



**SPACE Y
FIRST STAGE: RE USE**

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Executive Summary

Summary of Methodologies

This study seeks to identify key factors influencing successful rocket landings. The following methodologies were applied:

Data Collection: Data was gathered through the SpaceX REST API and web scraping techniques.

Data Wrangling: Data was processed to create a variable for success/failure outcomes.

Exploratory Data Analysis: Visualization techniques were used to analyze factors like payload, launch site, flight frequency, and annual trends.

SQL Analysis: Data was examined to calculate statistics, including total payload, payload range for successful launches, and counts of successful and failed landings.

Launch Site Success Rates: Examined success rates by launch site and their proximity to geographical landmarks.

Data Visualization: Mapped and highlighted launch sites with the highest success rates and successful payload ranges.

Predictive Modeling: Developed models to predict landing outcomes using logistic regression, support vector machine (SVM), decision tree, and K-nearest neighbor (KNN) algorithms.

Visualization/Analytics:

- Most launch sites are near the equator, and all are close to the coast

Predictive Analytics:

- All models performed similarly on the test set. The decision tree model slightly outperformed

Exploratory Data Analysis:

- Launch success has improved over time
- KSC LC-39A has the highest success rate among landing sites
- Orbit ES-L1, GEO, HEO, and SSO have a 100% success rate

Introduction

Background

As a pioneer in the space industry, SpaceX is committed to making space travel accessible and affordable. Its achievements include launching spacecraft to the International Space Station, deploying a satellite network to provide global internet access, and executing crewed space missions. SpaceX achieves cost-effective launches (around \$62 million each) by reusing the first stage of its Falcon 9 rocket, a significant advantage over other providers whose non-reusable rockets can cost over \$165 million per launch. By predicting the likelihood of first-stage landings, we can better estimate launch costs. This study uses public data and machine learning models to determine whether SpaceX—or a competitor—can successfully reuse the first stage of its rockets.

Key Areas of Exploration

How factors like payload mass, launch site, flight count, and orbital paths impact first-stage landing success.

Trends in landing success rates over time.

Identification of the most effective predictive model for successful landings (binary classification).

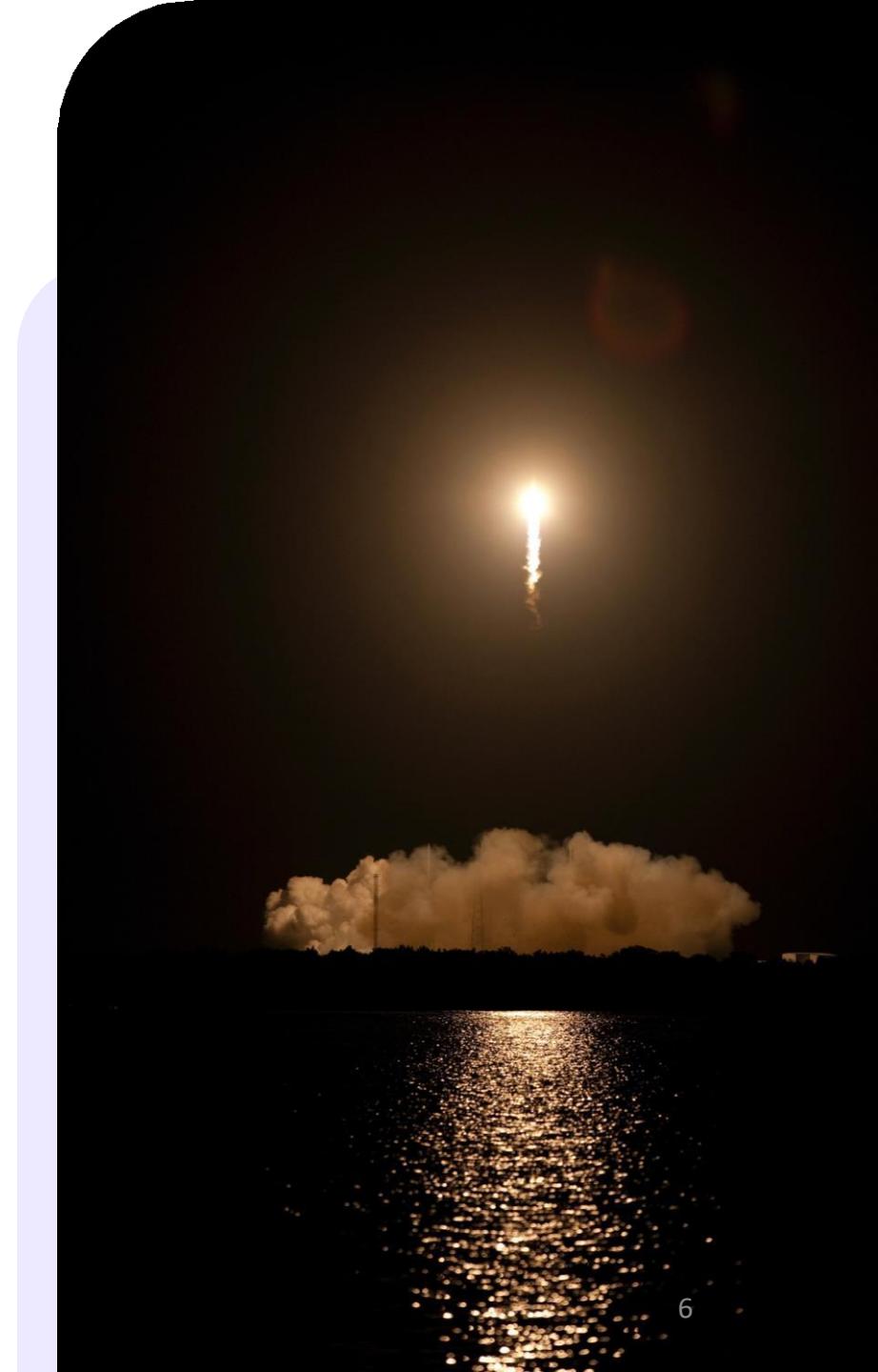
METHODOLOGY



Methodology

Steps

- **Collect** data using SpaceX REST API and web scraping techniques
- **Wrangle** data –by filtering the data, handling missing values and applying one hot encoding –to prepare the data for analysis and modeling
- **Explore** data via EDA with SQL and data visualization techniques
- **Visualize** the data using Folium and Plotly Dash
- **Build Models** to predict landing outcomes using classification models. Tune and evaluate models to find best model and parameters



Data Collection –API

Steps

- **Request data** from SpaceX API (rocket launch data)
- **Decode response** using .json() and convert to a dataframe using .json_normalize()
- **Request information** about the launches from SpaceX API using custom functions
- **Create dictionary** from the data
- **Create dataframe** from the dictionary
- **Filter dataframe** to contain only Falcon 9 launches
- **Replace missing values** of Payload Mass with calculated .mean()
- **Export data** to csv file



Data Collection –Web Scraping

Steps

- **Request data** (Falcon 9 launch data) from Wikipedia
- **Create BeautifulSoup object** from HTML response
- **Extract column names** from HTML table header
- **Collect data** from parsing HTML tables
- **Create dictionary** from the data
- **Create dataframe** from the dictionary
- **Export data** to csv file



Data Wrangling

Steps

- **Perform EDA** and determine data labels
- **Calculate:**
 - #of launches for each site
 - #and occurrence of orbit
 - #and occurrence of mission outcome per orbit type]
- **Create binary** landing outcome column (dependent variable)
- **Export data** to csv file

Landing Outcome

- Landing was not always successful
- **True Ocean:** mission outcome had a successful landing to a specific region of the ocean

Landing Outcome Cont.

- **False Ocean:** represented an unsuccessful landing to a specific region of ocean
- **True RTLS:** meant the mission had a successful landing on a ground pad
- **False RTLS:** represented an unsuccessful landing on a ground pad
- **True ASDS:** meant the mission outcome had a successful landing on a drone ship
- **False ASDS:** represented an unsuccessful landing on drone ship
- **Outcomes converted** into 1 for a successful landing and 0 for an unsuccessful landing

