# IBM Data Science Capstone Project

Analysis on Space X Falcon 9

#### **Table of contents**

01

#### **Executive Summary**

You can describe the topic of the section here

04

#### Results

You can describe the topic of the section here

02

#### Introduction

You can describe the topic of the section here

05

#### Conclusion

You can describe the topic of the section here

03

#### Methodology

You can describe the topic of the section here

06

#### **Appendix**

You can describe the topic of the section here

# 01

**Executive Summary** 

#### **Executive Summary**

This report mainly focuses on spacex falcon 9. Particularly, collected the data using API and webscraping, conducted data wrangling, exploratory data analysis using pandas and SQL, created visualizations and dashboard using plotly and dash, and created a model to predict if the falcon 9 will land successfully.



# 02 Introduction

#### Introduction

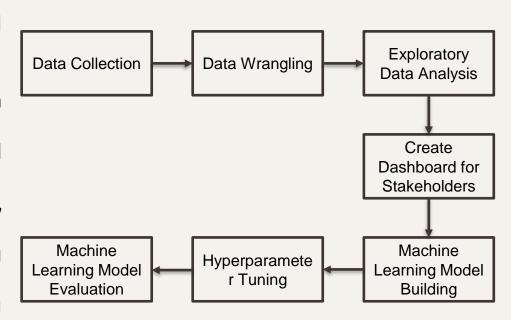
The cost of launching a rocket to space requires millions of dollars but aiming to lowert the cost would highly benefit the company. Providers cost more than 165 million dollars just to launch a rocket successfully, but spacex lower their cost by around 60%; the reason behind is the spacex can reuse the first stage of the process. Therefore, if we can determine if the first stage will land, the management would be able to approximately tell the cost of a launch.

This is the reason why we need to predict the successful launch using various machine learning algorithm. But before that, analysis, visualizations, and evaluations are needed to conclude.

# 03 Methodology

#### Methodology

- Collected the data using API and Webscraping;
- 2. Wrangled data using Pandas;
- 3. Conducted Exploratory Data Analysis using Pandas and SQL;
- 4. Created Dashboard using Dash and Plotly;
- 5. Used Logistic Regression, SVM, Decision Tree, and KNN;
- 6. Tuned hyperparameters using GridSearch CV;
- 7. And Evaluated Machine Learning Model using Testing set and CV Scores



# 04 Results

 The spacex launch site are called, CCAFS LC -40, VAFB SLC-4E, KSC LC-39A, and CCAFS SLC-40

```
[7]:  %%sql

SELECT DISTINCT Launch_Site FROM SPACEXTBL LIMIT 5

* sqlite:///my_data1.db
Done.

[7]: Launch_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40
```

 The first 5 observations are all CCAFS LC-40

```
%%sql
SELECT launch_site FROM spacextbl WHERE launch_site like "CCA%" LIMIT 5
 * sqlite:///my_data1.db
Done.
Launch Site
CCAFS LC-40
CCAFS LC-40
CCAFS LC-40
CCAFS LC-40
CCAFS LC-40
```

 The total number of payload is 619967(KG)

The average payload is
 2628.4 for F9 v1.1 Booster

```
%%sql
SELECT SUM(PAYLOAD_MASS__KG_) FROM spacextbl

* sqlite:///my_data1.db
Done.
SUM(PAYLOAD_MASS__KG_)
619967
```

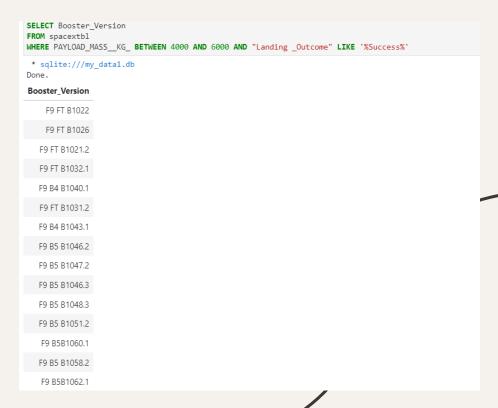
 The very first successful landing on ground pad recorded was on january 5, 2017

```
%%sql
SELECT min(date) FROM spacextbl WHERE "Landing _Outcome" LIKE '%Success%'

* sqlite:///my_data1.db
Done.
min(date)

01-05-2017
```

