

Winning Space Race with Data Science

<Name> <Date>



Outline

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

Executive Summary

- Summary of methodologies
- Summary of all results

Introduction

- Project background and context
- Problems you want to find answers



Methodology

Executive Summary

- Data collection methodology:
 - SpaceX API and web scraping
- Perform data wrangling
 - Removing duplicate, handle missing value, formatting columns consistency
- 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

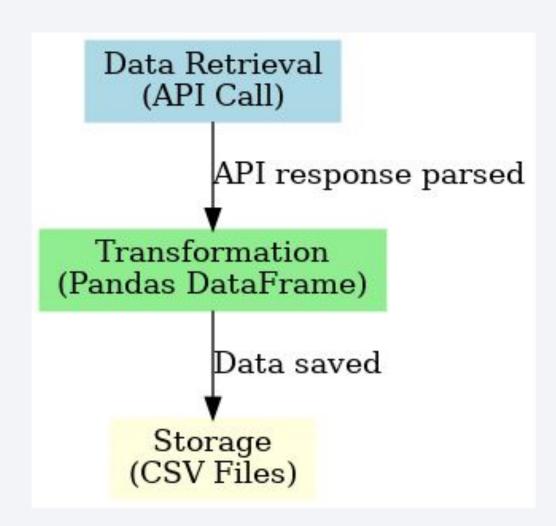
Data Collection

Sources:

- SpaceX API
- Web scraping

Types of data:

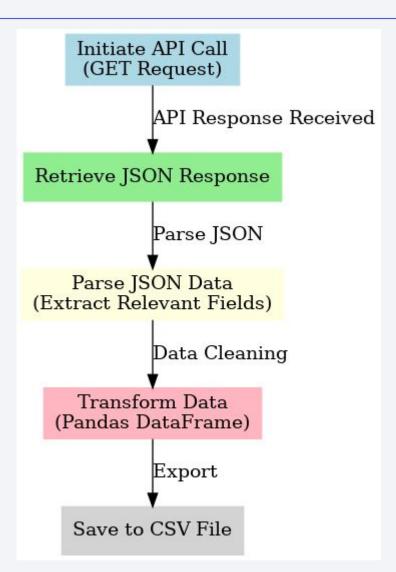
- Launch details
- Mission outcomes



Data Collection – SpaceX API

- Detailed informations about past rocket launches
- Payload details, launch outcomes and more

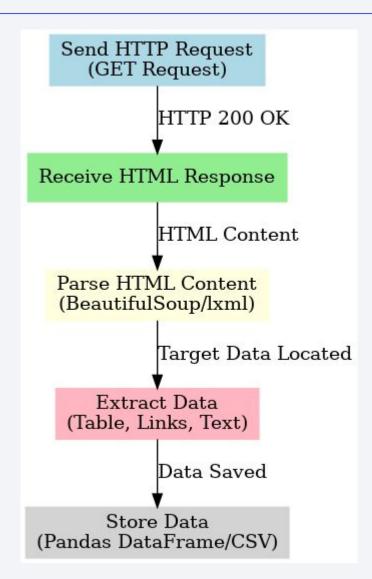
 https://github.com/Rhodham96/AppliedDataScien ceCapstone/blob/main/Lab1.1_CollectingData.ipy nb



Data Collection - Scraping

- Send HTTP Request
- Receive HTML Response
- Parse HTML content
- Extract Relevant Data
- Save Data

 https://github.com/Rhodham96/AppliedDa taScienceCapstone/blob/main/Lab1.2_W ebScrapping.ipynb



Data Wrangling

- Handling missing data
- removing duplicates
- Transformations

FlightNumber	int64
Date	object
File display ersion	object
PayloadMass	float64
Orbit	object
LaunchSite	object
Outcome	object
Flights	int64
GridFins	bool
Reused	bool
Legs	bool
LandingPad	object
Block	float64
ReusedCount	int64
Serial	object
Longitude	float64
Latitude	float64
dtype: object	

