



IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

David O'Connor
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Outline

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

Executive Summary

- Summary of methodologies:
 - Data Collection: SQL, Web API & Web Scraping
 - Data Wrangling: Cleaning and Preparation
 - Data Visualisation
 - Exploratory Data Analysis
 - Machine Learning Methods
- Summary of all results
 - EDA
 - Analytics
 - ML results

Introduction

- Project background and context:
 - Create a predictive model to assess the outcomes of first stage landing outcomes for Space X rocket launches
- Problems you want to find answers:
 - What method produces most accurate prediction of launch outcome

Section 1

Methodology

Methodology

Executive Summary

- Data collection methodology:
 - Python Web API @ <https://api.spacexdata.com/v4/rockets>
 - Python Web scraping @ https://en.wikipedia.org/w/index.php?title=List_of_Falcon_9_and_Falcon_Heavy_launches&oldid=1027686922
- Perform data wrangling
 - Check columns for percentage of missing data
 - Check data types
 - Assess number of launches
 - Create a landing class column based on successful or unsuccessful landing outcome.

Methodology

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Data Operations: Normalize, test/train split, train and evaluate using k-means, logistic regression, decision trees, and kNN

Data Collection

- Describe how data sets were collected:

Method 1: SpaceX API

- API get request -> pd json normalize -> Data Wrangling

Method 2: Wikipedia Webscraping

- http get request -> BeautifulSoup html parser -> Data Wrangling

Data Collection – SpaceX API

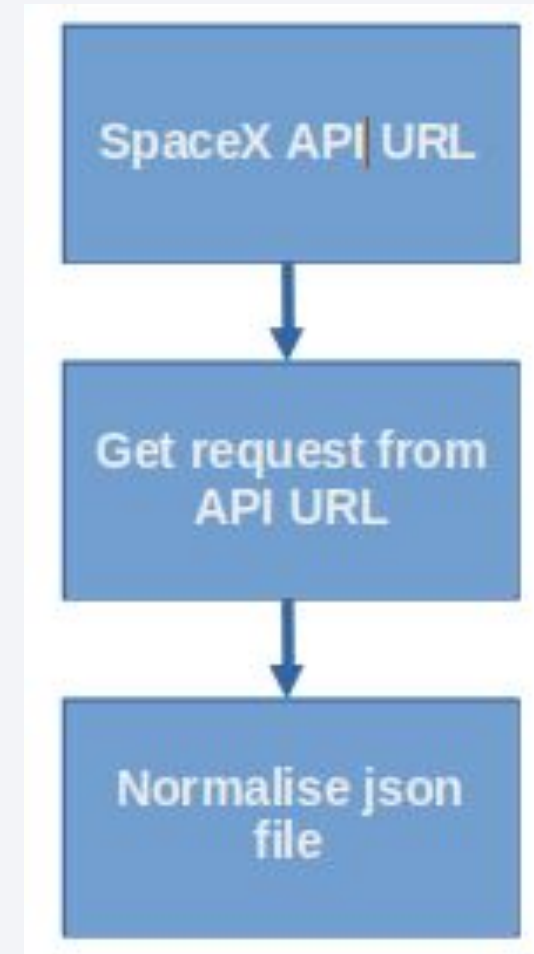
```
# Requesting data from API
```

```
spacex_url="https://api.spacexdata.com/v4/launches/past"
```

```
response = requests.get(spacex_url)
```

```
data = pd.json_normalize(response.json())
```

- https://github.com/skoopsy/ibm-ds-course/blob/main/a1_web_API_requests.ipynb



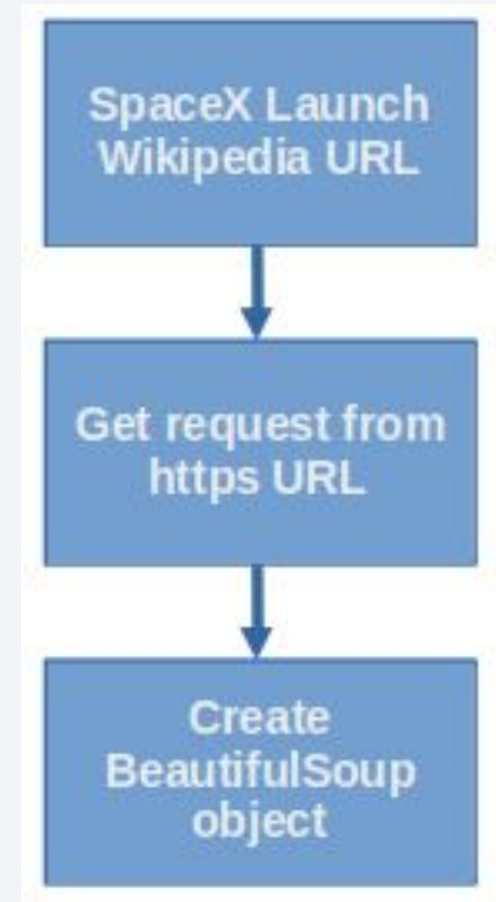
Data Collection - Scraping

```
static_url =  
"https://en.wikipedia.org/w/index.php?title=List_of  
_Falcon_9_and_Falcon_Heavy_launches&oldid=  
1027686922"
```

```
page=requests.get(static_url)
```

```
soup=BeautifulSoup(page.text, 'html.parser')
```

https://github.com/skoopsy/ibm-ds-course/blob/main/a2_web_scraping_wikipedia.ipynb