 List of successful launch booster version between 4000-6000 payload mass in KG



 There are 61 number of successful landing, while there are only 10 unsuccessful. The remaining are, controlled, prelude, and no attempt

 The two (2) failure drone ship occurred in January and April.

```
%%sal
SELECT CASE
     WHEN substr(Date,4,2) = "01" THEN 'January'
     WHEN substr(Date,4,2) = "02" THEN 'February'
     WHEN substr(Date,4,2) = "03" THEN 'March'
     WHEN substr(Date,4,2) = "04" THEN 'April'
     WHEN substr(Date,4,2) = "05" THEN 'May'
      WHEN substr(Date,4,2) = "06" THEN 'June'
     WHEN substr(Date,4,2) = "07" THEN 'July'
     WHEN substr(Date,4,2) = "08" THEN 'August'
     WHEN substr(Date,4,2) = "09" THEN 'September'
     WHEN substr(Date,4,2) = "10" THEN 'October'
     WHEN substr(Date,4,2) = "11" THEN 'November'
     WHEN substr(Date,4,2) = "12" THEN 'December'
     END AS Month, "Landing Outcome", Booster Version
FROM spacextbl
WHERE "Landing Outcome" = "Failure (drone ship)" AND substr(Date,7,4)='2015'
 * sqlite:///my data1.db
Done.
Month Landing Outcome Booster Version
January Failure (drone ship)
                            F9 v1.1 B1012
  April Failure (drone ship)
                           F9 v1.1 B1015
```

 Between April 06, 2010, and March 20, 2017, there are (5) successful drone ship, and three (3) successful ground pad

```
Rank the count of successful landing_outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

**Saql

SELECT "Landing _Outcome", COUNT("Landing _Outcome") FROM spacextbl

WHERE substr(date,7)||substr(date,4,2)||substr(date,1,2)

BETWEEN '20100604' AND '20170320' AND "Landing _Outcome" LIKE "%Success%"

GROUP BY "Landing _Outcome"

ORDER BY substr(date,7) DESC

* sqlite:///my_data1.db

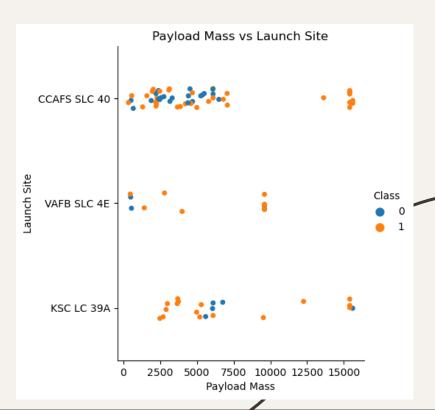
Done.

**Landing _Outcome COUNT("Landing _Outcome")

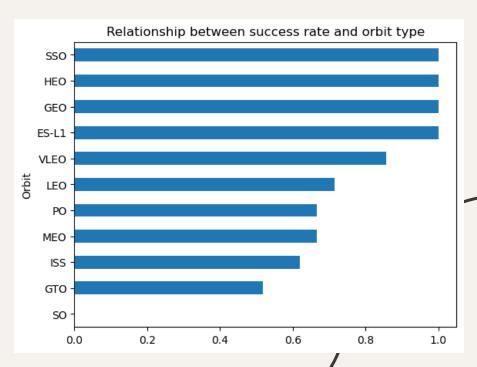
Success (drone ship) 5

Success (ground pad) 3
```

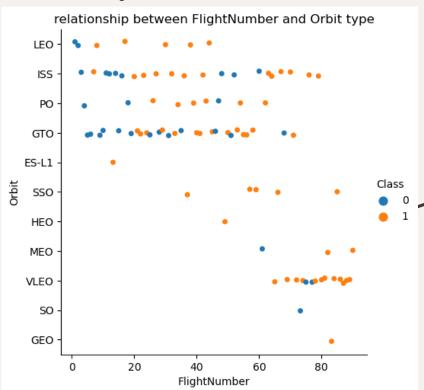
- There are no rocket launch in "VAFB SLC 4E" for heavy payload more than 10,000
- The lower the payload mass, there's a high probability that a launch won't succeed
- There are more successful launch in VAFB SLC 4E, and KSC LC 39 A.
- There are more attempt in CCAFS SLC 40