FlightNumber	0.000000
Date	0.000000
BoosterVersion	0.000000
PayloadMass	0.000000
Orbit	0.000000
LaunchSite	0.000000
Outcome	0.000000
Flights	0.000000
GridFins	0.00000
Reused	0.00000
Legs	0.000000
LandingPad	28.888889
Block	0.000000
ReusedCount	0.000000
Serial	0.000000
Longitude	0.000000
Latitude	0.000000
dtype: float64	

	FlightNumber	Date	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Flights	GridFins	Reused	Legs	LandingPac
display -	1	2010- 06- 04	Falcon 9	6104.959412	LEO	CCAFS SLC 40	None None	1	False	False	False	Nah
1	2	2012- 05- 22	Falcon 9	525.000000	LEO	CCAFS SLC 40	None None	1	False	False	False	Nah
2	3	2013- 03- 01	Falcon 9	677.000000	ISS	CCAFS SLC 40	None None	1	False	False	False	Naf
3	4	2013- 09- 29	Falcon 9	500.000000	PO	VAFB SLC 4E	False Ocean	1	False	False	False	Naf
4	5	2013- 12-03	Falcon 9	3170.000000	GTO	CCAFS SLC 40	None None	1	False	False	False	Nat
5	6	2014- 01- 06	Falcon 9	3325.000000	GTO	CCAFS SLC 40	None None	1	False	False	False	Nat
6	7	2014- 04- 18	Falcon 9	2296.000000	ISS	CCAFS SLC 40	True Ocean	1	False	False	True	Naf
7	8	2014- 07-14	Falcon 9	1316.000000	LEO	CCAFS SLC 40	True Ocean	1	False	False	True	Nat
8	9	2014- 08- 05	Falcon 9	4535.000000	GTO	CCAFS SLC 40	None None	1	False	False	False	Nat
9	10	2014- 09- 07	Falcon 9	4428.000000	GTO	CCAFS SLC 40	None None	1	False	False	False	Nal

• https://github.com/Rhodham96/AppliedDataScienceCapstone/blob/main/Lab2_DataWrangling.ipynb

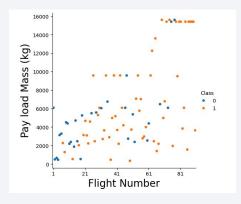
EDA with Data Visualization

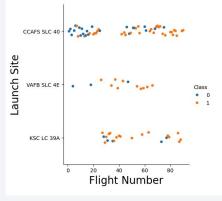
Scatter Plot: Flight Number vs. Launch Site, Flight Number vs. Payload

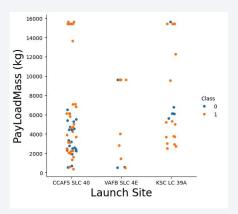
Mass, Launch Site vs. Payload Mass

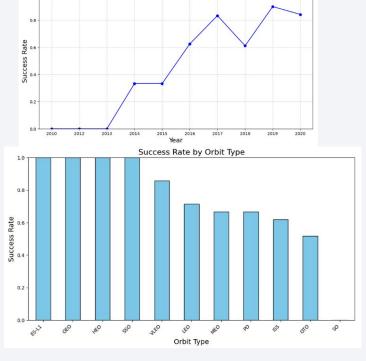
Bar Chart : Success rates across orbit types

Line Chart: Success rates over the years









Yearly Success Rate of Launches

• https://github.com/Rhodham96/AppliedDataScienceCapstone/blob/main/Lab4_ExplorePrepare.ipynb

