

Space Race with Data Science

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OUTLINE

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- Methodology
- Results
- Conclusion
- Appendix

EXECUTIVE SUMMARY

- For the capstone project, I predict if SpaceX Falcon 9 first stage is going to land successfully by applying different machine learning algorithms
- This project includes
 - Data collection, wrangling
 - Exploratory data analysis
 - Data visualizations
 - ML predictions
- The graphs provide some reason to believe that specific features of rocket launches are correlated with whether they succeed or fail
- The decision tree is potentially the best algorithm

INTRODUCTION

- For the capstone project, I predict if SpaceX Falcon 9 first stage is going to land successfully by applying different machine learning algorithms. SpaceX is capable of saving about 100 million dollars because they are able to reuse the first stage. Moreover, if I can decipher whether a rocket's first stage will land, I can help determine the cost of the launch. This would benefit rivaling companies of SpaceX.
- I'm looking to determine for a specific set of features in a rocket launch like the mass, orbit type, launch site, and others, will the first stage land?

Part 1

Methodology



METHODOLOGY

Executive Summary:

• Data collection methodology:

SpaceX API

WebScraping

Perform data wrangling:

Based on outcome data

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models:

Train and set tests -> classification algorithms -> make conclusion

Data Collection – SpaceX API

 SpaceX API o API provides different types of SpaceX rocket launches o Missing data replaced by mean o 90x17 of data

 https://github.com/agprays/Capstoneagprays/blob/main/1 jupyter-labs-spacex-data-collectionapi.ipynb

Data Collection - Scraping

Webscraping

https://github.com/agprays/Capstone-agprays/blob/main/2_jupyterlabs-webscraping.ipynb

Data Wrangling

• EDA -> summarizations -> Outcome Label

https://github.com/agprays/Capstone-agprays/blob/main/3 labs-jupyter-spacex-Data%20wrangling.ipynb

EDA with Data Visualization

• Use scatterplots and boxplots to show relationships https://github.com/agprays/Capstone-agprays/blob/main/5 edadataviz.ipynb

EDA with SQL

SQL queries o Launch site names

Total payload mass by NASA

Average payload mass by booster

https://github.com/agprays/Capstone-

agprays/blob/main/4 jupyter-labs-eda-sql-coursera sqllite.ipynb

Build an Interactive Map with Folium

- Used markers, circles, lines and marker clusters
- Shows successes, fails, distances, and locations

```
https://github.com/agprays/Capstone-
 agprays/blob/main/6 lab jupyter launch site location.ipynb
```

Build a Dashboard with Plotly Dash

- Lauches by site percentages
- Payload range
- Helps show the relationship between payload and launch site

Predictive Analysis (Classification)

Classification Model

Logistic regression

Support Vector Machine

Decision Tree

K-Nearest Neighbors

Data prep -> combinations -> conclusion

https://github.com/agprays/Capstoneagprays/blob/main/8 SpaceX Machine%20Learning%20Prediction Par t 5.ipynb