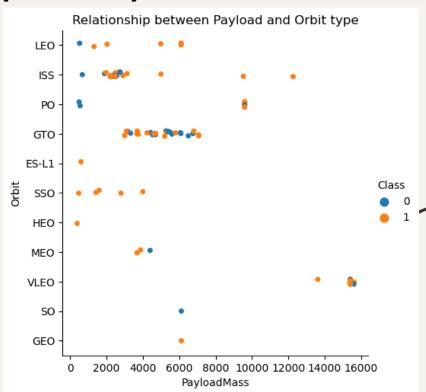
- SSO, HEO, GEO, and ES-L1, have a 100% launch success rate.
- Success rate of GTO is only approximately 50%
- LEO, PO, MEO, ISS, and VLEO's success rate are ranging from 60% to 90%



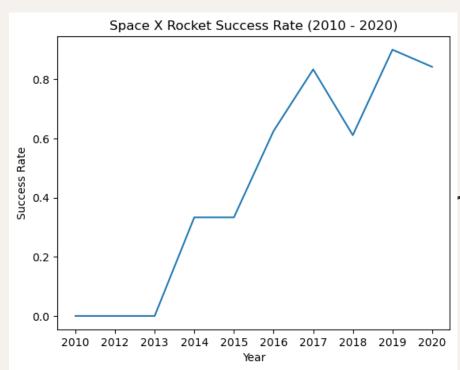
- LEO is related to the number of flights; as the number of flights increases LEO's success rate also increases
- However, this lack information to conclude that the orbit has relationship to the number of flights



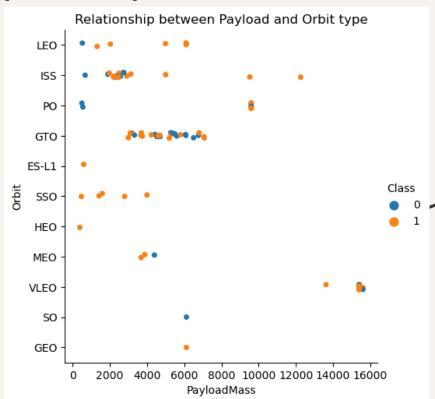
- With heavy payloads the successful landing rate are more for LEO, and ISS.
- In GTO, the amout of payload is not a factor to determine a successful landing.



 After 2013, the successrate of SpaceX kept on increasing.



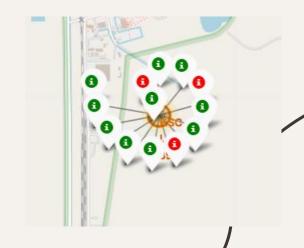
- With heavy payloads the successful landing rate are more for LEO, and ISS.
- In GTO, the amout of payload is not a factor to determine a successful landing.



#### Results on the Analysis (Geographical)

 The KSC LC-39A Has 3 unsuccessful landing but have 10 successful landings.

#### **KSC LC-39A Launch Site**



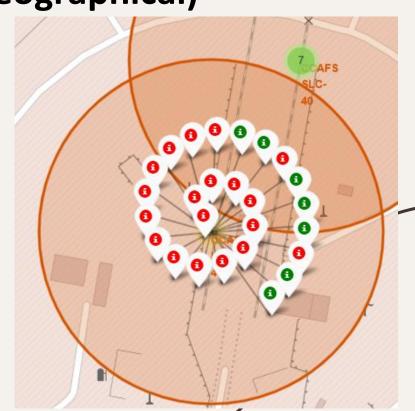
### Results on the Analysis (Geographical)