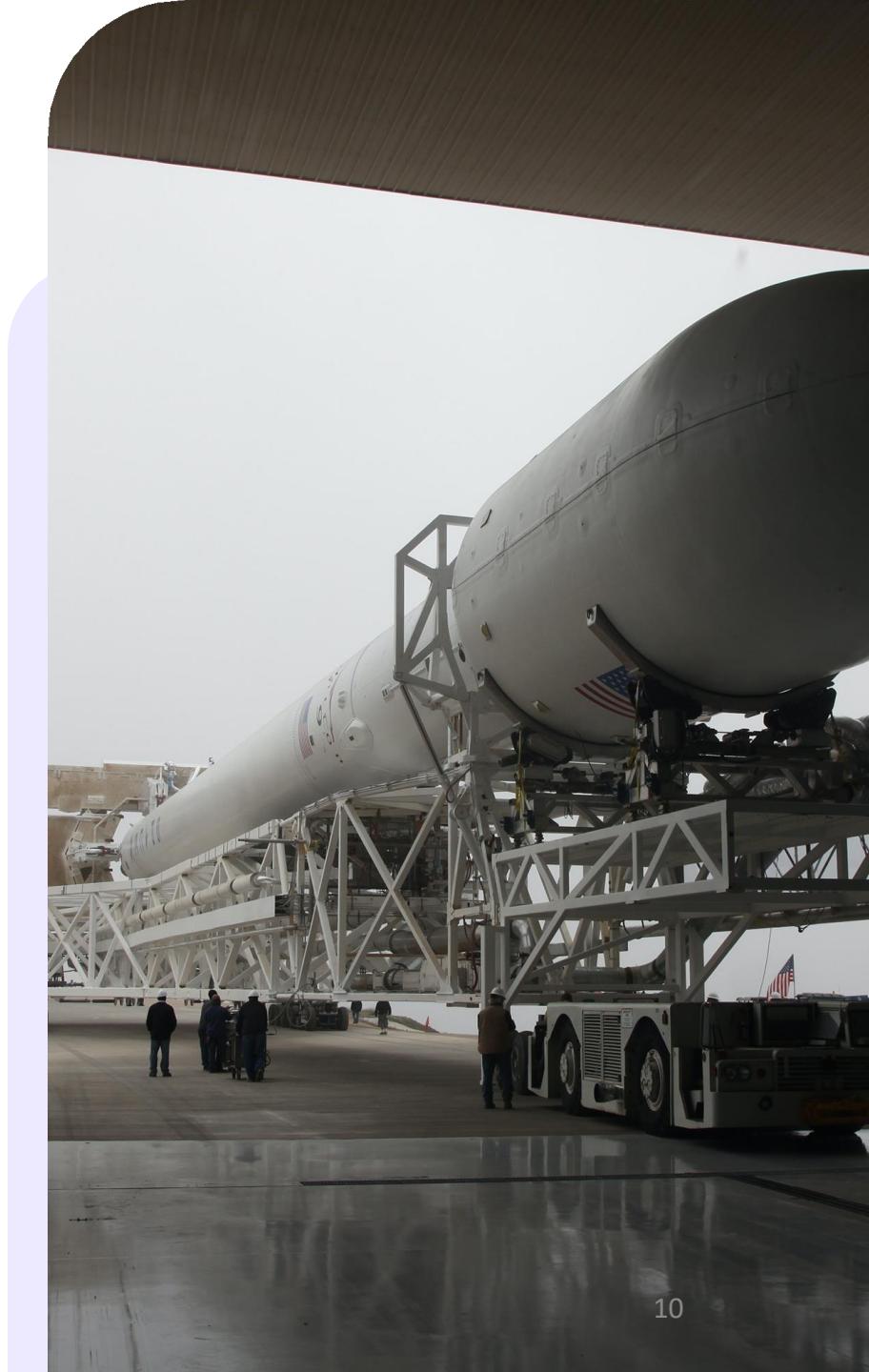
EDA with Visualization

Charts

- Flight Number vs. Payload
- Flight Number vs. Launch Site
- Payload Mass (kg) vs. Launch Site
- Payload Mass (kg) vs. Orbit type

Analysis

- **View relationship** by using **scatter plots**. The variables could be useful for machine learning if a relationship exists
- **Show comparisons** among discrete categories with **bar charts**. Bar charts show the relationships among the categories and a measured value.



EDA with SQL

Queries

Display:

- Names of unique launch sites
- 5 records where launch site begins with 'CCA'
- Total payload mass carried by boosters launched by NASA (CRS)
- Average payload mass carried by booster version F9 v1.1.

List:

- Date of first successful landing on ground pad
- Names of boosters which had success landing on drone ship and have payload mass greater than 4,000 but less than 6,000
- Total number of successful and failed missions
- Names of booster versions which have carried the max payload
- Failed landing outcomes on drone ship, their booster version and launch site for the months in the year 2015
- Count of landing outcomes between 2010-06-04 and 2017-03-20 (desc)



Map with Folium

Markers Indicating Launch Sites

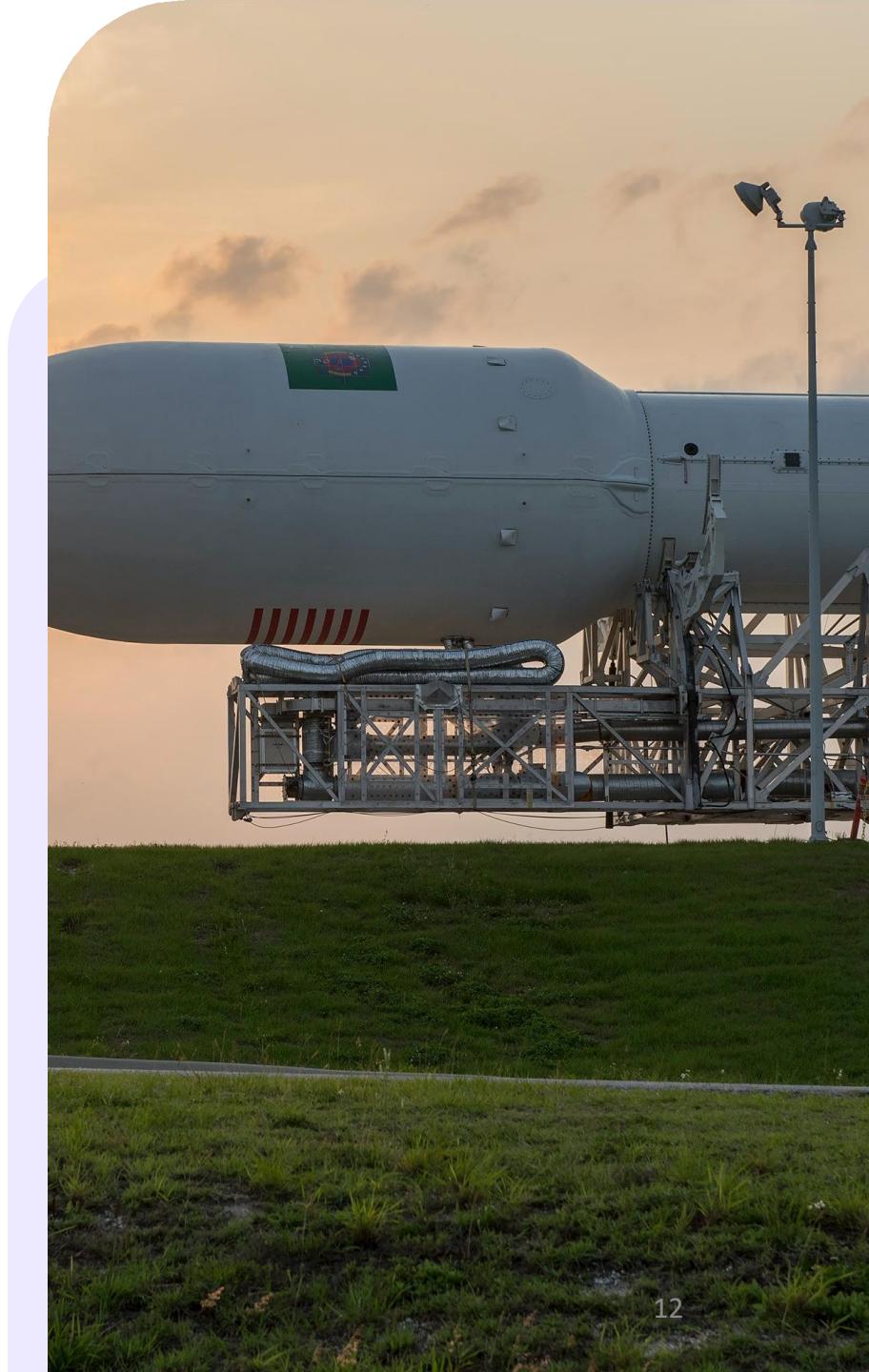
- Added **blue circle** at **NASA Johnson Space Center's coordinate** with a **popup label** showing its name using its latitude and longitude coordinates
- Added **red circles** at **all launch sites coordinates** with a **popup label** showing its name using its name using its latitude and longitude coordinates

Colored Markers of Launch Outcomes

- Added **colored markers** of **successful (green)** and **unsuccessful (red)** **launches** at each launch site to show which launch sites have high success rates

Distances Between a Launch Site to Proximities

- Added **colored lines** to **show distance between** launch site **CCAFS SLC-40** and its proximity to the **nearest coastline, railway, highway, and city**



Dashboard with Plotly Dash

Dropdown List with Launch Sites

- Allow user to select all launch sites or a certain launch site

Pie Chart Showing Successful Launches

- Allow user to see successful and unsuccessful launches as a percent of the total