RESULTS

- Four launch sites
- Two booster versions
- Improvement over years

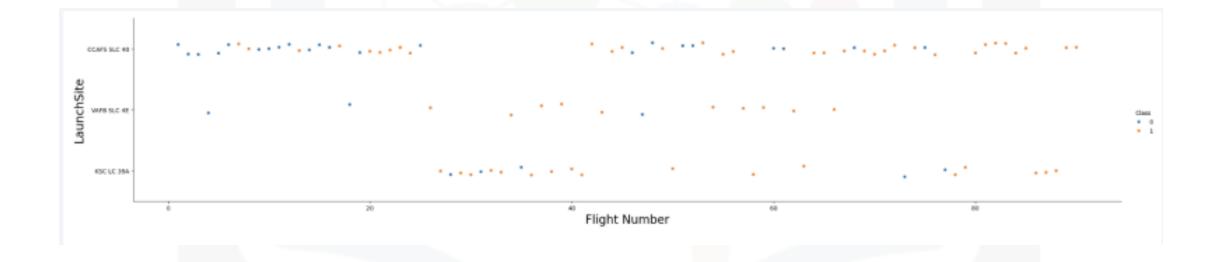
Part 2

Insights drawn fron EDA



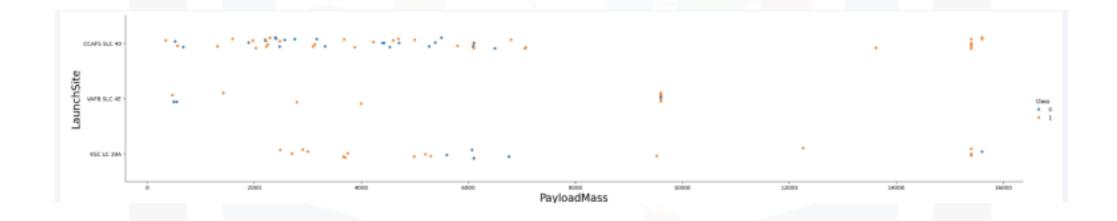
Flight Number vs. Launch Site

Flight Number vs. Launch Site o success over time



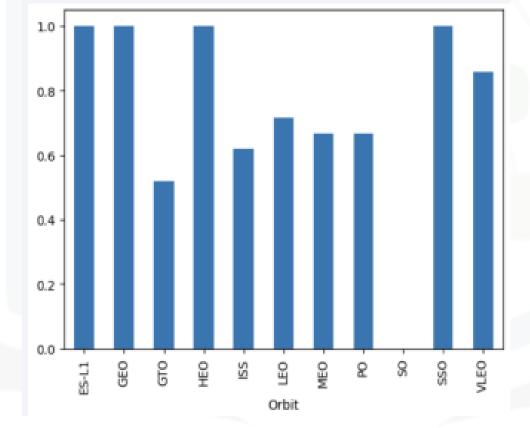
Payload vs. Launch Site

Payload vs. Launch Site o Payloads > 9,000kg have good success rate



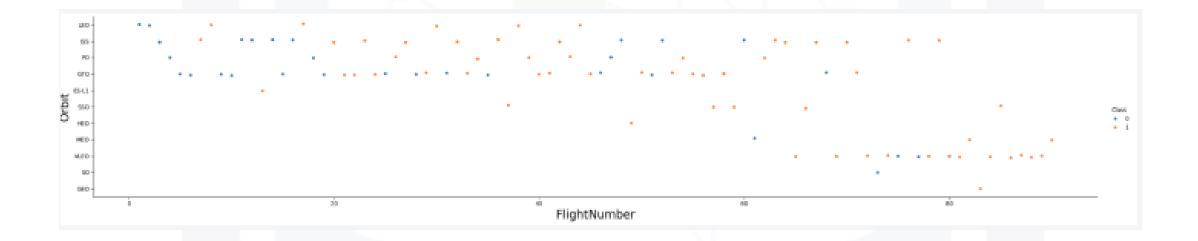
Success Rate vs. Orbit Type

ES-L1, GEO, HEO, SSO are all tied at the top



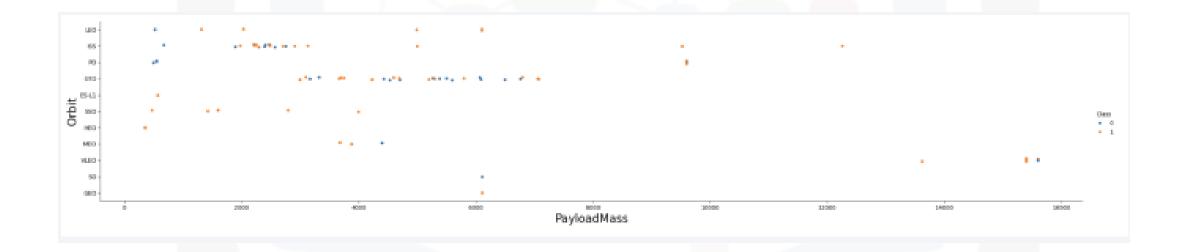
Flight Number vs. Orbit Type

Flight number vs. Orbit type o Success over time



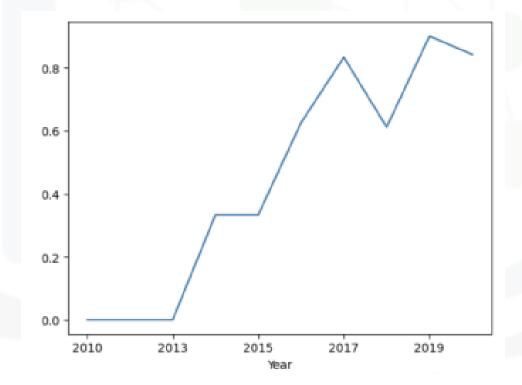
Payload vs. Orbit Type

payload vs. orbit type o No relation



Launch Success Yearly Tren

No improvement at 2010- 2013 but that increases a lot



All Launch Site Names

Four Launch sites

Launch_Site

CCAFS LC-40

CCAFS SLC-40

KSC LC-39A

VAFB SLC-4E

Launch Site Names Begin with 'CCA'

5 of these records

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASSK	G_	Orbit	Customer	Mission_Outcome	Landing_Outcon
2010- 06- 04	18:45:00	F9 v1.0 B0003	CCAFS LC- 40	Dragon Spacecraft Qualification Unit		0	LEO	SpaceX	Success	Failure (parachut
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 (parachut
2012- 05- 22	7:44:00	F9 v1.0 B0005	CCAFS LC- 40	Dragon demo flight C2	5	525	LEO (ISS)	NASA (COTS)	Success	No attem
2012- 10- 08	0:35:00	F9 v1.0 B0006	CCAFS LC- 40	SpaceX CRS-1	5	600	LEO (ISS)	NASA (CRS)	Success	No attem