EDA with SQL

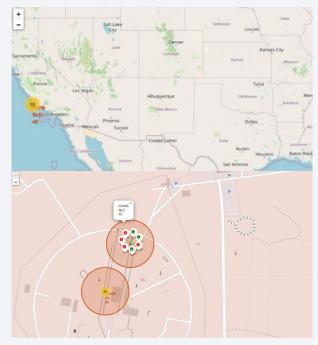
```
SELECT DISTINCT launch_site FROM SPACEXTBL;
SELECT * FROM SPACEXTBL WHERE launch_site LIKE 'CCA%' LIMIT 5;
SELECT SUM(PAYLOAD_MASS__KG_) AS total_payload_mass FROM SPACEXTBL WHERE Customer = 'NASA (CRS)';
SELECT AVG(PAYLOAD_MASS_KG_) AS total_payload_mass FROM SPACEXTBL WHERE Booster_Version LIKE 'F9 v1.1%';
SELECT MIN(Date) FROM SPACEXTBL WHERE Landing_Outcome = 'Success (ground pad)';
SELECT Booster_version FROM SPACEXTBL WHERE landing_outcome = 'Success (drone ship)' AND PAYLOAD_MASS_KG > 4000 AND PAYLOAD MASS KG < 6000;
SELECT landing outcome, COUNT(*) AS total FROM SPACEXTBL GROUP BY landing outcome;
SELECT landing_outcome, COUNT(*) AS total FROM SPACEXTBL WHERE landing_outcome LIKE 'Success%' OR landing_outcome LIKE 'Failure%' GROUP BY landing_outcome;
SELECT Booster_version FROM SPACEXTBL WHERE PAYLOAD_MASS__KG_ = (SELECT MAX(PAYLOAD_MASS__KG_) FROM SPACEXTBL)
```

• https://github.com/Rhodham96/AppliedDataScienceCapstone/blob/main/Lab3 SQL.ipynb

Build an Interactive Map with Folium

- All launch sites on a map
- Add success and failed launches markers
- Calculate distance between launch site and its proximities







Build a Dashboard with Plotly Dash

- Total launch success counts by site (pie chart)
 (Sites like KSC LC-39A and CCAFS LC-40 have higher success rate due to more frequent launches)
- Correlation between Payload and launch outcome (Scatter Plot)

(For lighter payloads (<5000kg), success is more consistent, For heavier payloads (>10000kg), success drop slightly, Certain booster versions perform better for heavier payloads.)

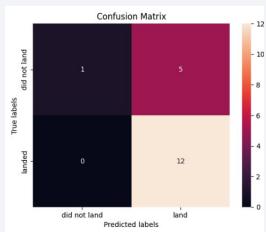


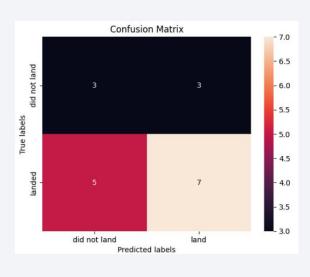
https://github.com/Rhodham96/AppliedDataScienceCapstone/blob/main/Lab6_Dashboard.py

Predictive Analysis (Classification)

- Logistic Regression (score = 0.888888888888...)
- Support Vector Machine (score = ?)
- Decision Tree Classifier (score = 0.722222222222222)
- K Nearest Neighbors (0.55555555555...)
- LR is better. (SVM I never could have a result)







https://github.com/Rhodham96/AppliedDataScienceCapstone/blob/main/Lab7_Predictions.ipynb

Results

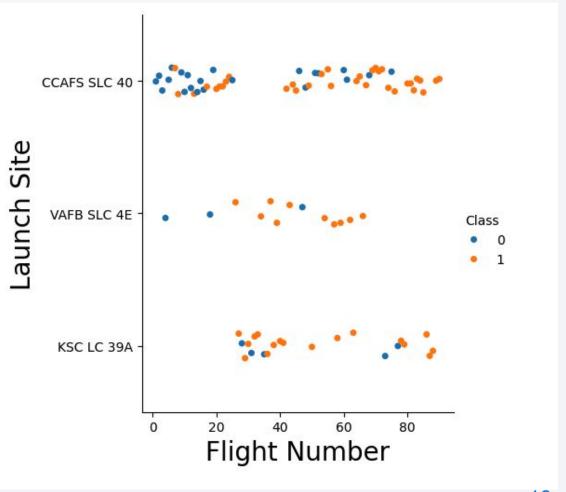
- Models used : Logistic Regression, SVM, Decision Tree, KNN
- Hyperparameter tuning with GridSearchCV
- Metrics: Accuracy, F1-score