Slider of Payload Mass Range

- Allow user to select payload mass range

Scatter Chart Showing Payload Mass vs. Success Rate by Booster Version

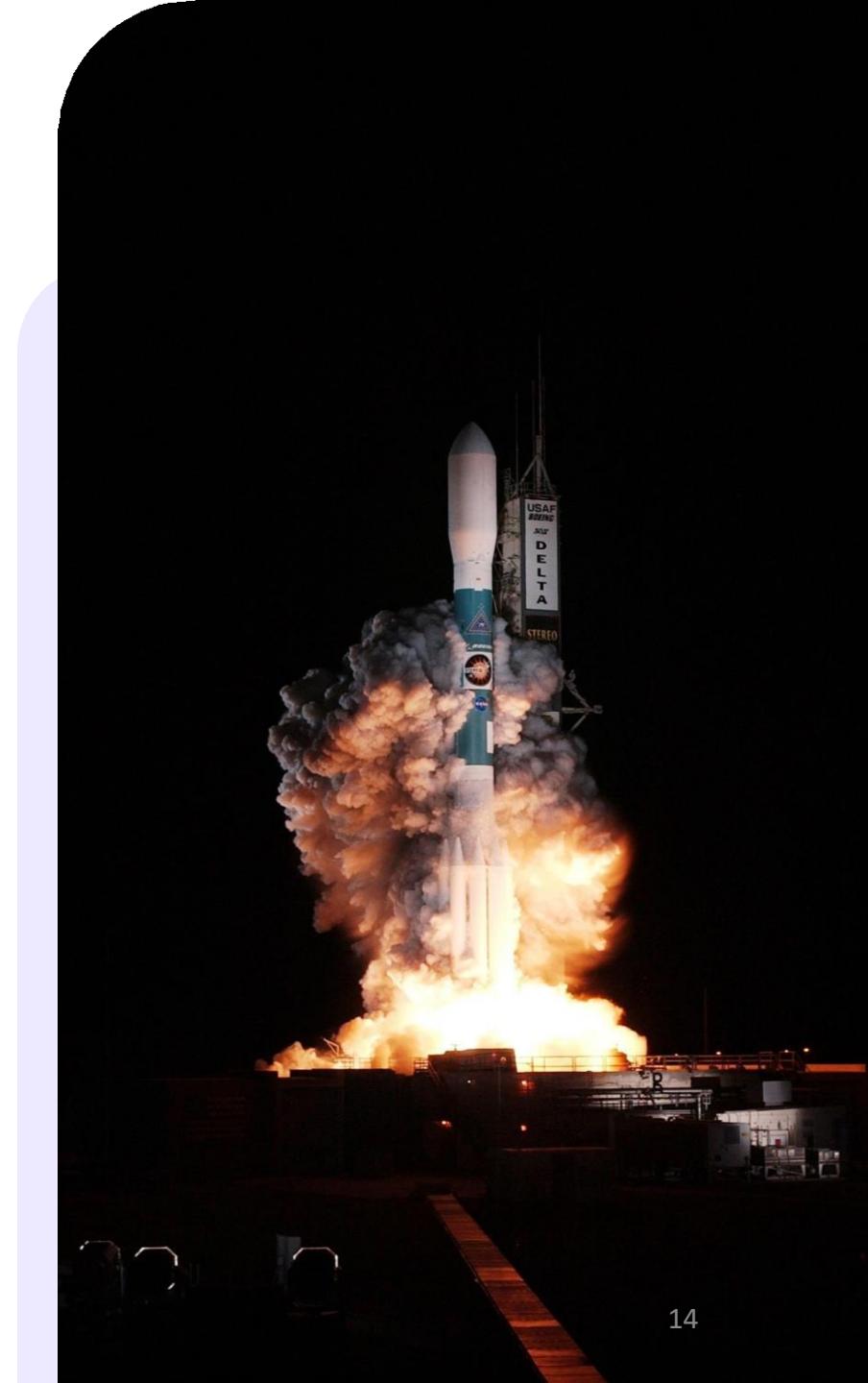
- Allow user to see the correlation between Payload and Launch Success



Predictive Analytics

Charts

- **Create** NumPy array from the Class column
- **Standardize** the data with StandardScaler. Fit and transform the data.
- **Split** the data using train_test_split
- **Create** a GridSearchCV object with cv=10 for parameter optimization
- **Apply** GridSearchCV on different algorithms: logistic regression (LogisticRegression()), support vector machine (SVC()), decision tree (DecisionTreeClassifier()), K-Nearest Neighbor (KNeighborsClassifier())
- **Calculate** accuracy on the test data using .score() for all models
- **Assess** the confusion matrix for all models
- **Identify** the best model using Jaccard_Score, F1_Score and Accuracy



Results



Results Summary

Exploratory Data Analysis

Over time, the success rate of launches has shown improvement.

Among landing sites, KSC LC-39A boasts the highest success rate.

The orbits ES-L1, GEO, HEO, and SSO have achieved a perfect success rate of 100%.

Visual Analytics

The majority of launch sites are located near the equator and in close proximity to coastal areas.

These launch sites are strategically positioned far enough from potential hazards (such as cities, highways, and railways) to mitigate the risk of damage from failed launches, while remaining accessible for transporting personnel and materials needed for launch operations.

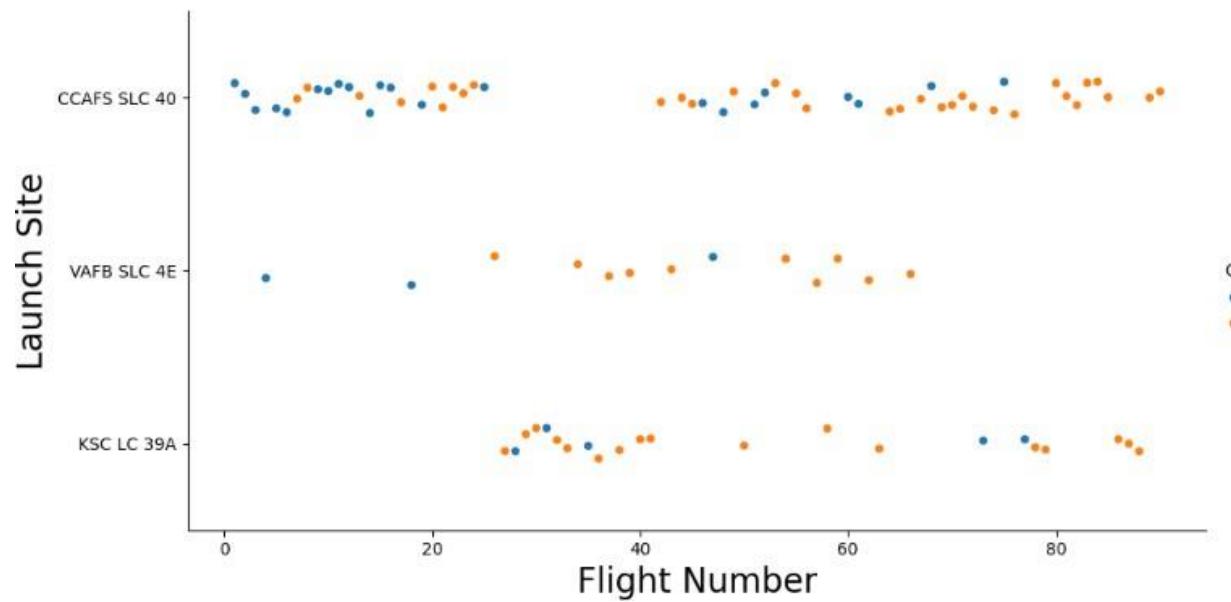
Predictive Analytics

The Decision Tree model stands out as the most effective predictive model for this dataset.

Flight Number vs. Launch Site

Exploratory Data Analysis

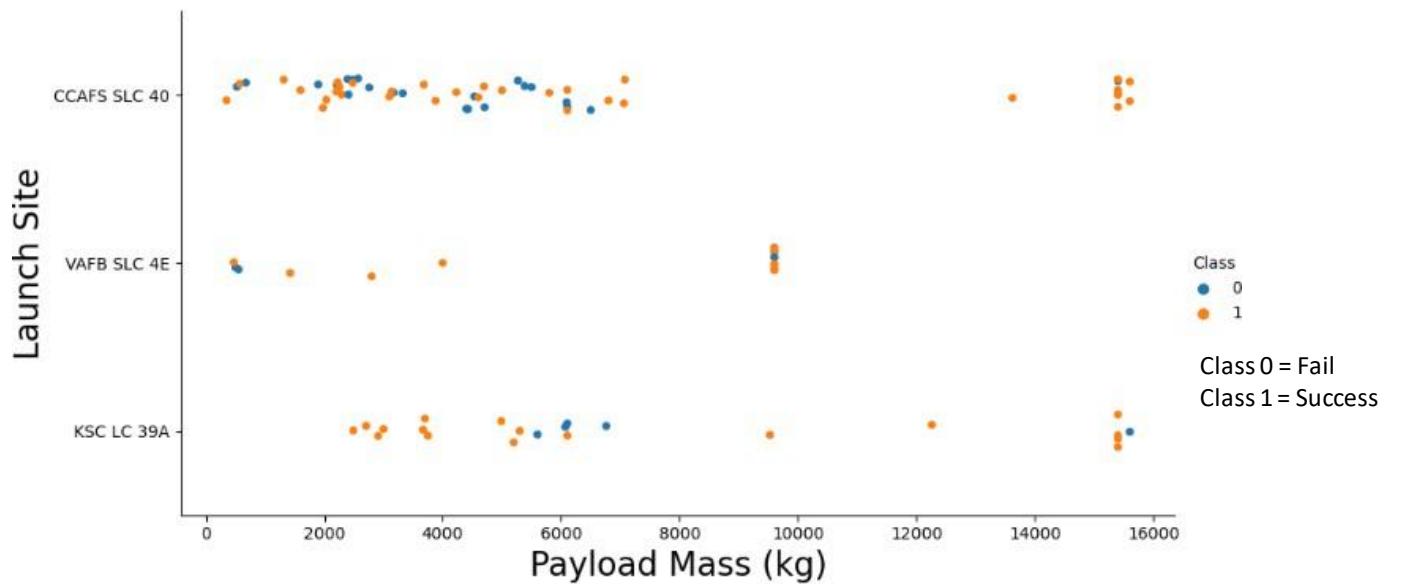
- **Earlier flights** had a **lower success rate** (**blue = fail**)
- **Later flights** had a **higher success rate** (**orange = success**)
- Around half of launches were from CCAFS SLC 40 launch site
- VAFB SLC 4E and KSC LC 39A have higher success rates
- We can infer that new launches have a higher success rate



