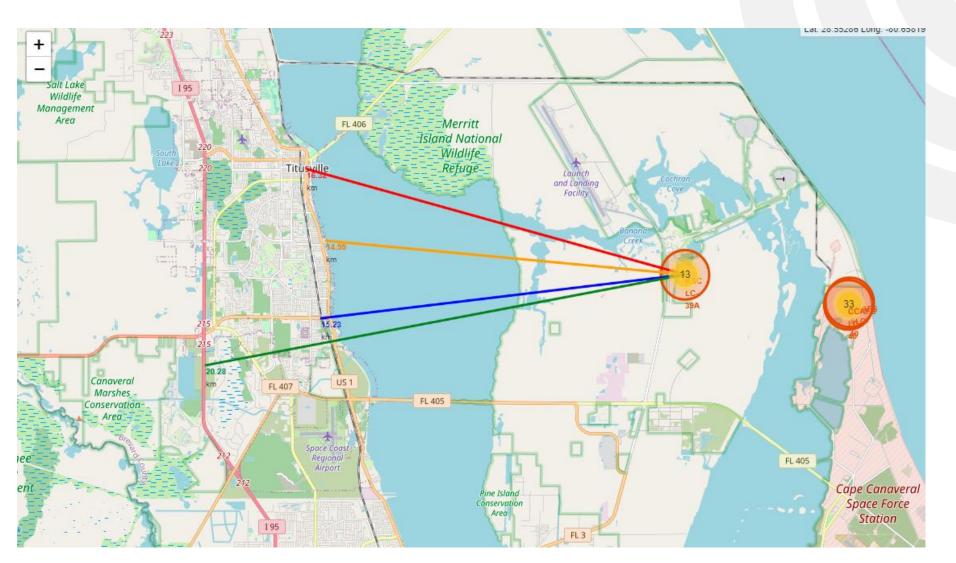
**CCAFS LC-40** 

Failed Launches





# Distance Between KSC LC-39 To The Nearest Highway, Railway, City And Coastline



Railway 15.23km

**City 16.32km** 

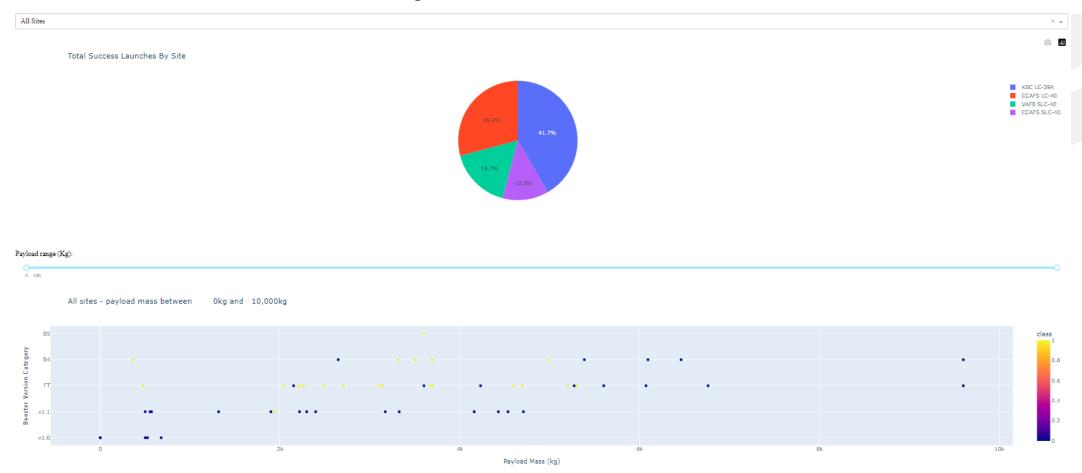
Highway 20.28km

Coastline 14.99km

# **DASHBOARD**

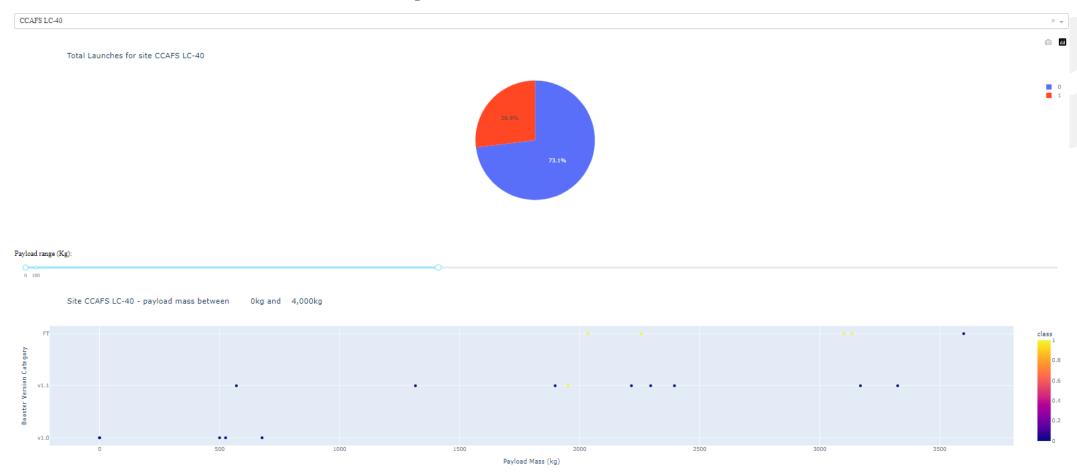
#### Successful Launches by Sites with Payload Mass Between 0-10000kg

#### SpaceX Launch Records Dashboard



#### Successful Launches in CCAFS LC-40 with Payload Mass between 0-4000 kg

#### SpaceX Launch Records Dashboard



# PREDICTIVE ANALYSIS

#### **CLASSIFICATION**

Comparing the different ML models with evaluation metrics

Whole Dataset Test Data

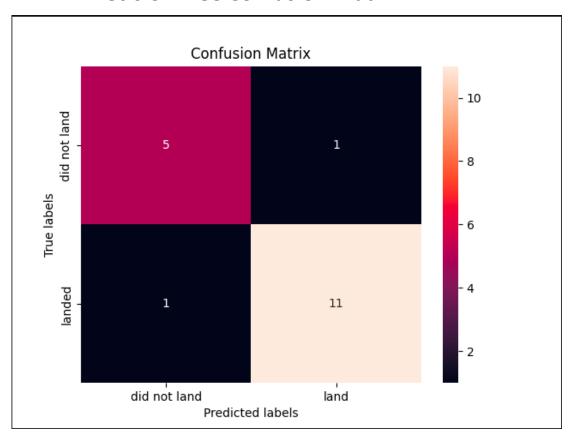
[33]:		LogReg	SVM	Tree	KNN
	Jaccard_Score	0.833333	0.845070	0.841270	0.819444
	F1_Score	0.909091	0.916031	0.913793	0.900763
	Accuracy	0.866667	0.877778	0.888889	0.855556

Jaccard_Score         0.800000         0.800000         0.846154         0.800000           F1_Score         0.888889         0.888889         0.916667         0.888889           Accuracy         0.833333         0.833333         0.888889         0.833333	[32]:		LogReg	SVM	Tree	KNN