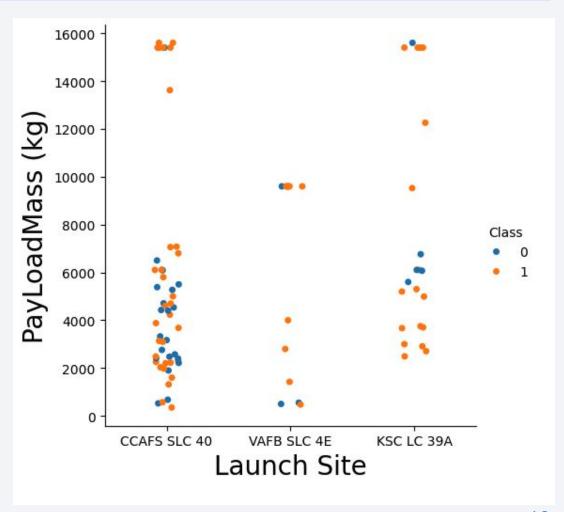
Flight Number vs. Launch Site

- When flight number gets bigger, success grow too
- Some launch sites where't user at beginning or failed but all became better



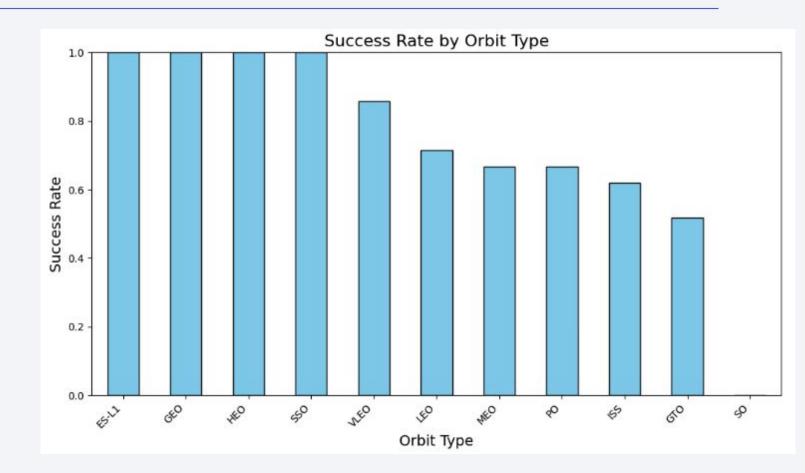
Payload vs. Launch Site

- CCAFS SLC 40 highest density of launches, various payload ranges, successful launches across payload ranges
- VAFBSLC 4E fewer launches, lighter payloads
- KSC LC 39A launches across the entire payload spectrum, success rate remain high, site is optimized for complex missions



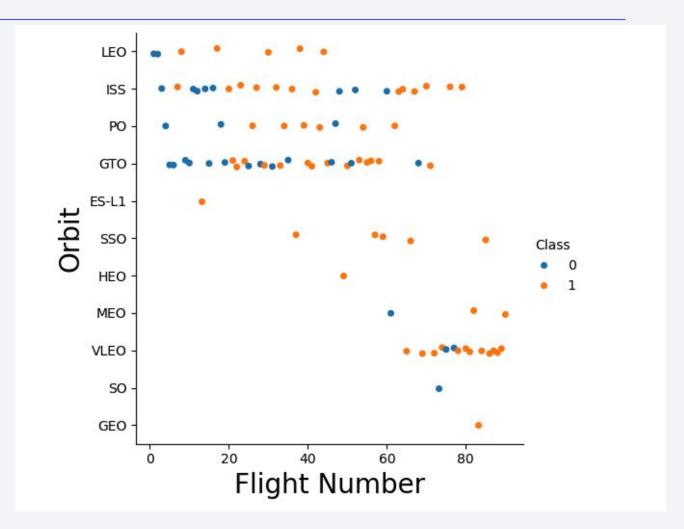
Success Rate vs. Orbit Type

- Orbits ES-L1, GEO, HEO ans SSO have perfect success rate
- VLEO, LEO, MEO, PO, success between 70%-90%
- GTO, SO lowest success rate (higher risk)