Payload vs. Launch Site

Exploratory Data Analysis

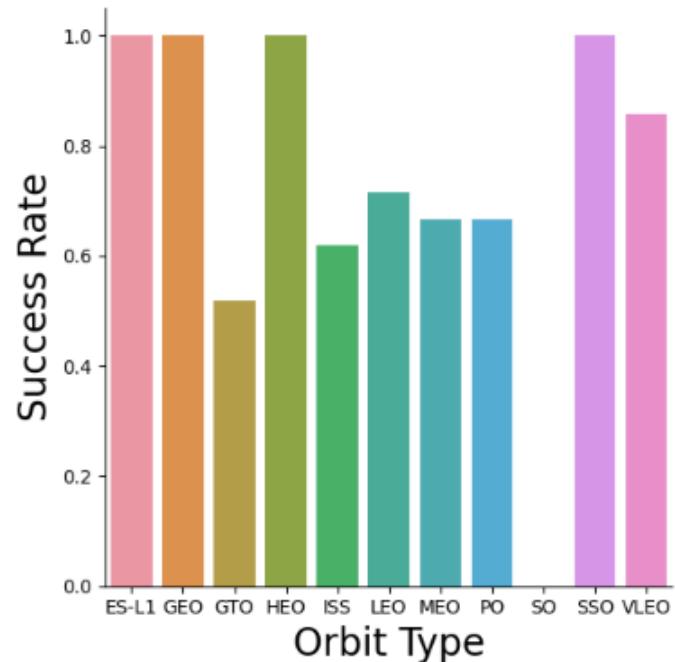
- Typically, the **higher** the **payload mass** (kg), the **higher** the **success rate**
- Most launches with a payload greater than 7,000 kg were successful
- KSC LC 39A has a 100% success rate for launches less than 5,500 kg
- VAFB SKC 4E has not launched anything greater than ~10,000 kg



Success Rate by Orbit

Exploratory Data Analysis

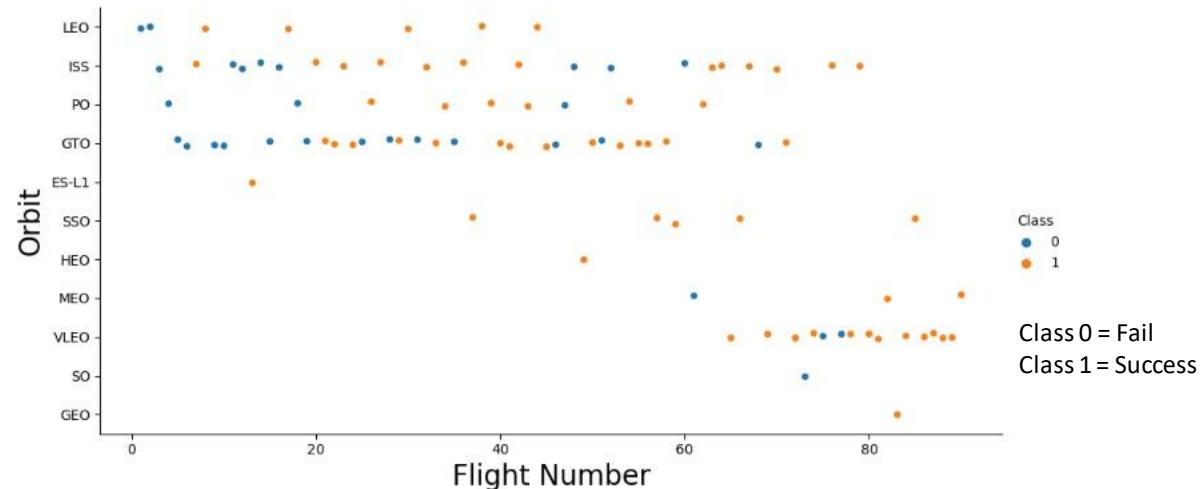
- **100% Success Rate:** ES-L1, GEO, HEO and SSO
- **50%-80% Success Rate:** GTO, ISS, LEO, MEO, PO
- **0% Success Rate:** SO



Flight Number vs. Orbit

Exploratory Data Analysis

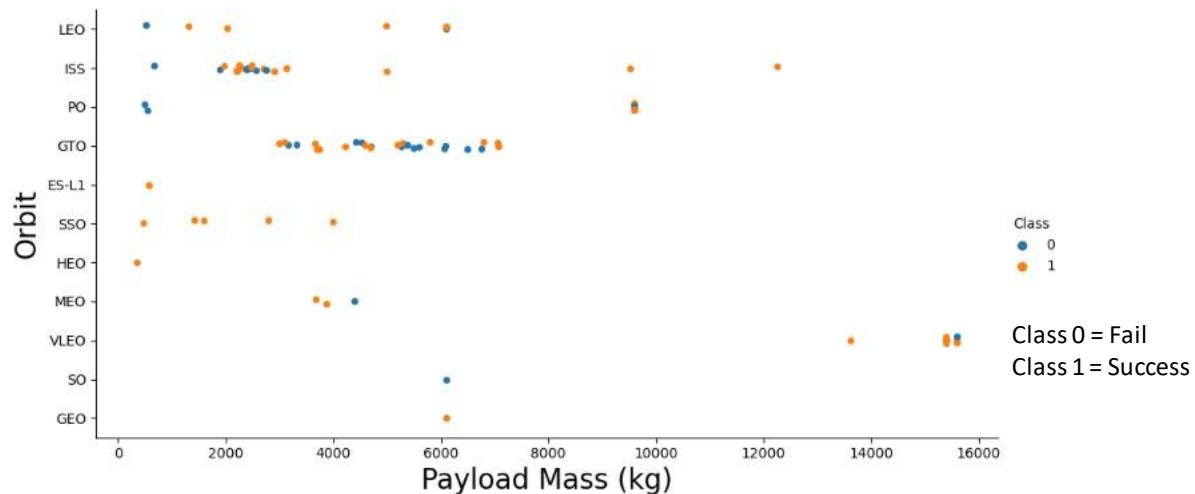
- The success rate typically increases with the number of flights for each orbit
- This relationship is highly apparent for the LEO orbit
- The GTO orbit, however, does not follow this trend



Payload vs. Orbit

Exploratory Data Analysis

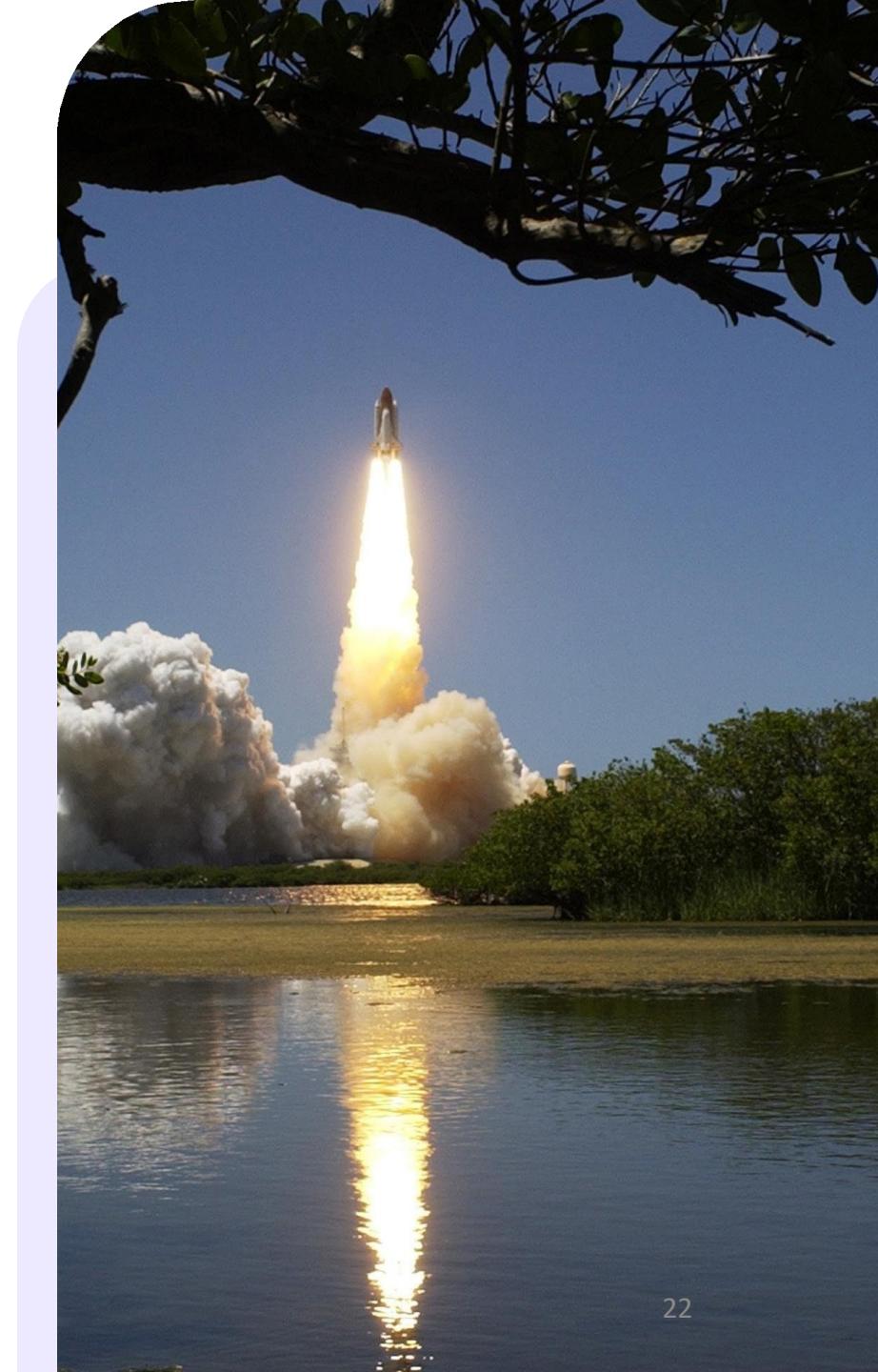
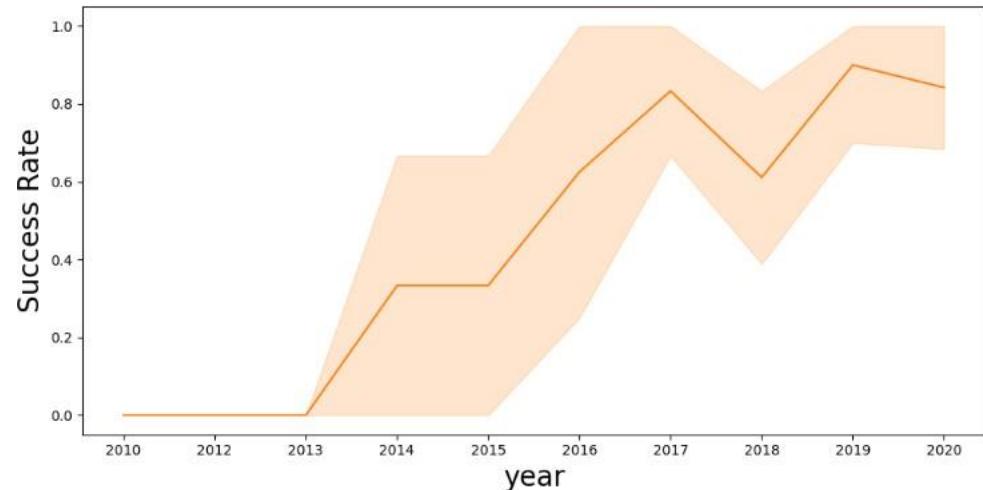
- Heavy payloads are better with LEO, ISS and PO orbits
- The GTO orbit has mixed success with heavier payloads