		Jaccard_Score	0.800000	0.800000	0.846154	0.800000
Accuracy 0.833333 0.833333 0.888889 0.833333		F1_Score	0.888889	0.888889	0.916667	0.888889
		Accuracy	0.833333	0.833333	0.888889	0.833333

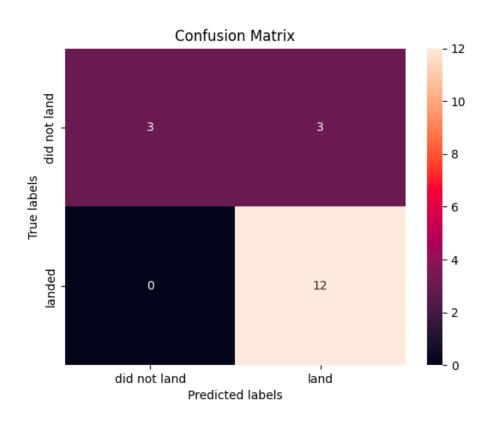
☐ From the above image, Decision Tree shows to be the best classification model to use while Scoring on the Test dataset and the whole dataset

#### **Confusion Matrix**

#### **Decision Tree Confusion Matrix**



#### **Logistic Regression Confusion Matrix**



This model correctly predicted 91.67% of the true positive cases and 83.3% of the true negative cases

This model correctly predicted 100% of the true positive cases and 50% of the true negative cases

# **CONCLUSION**

- The success rate increased over time in each launch-site
- The launch site with the highest success rate is KSC LC- 39A
- ES-L1, GEO, HEO and SSO have a 100% success rate
- Decision tree is the most optimal classification model in this case.
- the launch sites are located near the sea to avoid populated areas





# THANK YOU!



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https://github.com/darexboy/