 The CCAFS SLC – 40 has 4 unsuccessful landings and 3 successful landings

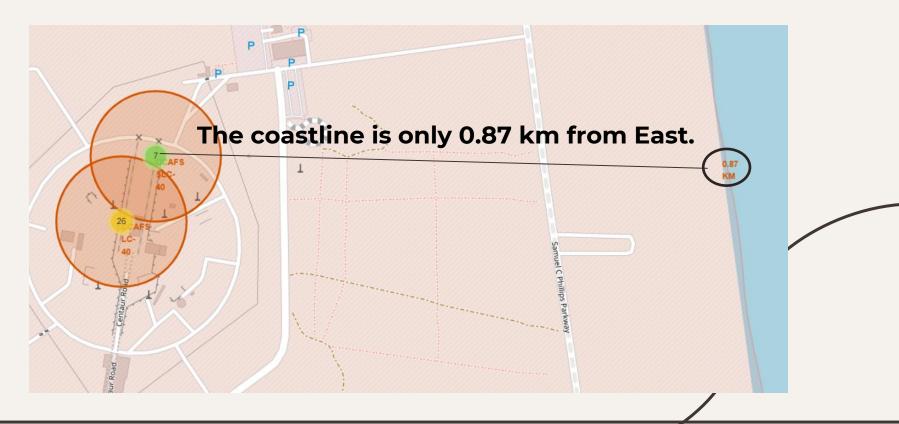


### Results on the Analysis (Geographical)

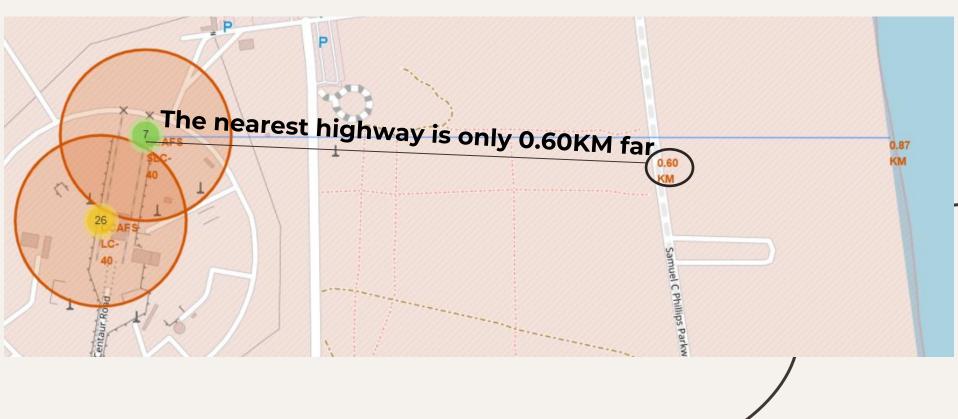
 The CCAFS LC – 40 has the majority of unsuccessful landings. With only seven (7) successful landings out of 26 attempts.



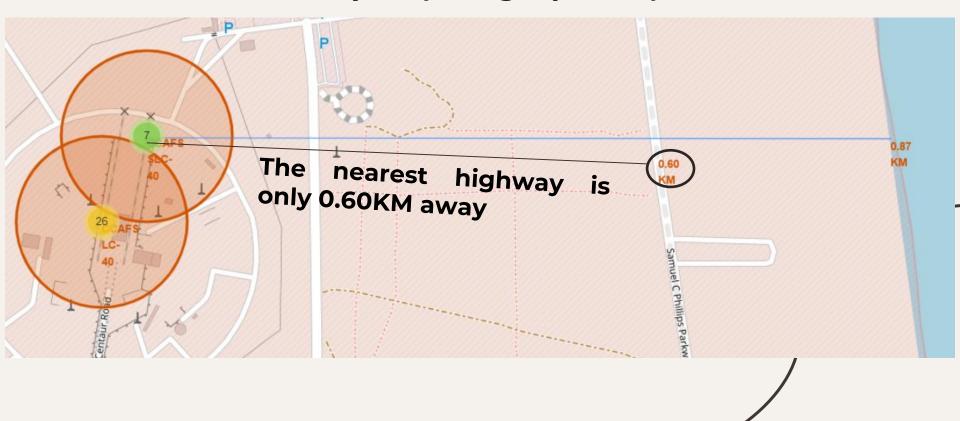
# Results on the Analysis (Geographical) - CCAFS SLC-40



# Results on the Analysis (Geographical) - CCAFS SLC-40



### Results on the Analysis (Geographical) - CCAFS SLC-40



### Results on the Analysis (Dashboard)

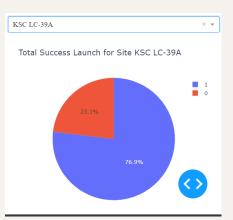
 In all sites, the sites with the highest successful landing rate was the KSC LC-39A