Launch Success over Time

Exploratory Data Analysis

- The success rate improved from 2013-2017 and 2018-2019
- The success rate decreased from 2017-2018 and from 2019-2020
- Overall, the success rate has improved since 2013



Launch Site Information

Launch Site Names

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

Landing Outcome Cont.

```
[30]: %sql ibm_db_sa://yyy33800:dwNkg8J3L0IBd6CP@1bbf73c5  
%sql SELECT Unique(LAUNCH_SITE) FROM SPACEXTBL;  
  
* ibm_db_sa://yyy33800:***@1bbf73c5-d84a-4bb0-85b6  
sqlite:///my_data1.db  
Done.  
  
[30]: launch_site  
CCAFS LC-40  
CCAFS SLC-40  
KSC LC-39A  
VAFB SLC-4E
```

Records with Launch Site Starting with CCA

- Displaying 5 records below

```
%sql SELECT * \  
FROM SPACEXTBL \  
WHERE LAUNCH_SITE LIKE 'CCA%' LIMIT 5;
```

* ibm_db_sa://yyy33800:***@1bbf73c5-d84a-4bb0-85b6-ab1a4348f4a4.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:32286/BLUDB
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	07:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
2012-10-08	00:35:00	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

Payload Mass

Total Payload Mass

- **45,596 kg** (total) carried by boosters launched by NASA (CRS)

```
%sql SELECT SUM(PAYLOAD_MASS__KG_) \
    FROM SPACEXTBL \
    WHERE CUSTOMER = 'NASA (CRS)';

* ibm_db_sa://yyy33800:***@1bbf73c5-d84a-41
sqlite:///my_data1.db
Done.

1
45596
```

Average Payload Mass

- **2,928 kg** (average) carried by booster version F9 v1.1

```
%sql SELECT AVG(PAYLOAD_MASS__KG_) \
    FROM SPACEXTBL \
    WHERE BOOSTER_VERSION = 'F9 v1.1';

* ibm_db_sa://yyy33800:***@1bbf73c5-d84a-41
sqlite:///my_data1.db
Done.

1
2928
```



Landing & Mission Info

1st Successful Landing in Ground Pad

- 12/22/2015

```
%sql SELECT MIN(DATE) \
FROM SPACEXTBL \
WHERE LANDING_OUTCOME = 'Success_(ground_pad)'

* ibm_db_sa://yyy33800:***@1bbf73c5-d84a-4bb0-85b9
sqlite:///my_data1.db

Done.

1
2015-12-22
```

Booster Drone Ship Landing

- Booster mass greater than 4,000 but less than 6,000
- JSCAT-14, JSCAT-16, SES-10, SES-11 / EchoStar 105

```
%sql SELECT PAYLOAD \
FROM SPACEXTBL \
WHERE LANDING_OUTCOME = 'Success_(drone_ship)' \
AND PAYLOAD_MASS_KG BETWEEN 4000 AND 6000;

* ibm_db_sa://yyy33800:***@1bbf73c5-d84a-4bb0-85b9
sqlite:///my_data1.db

Done.

payload
_____
JCSAT-14
JCSAT-16
SES-10
SES-11 / EchoStar 105
```

Total Number of Successful and Failed Mission Outcomes

- 1 Failure in Flight
- 99 Success
- 1 Success (payload status unclear)

```
%sql SELECT MISSION_OUTCOME, COUNT(*) as total_number \
FROM SPACEXTBL \
GROUP_BY MISSION_OUTCOME;

* sqlite:///my_data1.db

Done.

Mission_Outcome  total_number
_____
Failure (in flight)    1
Success           98
Success           1
Success (payload status unclear) 1
```



Boosters

Carrying Max Payload

- 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

```
%sql SELECT BOOSTER_VERSION \
FROM SPACEXTBL \
WHERE PAYLOAD_MASS_KG = (SELECT MAX(PAYLOAD_MASS_KG) FROM SPACEXTBL);  
* sqlite:///my_data1.db  
Done.  
  
Booster_Version  
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
```



Failed Landings on Drone Ship

In 2015

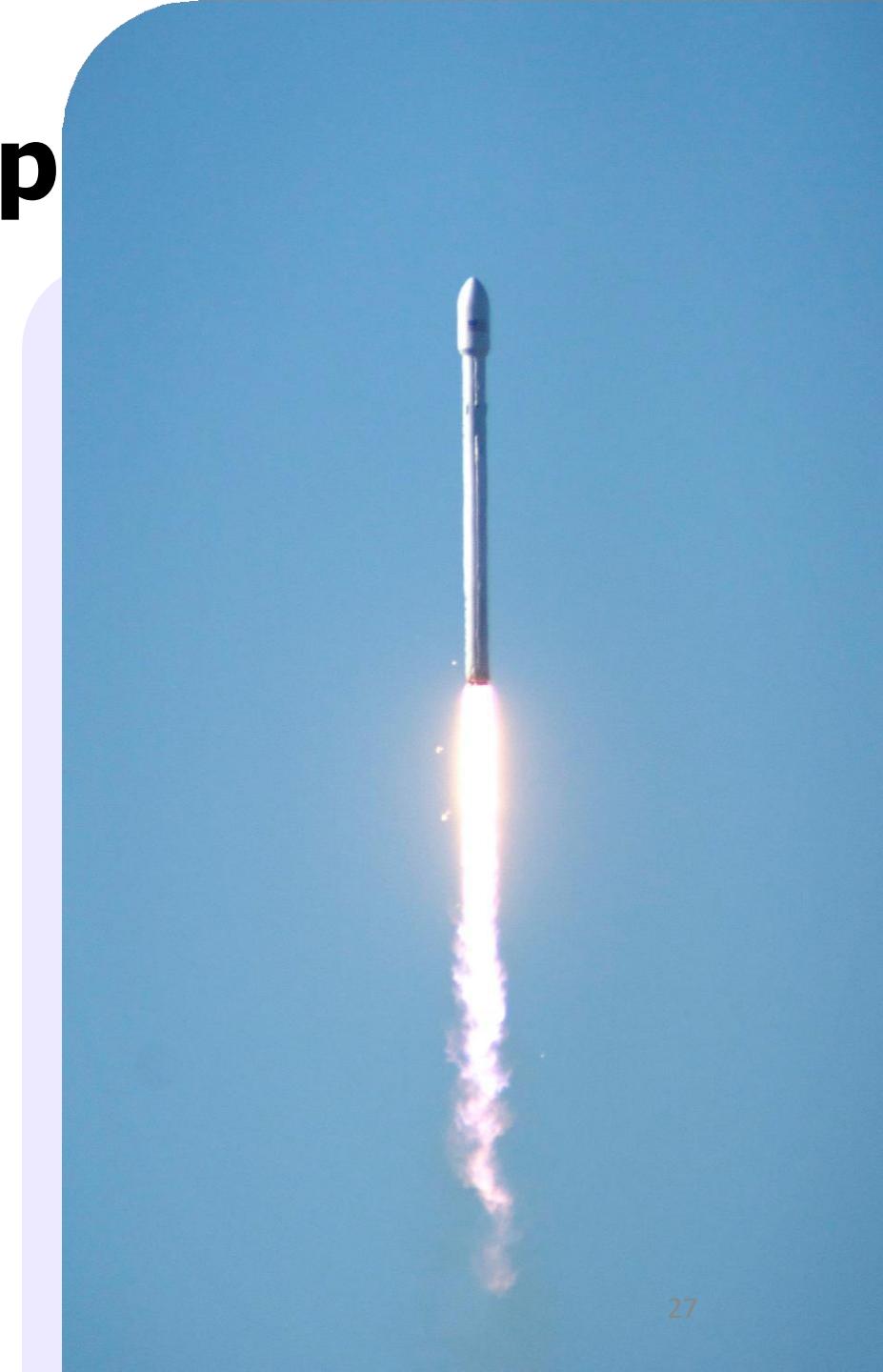
- Showing month, date, booster version, launch site and landing outcome

```
%sql SELECT substr(Date,4,2) as month, DATE, BOOSTER_VERSION, LAUNCH_SITE, [Landing _Outcome] \
FROM SPACEXTBL \
where [Landing _Outcome] = 'Failure (drone ship)' and substr(Date,7,4)='2015';
```