2013- 03- 01	15:10:00	F9 v1.0 B0007	CCAFS LC- 40	SpaceX CRS-2	6	377	LEO (ISS)	NASA (CRS)	Success	No attem

Total Payload Mass

the total payload carried by boosters from NASA

TOTAL_PAYLOAD

111268

Average Payload Mass by F9 v1.1

The average payload mass carried by booster version F9 v1.1

AVG_PAYLOAD

2928.4

First Successful Ground Landing Date

first successful landing outcome on ground pad

List the date when the first succesful landing outcome in ground pad was acheived.

Hint:Use min function

sql SELECT MIN(DATE) AS FIRST_SUCCESS_GP FROM SPACEXTBL WHERE LANDING__OUTCOME = 'Success (ground pad)';

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

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

sql SELECT DISTINCT BOOSTER VERSION FROM SPACEXTBL WHERE PAYLOAD_MASS_KG_ BETWEEN 4000 AND 6000 AND LANDING_OUTCOME =



Total Number of Successful and Failure Mission Outcomes

Total number of successful and failure mission outcomes

	Mission_Outcome	QTY
	Failure (in flight)	1
	Success	98
	Success	1
Succ	ess (payload status unclear)	1

Boosters Carried Maximum Payload

Booster which have carried the maximum payload mass

F9 B5 B1048.4
F9 B5 B1048.5
F9 B5 B1049.4
F9 B5 B1049.5
F9 B5 B1049.7
F9 B5 B1051.3
F9 B5 B1051.4
F9 B5 B1051.6
F9 B5 B1056.4
F9 B5 B1060.2
F9 B5 B1060.3

2015 Launch Records

The failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015

> 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.

Note: SQLLite does not support monthnames. So you need to use substr(Date, 6,2) as month to get the months and substr(Date,0,5)='2015' for year.