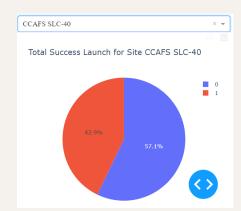
# SpaceX Launch Records Dashboard



# Results on the Analysis (Dashboard) Total Succ

 CCAFS SLC-40 has the lowest landing success rate. While the other three have over 60% success rates



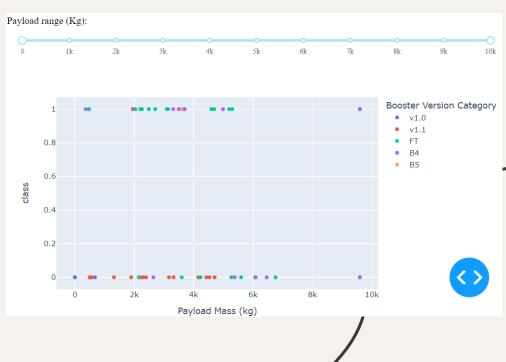






# Results on the Analysis (Dashboard)

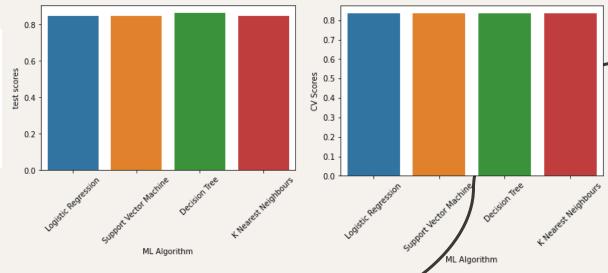
- There are more attempt using low payloads
- it also shows that the success for massive payloads is lower than that for low payloads.
- Lastly, v 1.0, and B5 are not yet launched with massive payloads



#### **Results on Model Evaluation**

• The tables below shown are scores from testing and cross validation using GridSearch CV. Specifically, lookingat the CV Scores, all accuracies are the same but in testing, the decision tree yield the highest percentage (86.25% Accuracy)

	ML Algorithm	CV Scores	test scores
0	Logistic Regression	0.833333	0.846429
1	Support Vector Machine	0.833333	0.848214
2	Decision Tree	0.833333	0.862500
3	K Nearest Neighbours	0.833333	0.848214



### **Results on Model Evaluation (Cont.)**

- The best hyperparameter of the decision tree are the following:
  - o Criterion gini
  - max\_depth 6
  - Max\_features auto
  - o Min\_samples\_leaf 4
  - o min\_samples\_split 10
  - Splitter random

This hyperparameter yield a 86.25 % accuracy for our testing data

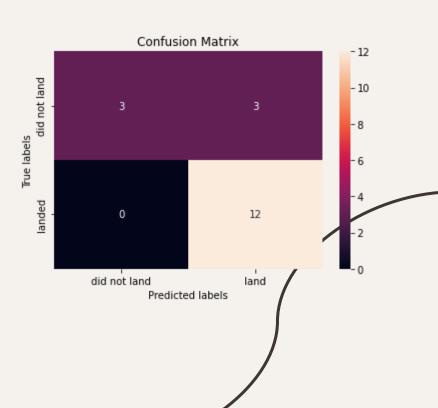
# 05 Conclusions

#### **Conclusion**

 Even if the decision tree has the highest accuracy on the testing or the evaluation. All confusion matrix of every model are just the same. Therefore, more data is needed to further analyze the machine learning model and to choose what is the best machine learning model for this scenario

#### Conclusion

- Moreover, there are 15 correct predictions (True Positive and True Negative) and there are 3 wrong predictions (False Negatives).
- This tell us that using 18 observations, our model correctly classified 15 observations.



# 06 Appendix

# Appendix

- Machine Learning Model Link: <a href="https://jp-tok.dataplatform.cloud.ibm.com/analytics/notebooks/v2/2cf6ae/47-f6c2-4f44-alfd-5lel3f0dabel/view?access\_token=62b443lf8l7l48ceaab0782l5a/2d44ef90d669d80cc55la32b0l302a74l0ed50</a>
- Github Link: https://github.com/Justinjay282/IBM-Data-Science-Certificate-Capstone-Project