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

```
Done.
```

month	Date	Booster_Version	Launch_Site	Landing_Outcome
01	10-01-2015	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
04	14-04-2015	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)



Count of Successful Landings

Ranked Descending

- Count of landing outcomes between 2010-06-04 and 2017-03-20 in descending order

```
%sql SELECT [Landing _Outcome], count(*) as count_outcomes \
FROM SPACEXTBL \
WHERE DATE between '04-06-2010' and '20-03-2017' group by [Landing _Outcome] order by count_outcomes DESC;
```

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

```
Done.
```

Landing _Outcome	count_outcomes
Success	20
No attempt	10
Success (drone ship)	8
Success (ground pad)	6
Failure (drone ship)	4
Failure	3
Controlled (ocean)	3
Failure (parachute)	2
No attempt	1



Launch Site Analysis



Launch Sites

With Markers

- **Near Equator:** the closer the launch site to the equator, the **easier** it is **to launch** to equatorial orbit, and the more help you get from Earth's rotation for a prograde orbit. Rockets launched from sites near the equator get an **additional natural boost** - due to the rotational speed of earth - that **helps save the cost** of putting in extra fuel and boosters.



Launch Outcomes

At Each Launch Site

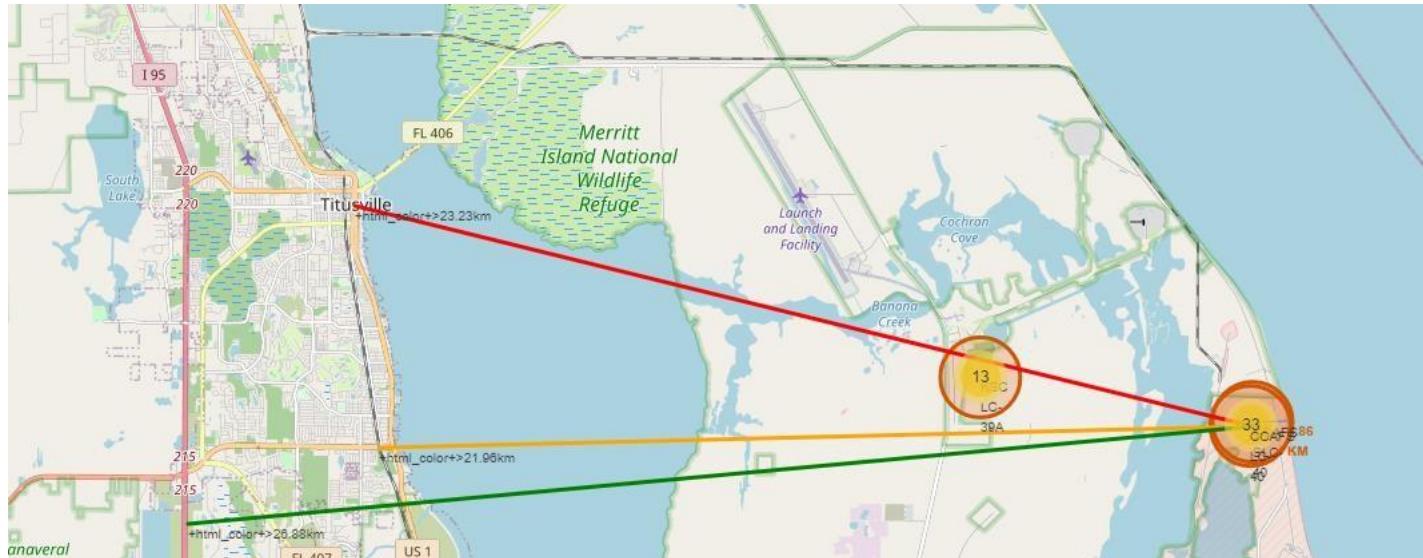
- **Outcomes:**
- **Green** markers for successful launches
- **Red** markers for unsuccessful launches
- Launch site **CCAFS SLC-40** has a **3/7 success rate (42.9%)**



Distance to Proximities

CCAFS SLC-40

- **.86 km** from nearest coastline
- **21.96 km** from nearest railway
- **23.23 km** from nearest city
- **26.88 km** from nearest highway



2023

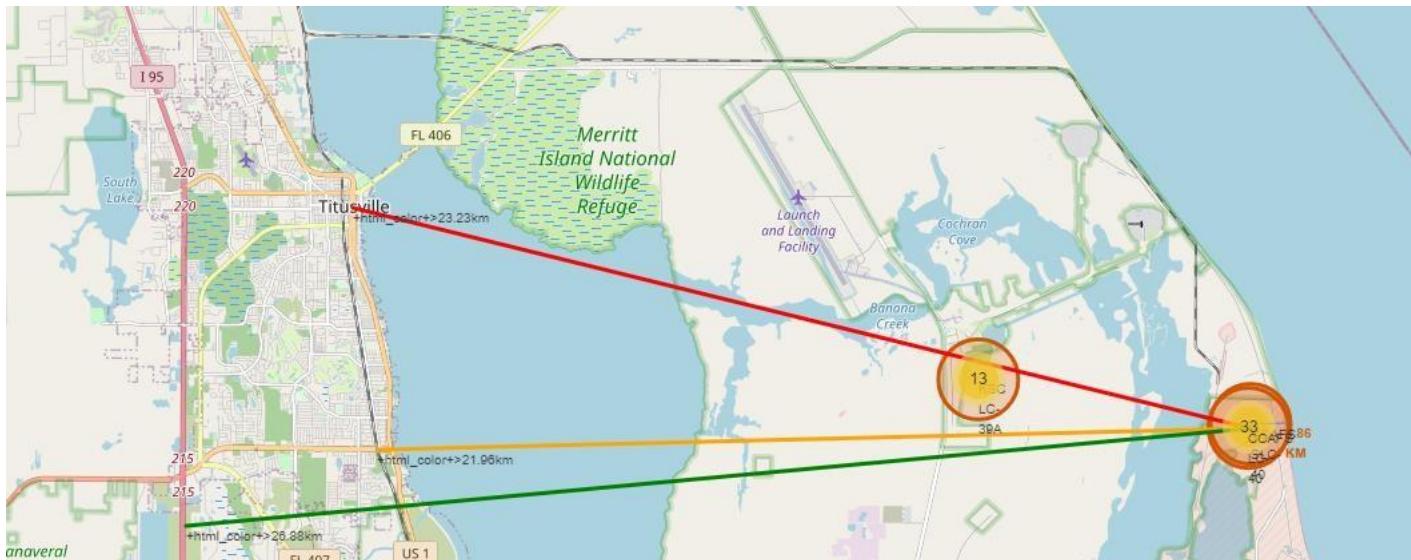


32

Distance to Proximities

CCAFS SLC-40

- **Coasts:** help ensure that spent stages dropped along the launch path or failed launches don't fall on people or property.
- **Safety / Security:** needs to be an exclusion zone around the launch site to keep unauthorized people away and keep people safe.
- **Transportation/Infrastructure and Cities:** need to be away from anything a failed launch can damage, but still close enough to roads/rails/docks to be able to bring people and material to or from it in support of launch activities.