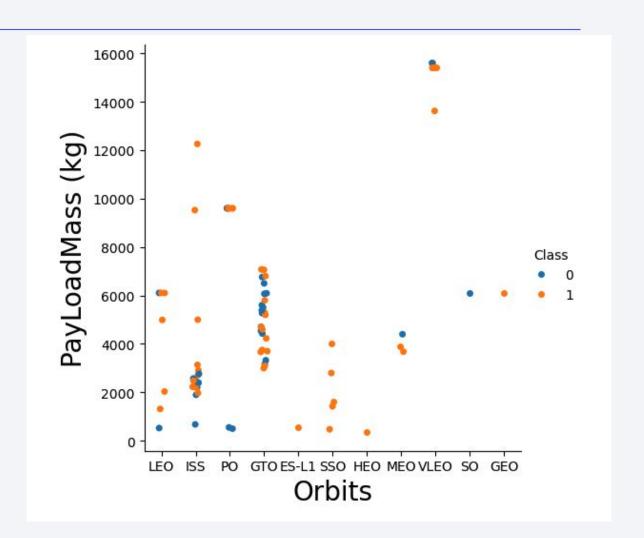
Flight Number vs. Orbit Type

- LEO and ISS frequent launches success dominate
- GTO failure rates are higher at earlier flight number, improvement over time
- SSO All successful
- For most orbit, success rate improve with higher flight number



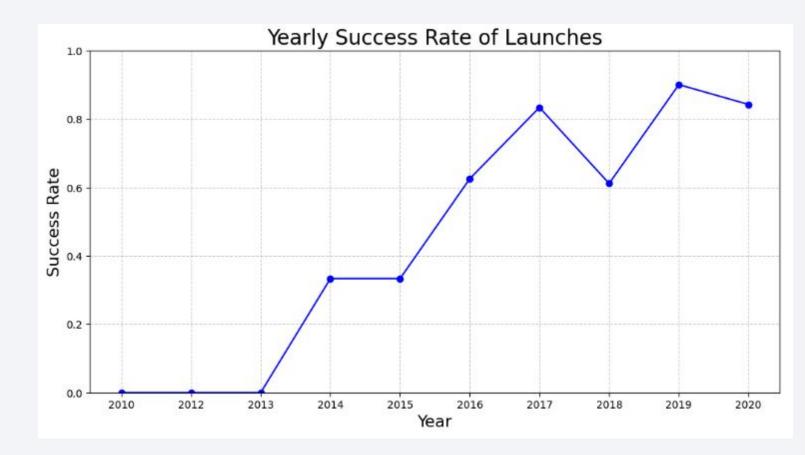
Payload vs. Orbit Type

- LEO wide range of payloads majority successful
- GTO mid range load mix success failure
- ISS lighter payloads (<8000kg)
- SSO more light (<6000kg)



Launch Success Yearly Trend

- Success increase over time
- Slight variablility in 2017-2018
- Stability in 2019-2020



All Launch Site Names

%sql SELECT DISTINCT launch_site FROM SPACEXTBL;

Launch_Site CCAFS LC-40 VAFB SLC-4E KSC LC-39A CCAFS SLC-40

Launch Site Names Begin with 'CCA'

 %sql SELECT * FROM SPACEXTBL WHERE launch_site LIKE 'CCA%' LIMIT 5;