sql SELECT BOOSTER_VERSION, LAUNCH_SITE FROM SPACEXTBL WHERE LANDING_OUTCOME = 'Failure (drone ship)' AND DATE_PART(

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

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

Task 10

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

sql SELECT LANDING DUTCOME, COUNT(*) AS DTY FROM SPACEXTBL MHERE

* solite:///www.datal.db

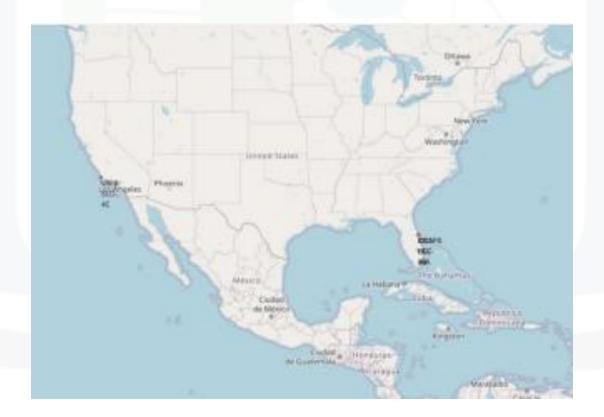
Part 3

Launch Sites Proximities Analysis



Launch Sites

Sites are on the coasts



Launch Successes and Fails



Distance from Site to Proximity



Build a Dashboard with Plotly Dash

SpaceX Launch Records Dashboard

All Sites

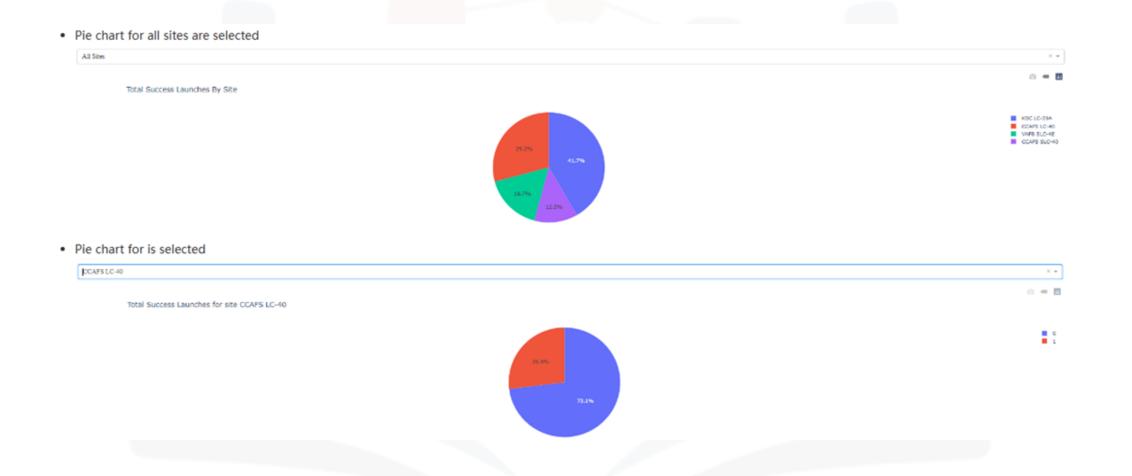
All Sites