PLOTLY DASHBOARD

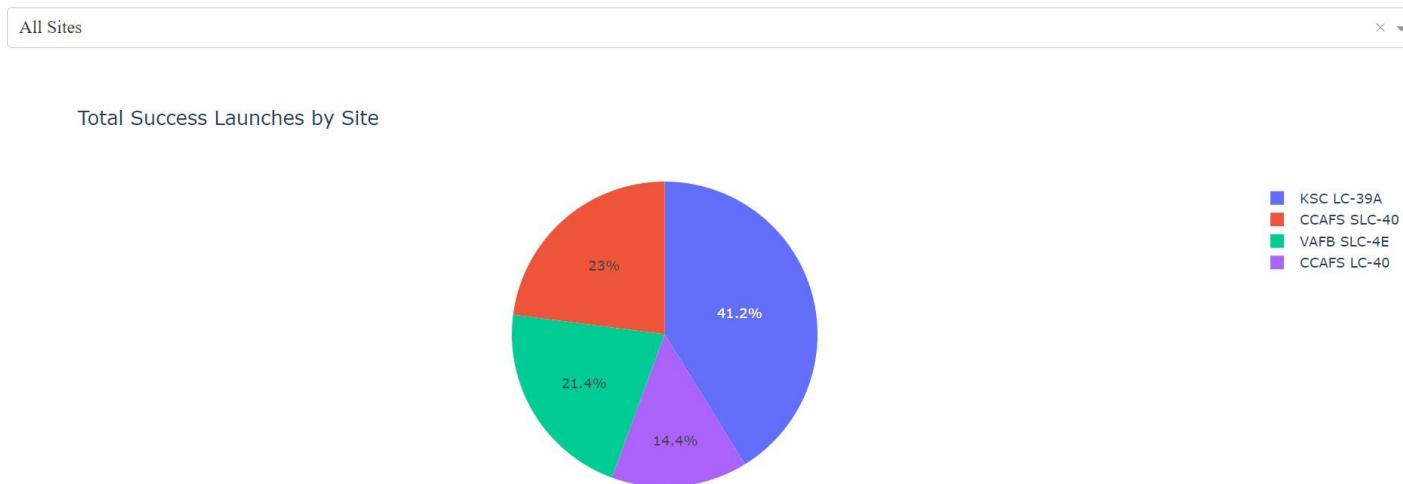


Launch Success by Site

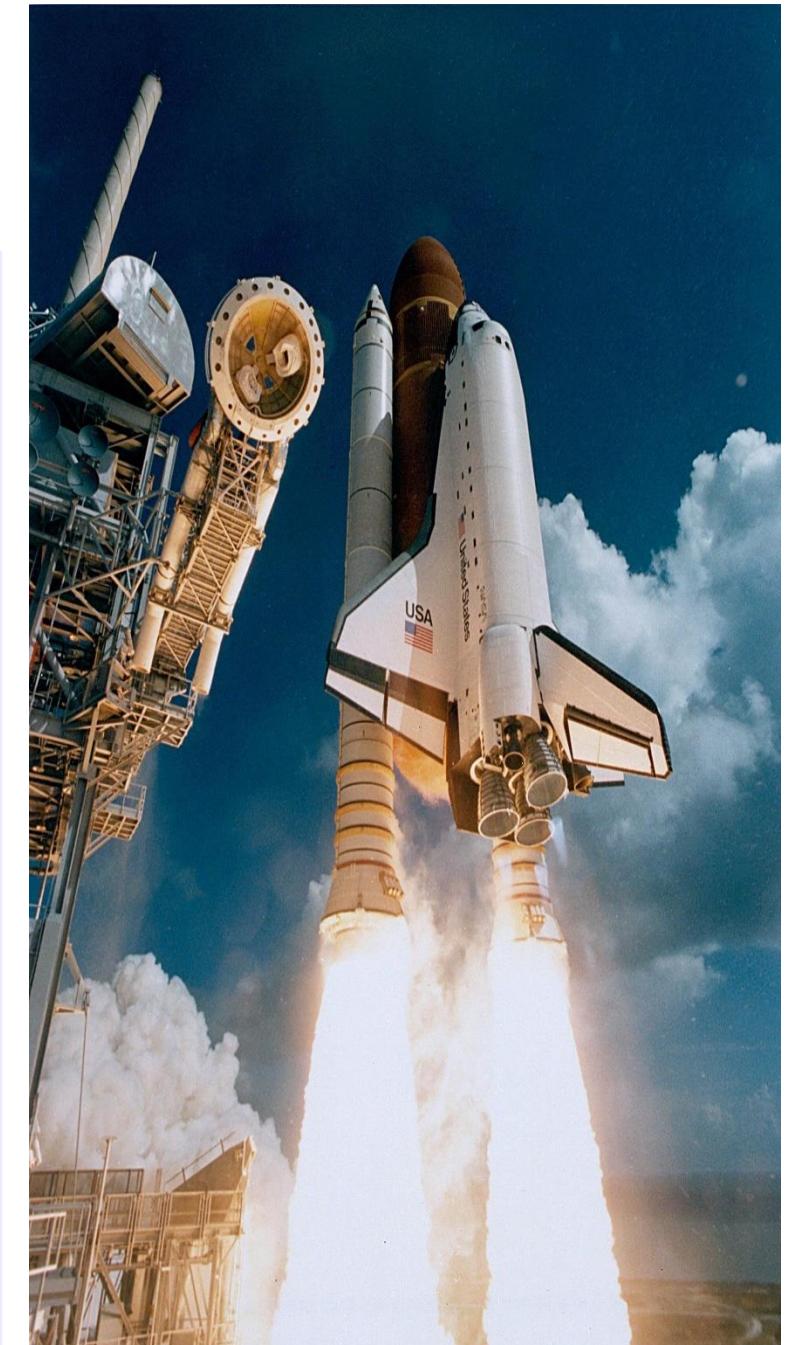
Success as Percent of Total

- **KSC LC-39A** has the **most successful launches** amongst launch sites (**41.2%**)

SpaceX Launch Records Dashboard



2023

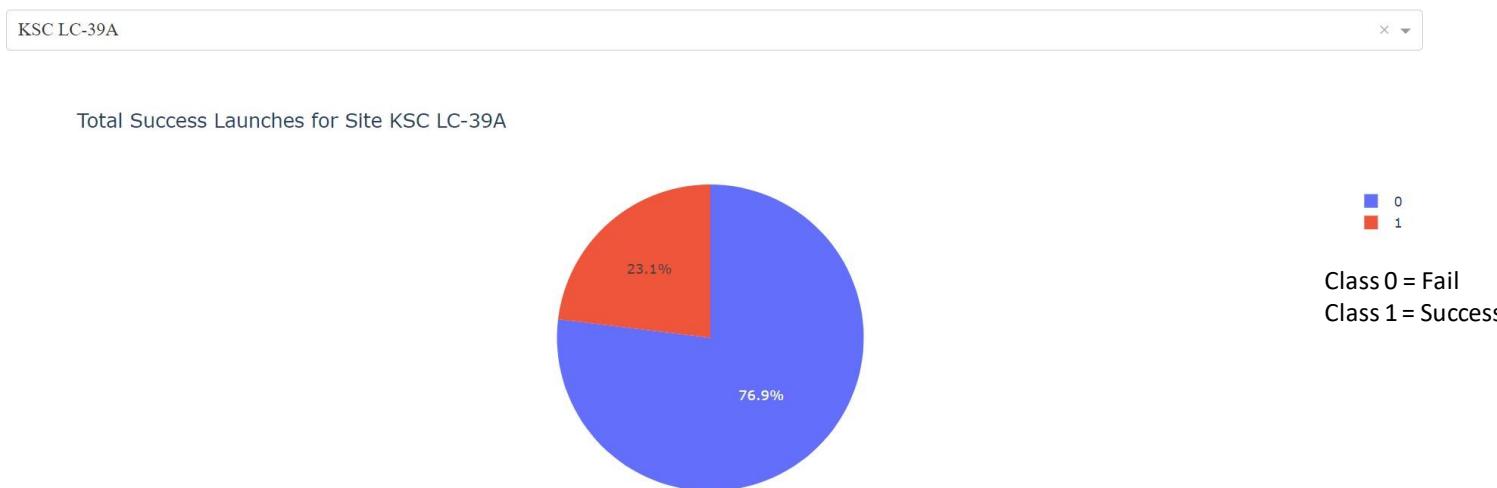


Launch Success (KSC LC-29A)

Success as Percent of Total

- **KSC LC-39A** has the **highest success rate** amongst launch sites (**76.9%**)
- 10 successful launches and 3 failed launches

SpaceX Launch Records Dashboard



Payload Mass and Success

By Booster Version

- **Payloads between 2,000 kg and 5,000 kg have the highest success rate**
- 1 indicating successful outcome and 0 indicating an unsuccessful outcome