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASSKG_	Orbit	Customer	Mission_Outcome	Landing_Outcome
2010-06- 04	18:45:00	F9 v1.0 B0003	CCAFS LC- 40	Dragon Spacecraft Qualification Unit	C) 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	C	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

 %sql SELECT SUM(PAYLOAD_MASS__KG_) AS total_payload_mass FROM SPACEXTBL WHERE Customer = 'NASA (CRS)';

> total_payload_mass 45596

Average Payload Mass by F9 v1.1

- %sql SELECT AVG(PAYLOAD_MASS__KG_) AS total_payload_mass FROM SPACEXTBL WHERE Booster_Version LIKE 'F9 v1.1%';
- Present your query result with a short explanation here

total_payload_mass

2534.666666666665

First Successful Ground Landing Date

%sql SELECT MIN(Date) FROM SPACEXTBL WHERE Landing_Outcome
 = 'Success (ground pad)';



Successful Drone Ship Landing with Payload between 4000 and 6000

 %sql SELECT Booster_version FROM SPACEXTBL WHERE landing_outcome = 'Success (drone ship)' AND PAYLOAD_MASS__KG_ > 4000 AND PAYLOAD_MASS__KG_ < 6000;



Total Number of Successful and Failure Mission Outcomes

 %sql SELECT landing_outcome, COUNT(*) AS total FROM SPACEXTBL GROUP BY landing_outcome;

Landing_Outcome	total
Controlled (ocean)	5
Failure	3
Failure (drone ship)	5
Failure (parachute)	2
No attempt	21
No attempt	1
Precluded (drone ship)	1
Success	38
Success (drone ship)	14
Success (ground pad)	9
Uncontrolled (ocean)	2

Boosters Carried Maximum Payload

 %sql SELECT Booster_version FROM SPACEXTBL WHERE PAYLOAD_MASS__KG_ = (SELECT MAX(PAYLOAD_MASS__KG_) FROM SPACEXTBL);

> F9 B5 B1048.4 F9 B5 B1049.4 F9 B5 B1051.3 F9 B5 B1056.4 F9 B5 B1048.5 F9 B5 B1051.4 F9 B5 B1049.5 F9 B5 B1060.2 F9 B5 B1058.3 F9 B5 B1051.6 F9 B5 B1060.3 F9 B5 B1049.7

Booster_Version

2015 Launch Records

SELECT month_mapping.month_name, Booster_version, Launch_site, landing_outcome FROM SPACEXTBL JOIN month_mapping ON substr(Date, 6, 2) = month_mapping.month_num WHERE substr(Date, 1, 4) = '2015' AND landing_outcome = 'Failure (drone ship)';

month_name	Booster_Version	Launch_Site	Landing_Outcome
January	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
April	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

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

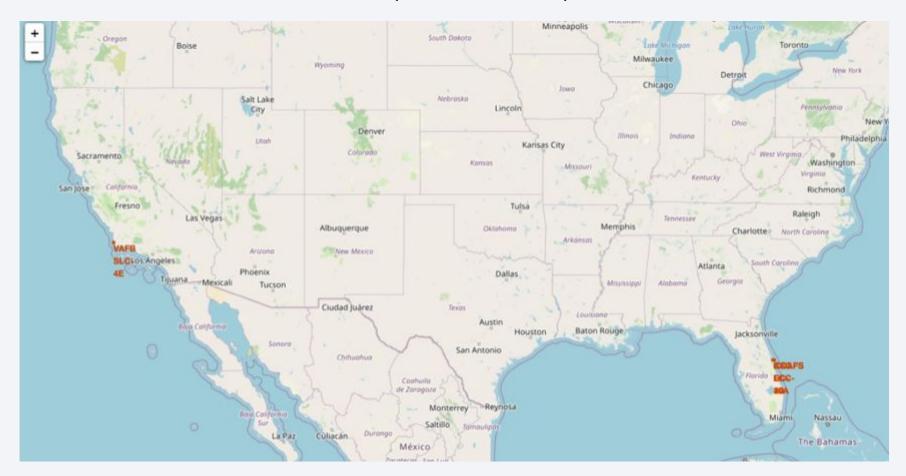
 %%sql SELECT landing_outcome, COUNT(*) AS outcome_count FROM SPACEXTBL WHERE Date BETWEEN '2010-06-04' AND '2017-03-20' GROUP BY landing_outcome ORDER BY outcome_count DESC;