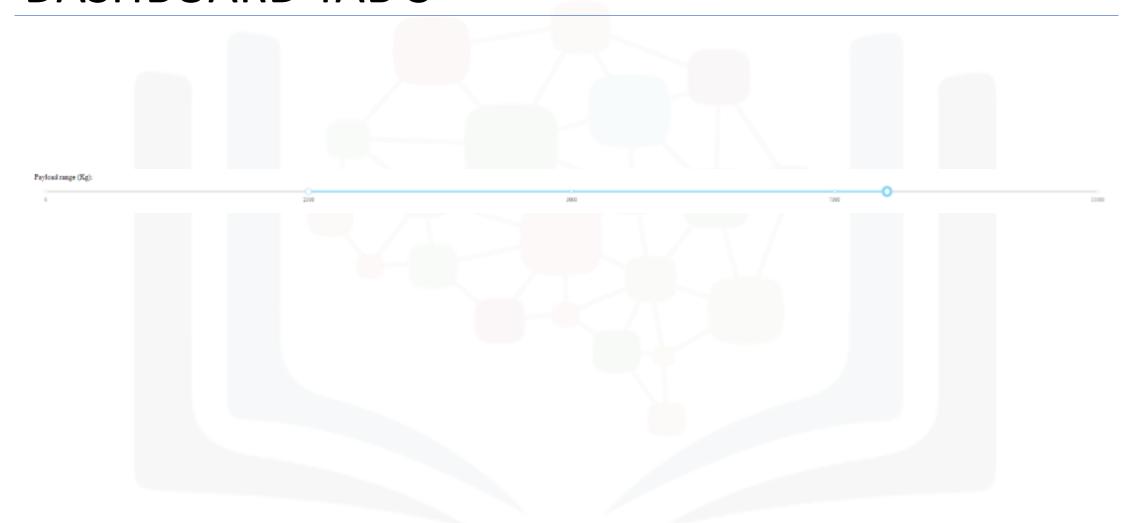
CCAFS LC-40

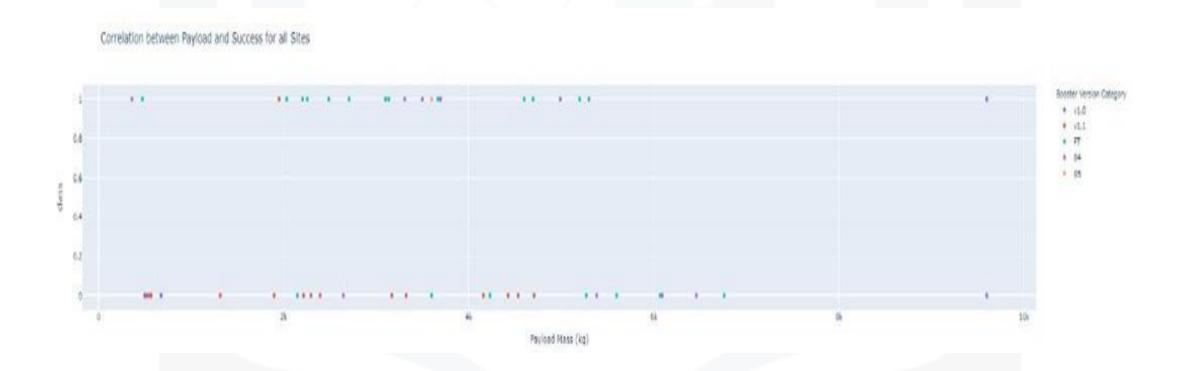
VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40







Part 5

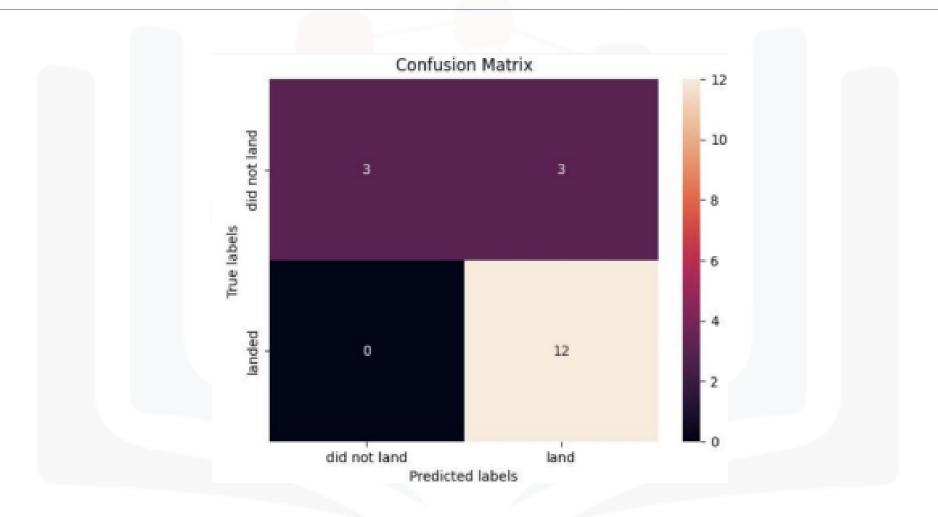
Predictive Analysis (Classification)

Classification Accuracy

They were all equal

Logistic_Reg	0.8333
SVM	0.8333
Decision Tree	0.8333
KNN	0.8333

Confusion Matrix



CONCLUSION

- Data was from SpaceX API and Wikipedia
- KSC LC-39A was the strongest site
- Launches under 7,000kg has more danger
- Over time, the data improved
- Sites are located near coasts; ensure less things in proximity
- More tests and algorithms are recommended for more accurate conclusions

APPENDIX

Numpy

Pandas

Matplotlib