Predictive Analytics



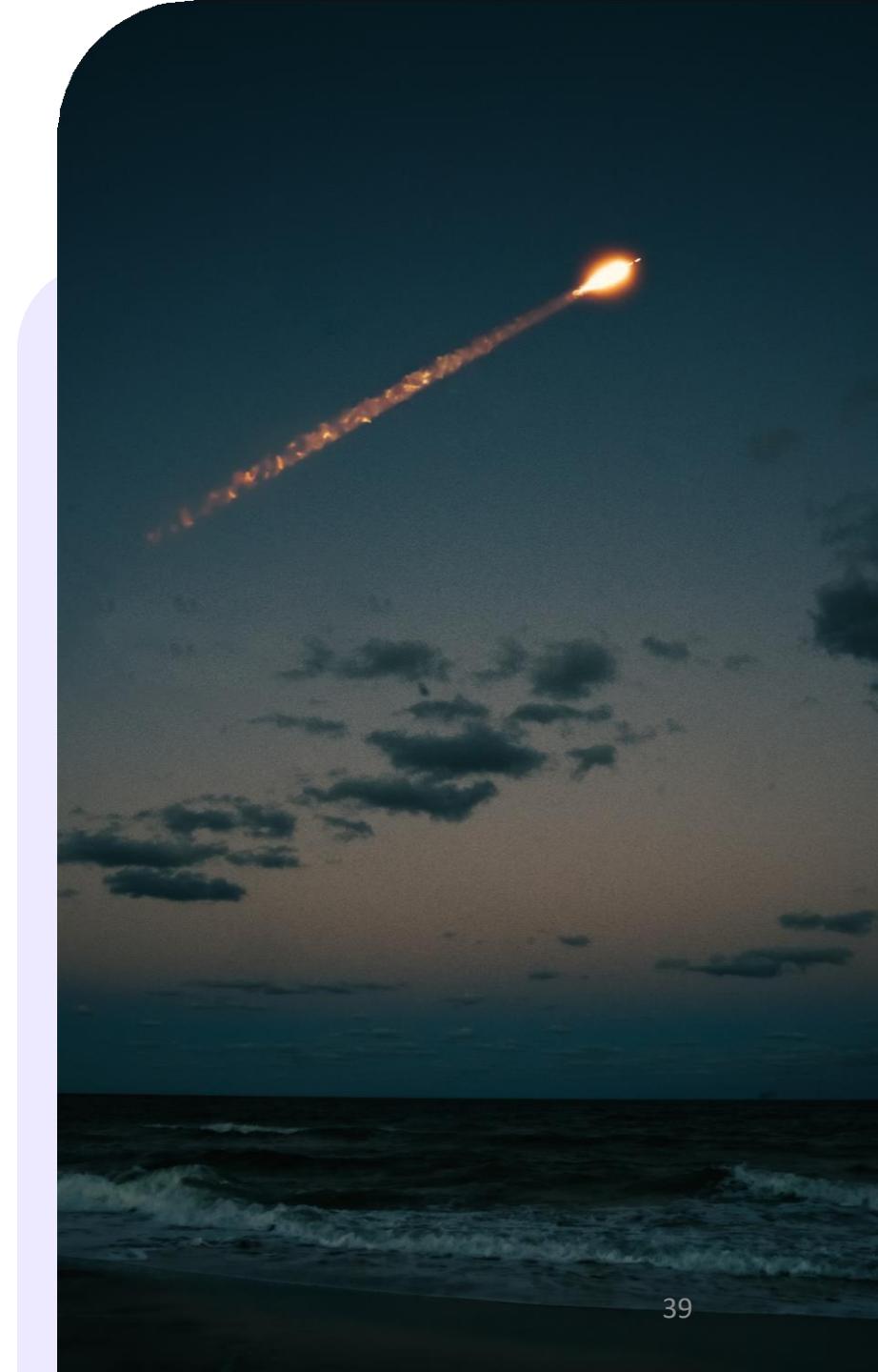
Classification

Accuracy

- All the **models** performed at about the same level and had the **same scores** and **accuracy**. This is likely due to the **small dataset**. The **Decision Tree model slightly outperformed** the rest when looking at `.best_score_`
- `.best_score_` is the average of all cv folds for a single combination of the parameters

	LogReg	SVM	Tree	KNN
Jaccard_Score	0.800000	0.800000	0.800000	0.800000
F1_Score	0.888889	0.888889	0.888889	0.888889
Accuracy	0.833333	0.833333	0.833333	0.833333

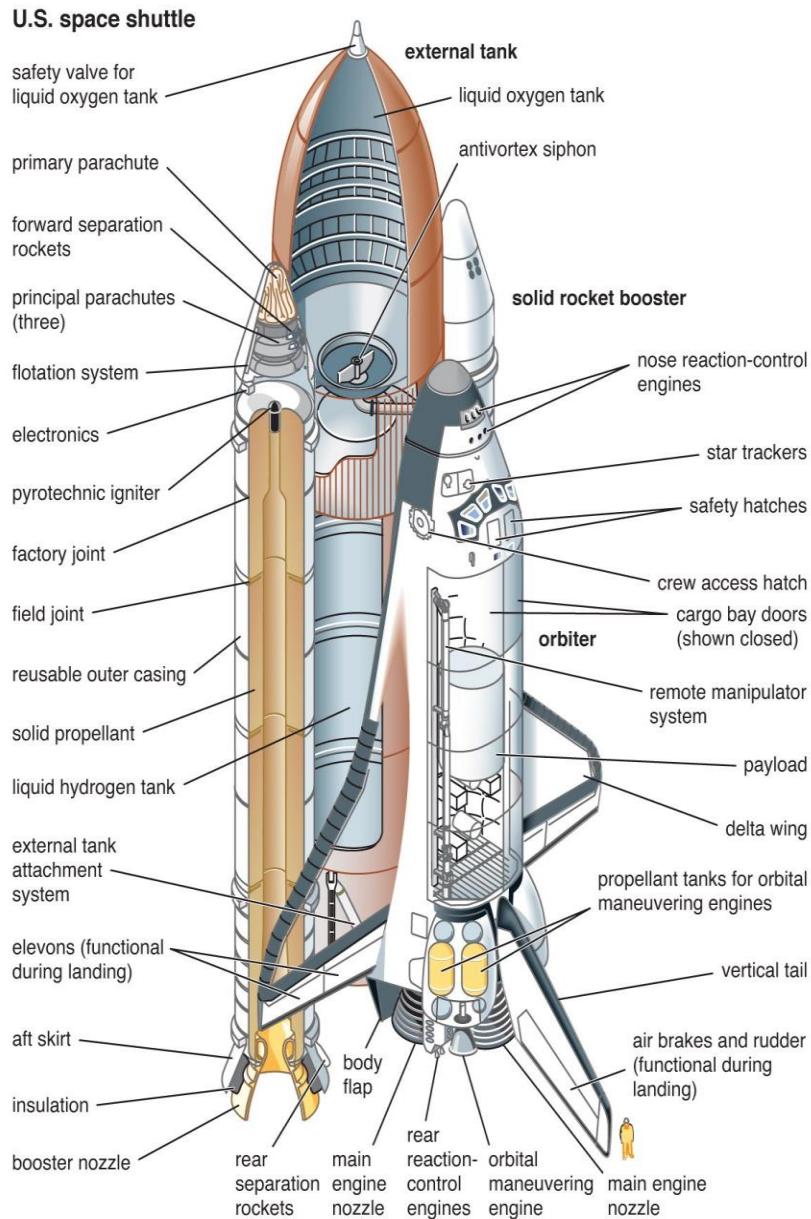
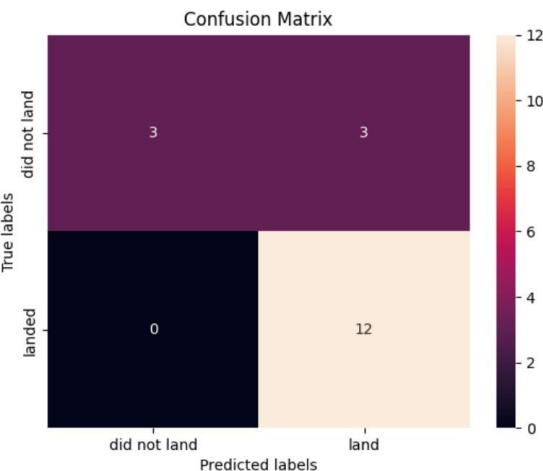
```
: models = {'KNeighbors':knn_cv.best_score_,  
           'DecisionTree':tree_cv.best_score_,  
           'LogisticRegression':logreg_cv.best_score_,  
           'SupportVector': svm_cv.best_score_}  
  
bestalgorithm = max(models, key=models.get)  
print('Best model is', bestalgorithm, 'with a score of', models[bestalgorithm])  
if bestalgorithm == 'DecisionTree':  
    print('Best params is :', tree_cv.best_params_)  
if bestalgorithm == 'KNeighbors':  
    print('Best params is :', knn_cv.best_params_)  
if bestalgorithm == 'LogisticRegression':  
    print('Best params is :', logreg_cv.best_params_)  
if bestalgorithm == 'SupportVector':  
    print('Best params is :', svm_cv.best_params_)  
  
Best model is DecisionTree with a score of 0.9017857142857142  
Best params is : {'criterion': 'gini', 'max_depth': 16, 'max_features': 'auto', 'min_samples_leaf': 4, 'min_samples_split': 10, 'splitter': 'random'}
```



Confusion Matrices

Performance Summary

- A confusion matrix summarizes the performance of a classification algorithm
- All the confusion matrices were identical
- The fact that there are false positives (Type 1 error) is not good
- Confusion Matrix Outputs:
 - 12 True positive
 - 3 True negative
 - **3 False positive**
 - 0 False Negative
- **Precision** = $TP / (TP + FP)$
 - $12 / 15 = .80$
- **Recall** = $TP / (TP + FN)$
 - $12 / 12 = 1$
- **F1 Score** = $2 * (\text{Precision} * \text{Recall}) / (\text{Precision} + \text{Recall})$
 - $2 * (.8 * 1) / (.8 + 1) = .89$
- **Accuracy** = $(TP + TN) / (TP + TN + FP + FN) = .833$



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Conclusion

Research

- **Model Performance:** The models performed similarly on the test set with the decision tree model slightly outperforming
- **Equator:** Most of the launch sites are near the equator for an additional natural boost - due to the rotational speed of earth - which helps save the cost of putting in extra fuel and boosters
- **Coast:** All the launch sites are close to the coast
- **Launch Success:** Increases over time
- **KSC LC-39A:** Has the highest success rate among launch sites. Has a 100% success rate for launches less than 5,500 kg
- **Orbits:** ES-L1, GEO, HEO, and SSO have a 100% success rate
- **Payload Mass:** Across all launch sites, the higher the payload mass (kg), the higher the success rate

Conclusion

Considerations

Dataset Size: Expanding the dataset could enhance the predictive analytics results, helping determine if the findings are applicable to a broader scope.

Feature Analysis and PCA: Conducting further feature analysis or applying Principal Component Analysis (PCA) may improve model accuracy.

XGBoost Model: Although not utilized in this study, XGBoost is a robust model worth exploring to see if it performs better than the current classification models.