outcome_count
10
5
5
3
3
2
2
1



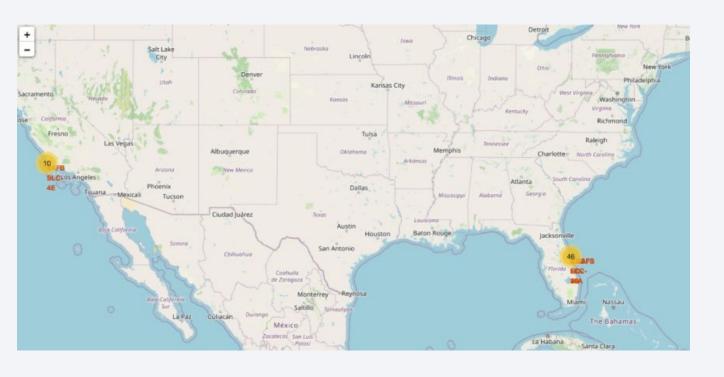
Mark all launch sites on a map

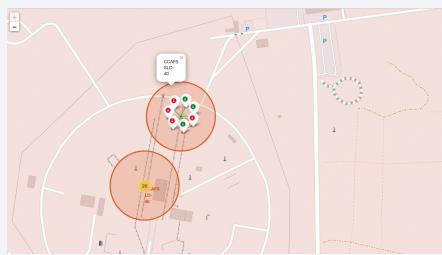
We can see where are all launch site (near the see)



Marks the success/failed launches on the map

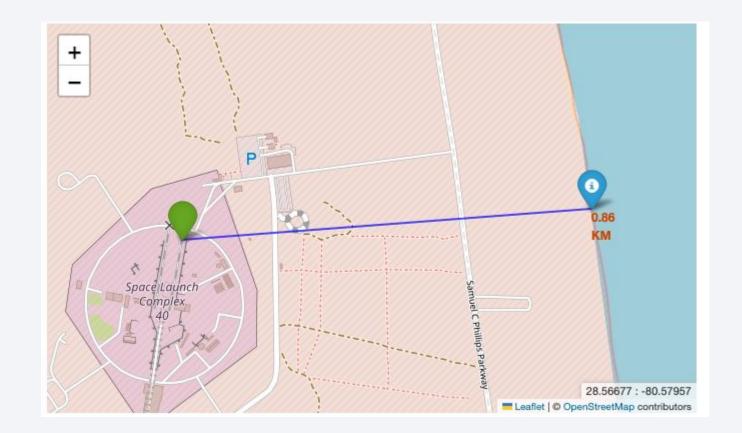
• We can see launches are on launch sites, if we zoom we can click on them and know if success or failed





Calculate the distance between launch sites and proximities

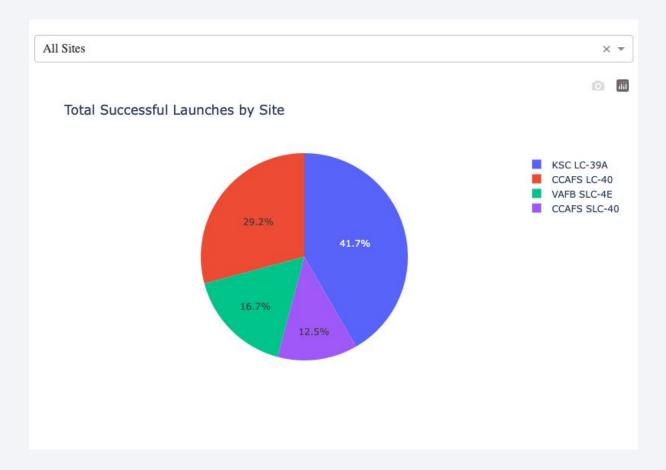
• There is 0.86KM between launch site and coast line





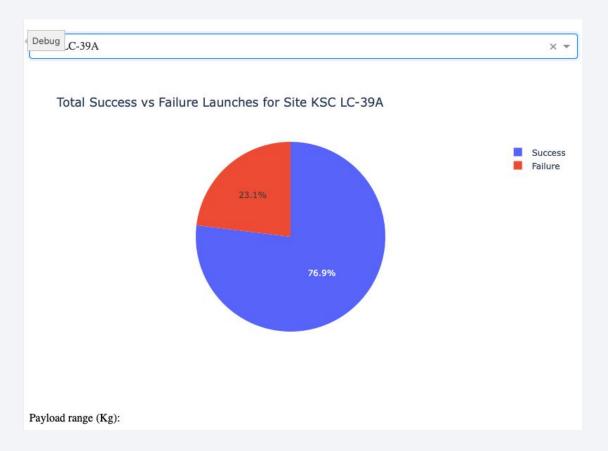
Launch sites repartition of success

KSC Has more success



Success vs. Failure for KSC LC-39A

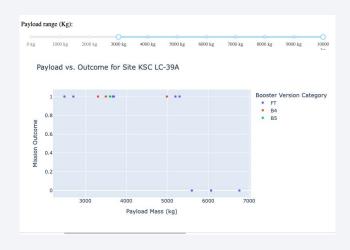
Almost 72% success



Payload vs. Outcome for KSC LC-39A

• We can see differences between booster version with different payload.









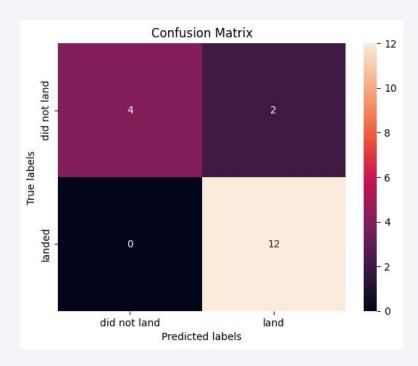
Classification Accuracy

 Visualize the built model accuracy for all built classification models, in a bar chart

 Find which model has the highest classification accuracy

Confusion Matrix

• Its almost perfect: 0 false didn't land and 2 false land



Conclusions

1. Launch Site Analysis:

- Florida-based launch sites (KSC LC-39A and CCAFS LC-40) have higher success rates, reflecting the optimization for frequent missions like geostationary and low Earth orbit launches.
- The West Coast site (VAFB SLC-4E) focuses on polar orbit missions, with fewer launches but consistent success rates for lighter payloads.

2. Payload and Outcome:

- Lighter payloads (<5000 kg) are generally more successful across all sites.
- For heavier payloads (>10,000 kg), low Earth orbits (LEO) dominate and maintain high success rates, indicating advancements in handling complex missions.
- Geostationary Transfer Orbits (GTO) show a higher risk profile with mixed success and failures, particularly in the mid-payload range (4000–6000 kg).

3. Orbit-Specific Success Trends:

- Orbits like **SSO**, **GEO**, and **ES-L1** exhibit near-perfect success rates, suggesting reliability in specialized missions.
- GTO and SO orbits experience lower success rates due to higher complexity and risk.

4. Trends Over Time:

- Success rates have improved significantly since 2013, with consistent performance above 80% since 2016.
- Early failures reflect the learning curve, while recent stability showcases operational maturity.

5. **Predictive Modeling Insights**:

- Among classification models, Logistic Regression achieved the highest accuracy (88.89%).
- Predictive insights indicate that payload, orbit type, and launch site are key factors in determining mission success.

Appendix

https://github.com/Rhodham96/AppliedDataScienceCapstone