Data Wrangling

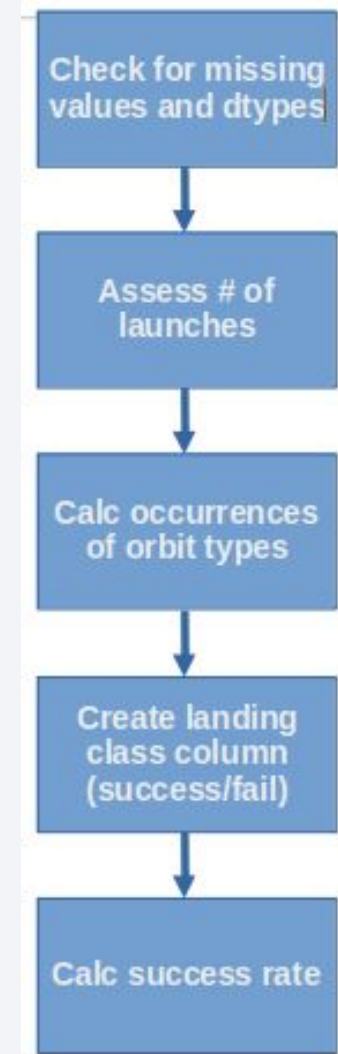
- Describe how data were processed:

- See flow chart

- Code:

```
df=pd.read_csv("https://cf-courses-data.s3.us.c
df.isnull().sum() / df.count()*100
df.dtypes
df.LaunchSite.value_counts()
df.Orbit.value_counts()
landing_outcomes = df.Outcome.value_counts()
landing_class = []
for i, outcome in enumerate(df.Outcome):
    if outcome in bad_outcomes:
        landing_class.append(0)
    else:
        landing_class.append(1)
df["Class"].mean().round(2)
```

- https://github.com/skoopsy/ibm-ds-course/blob/main/a3_data_wrangling-exploratory_data_analysis.ipynb



EDA with Data Visualization

Summarize what charts were plotted and why you used those charts:

- Flight Number vs Launch Site: Chronology of launch site usage
- Payload vs Launch Site: Payload capacity of each launch site
- Success rate vs Orbit type: Are certain orbit types more successful
- Flight Number vs Orbit Type: Chronology of orbit type launches
- Payload vs Orbit Type: Payload capacity of orbit types
- Success Rate vs Year: Is SpaceX becoming better at landing rockets?
- https://github.com/skoopsy/ibm-ds-course/blob/main/a5_eda_visualisation_seaborn.ipynb

EDA with SQL

- Using bullet point format, summarize the SQL queries you performed

```
df = pd.read_csv("https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/l  
df.to_sql("SPACEXTBL", con, if_exists='replace', index=False, method="multi")  
%sql create table SPACEXTABLE as select * from SPACEXTBL where Date is not null  
%sql SELECT DISTINCT LAUNCH_SITE from SPACEXTABLE;  
%sql SELECT LAUNCH_SITE from SPACEXTBL where (LAUNCH_SITE) LIKE 'CCA%' LIMIT 5;  
%sql select sum(PAYLOAD_MASS_KG) from SPACEXTBL where customer='NASA (CRS)';  
%sql SELECT AVG(PAYLOAD_MASS_KG) from SPACEXTBL WHERE BOOSTER_VERSION='F9 v1.1';  
%sql SELECT MIN(DATE) FROM SPACEXTBL WHERE LANDING_OUTCOME LIKE 'Success (ground pad)';  
%sql SELECT BOOSTER_VERSION from SPACEXTBL where LANDING_OUTCOME='Success (drone ship)' and PAYLOAD_MASS_KG bet  
%sql select count(MISSION_OUTCOME), mission_outcome from SPACEXTBL GROUP BY MISSION_OUTCOME like "Success%";  
%sql select booster_version from SPACEXTBL where payload_mass_kg =(select MAX(payload_mass_kg) from SPACEXTBL)  
%sql select LANDING_OUTCOME, booster_version, launch_site, date from SPACEXTBL where date like "2015%" and LANDIN  
%sql select LANDING_OUTCOME, count(*) from SPACEXTBL where date between "2010-06-04" and "2017-03-20"
```

- https://github.com/skoopsy/ibm-ds-course/blob/main/a4_sql_exploratory_data_analysis.ipynb

Build an Interactive Map with Folium

- Summarize what map objects such as markers, circles, lines, etc. you created and added to a folium map
 - Mark all launch sites on map
 - use circle to mark the launch site with popup
 - create marker clusters for different zoom levels
 - successful launch = green marker
 - failed launch = red marker
- Explain why you added those objects
 - Objects added to create a visual representation of where successful and failed launch sites occur to see if there is any visual indication of predicting launch outcome
- https://github.com/skoopsy/ibm-ds-course/blob/main/a6_folium.ipynb

Build a Dashboard with Plotly Dash

- Summarize what plots/graphs and interactions you have added to a dashboard
 - Pie chart total successful launches
 - Pie chart for failed vs successful launches for specific launch site
 - Slider to select payload ranges
 - Payload vs Launch success
- Explain why you added those plots and interactions
 - To uncover if payload and launch site impact launch success
- https://github.com/skoopsy/ibm-ds-course/blob/main/a7_plotly_dash.py

Predictive Analysis (Classification)

- Summarize how you built, evaluated, improved, and found the best performing classification model



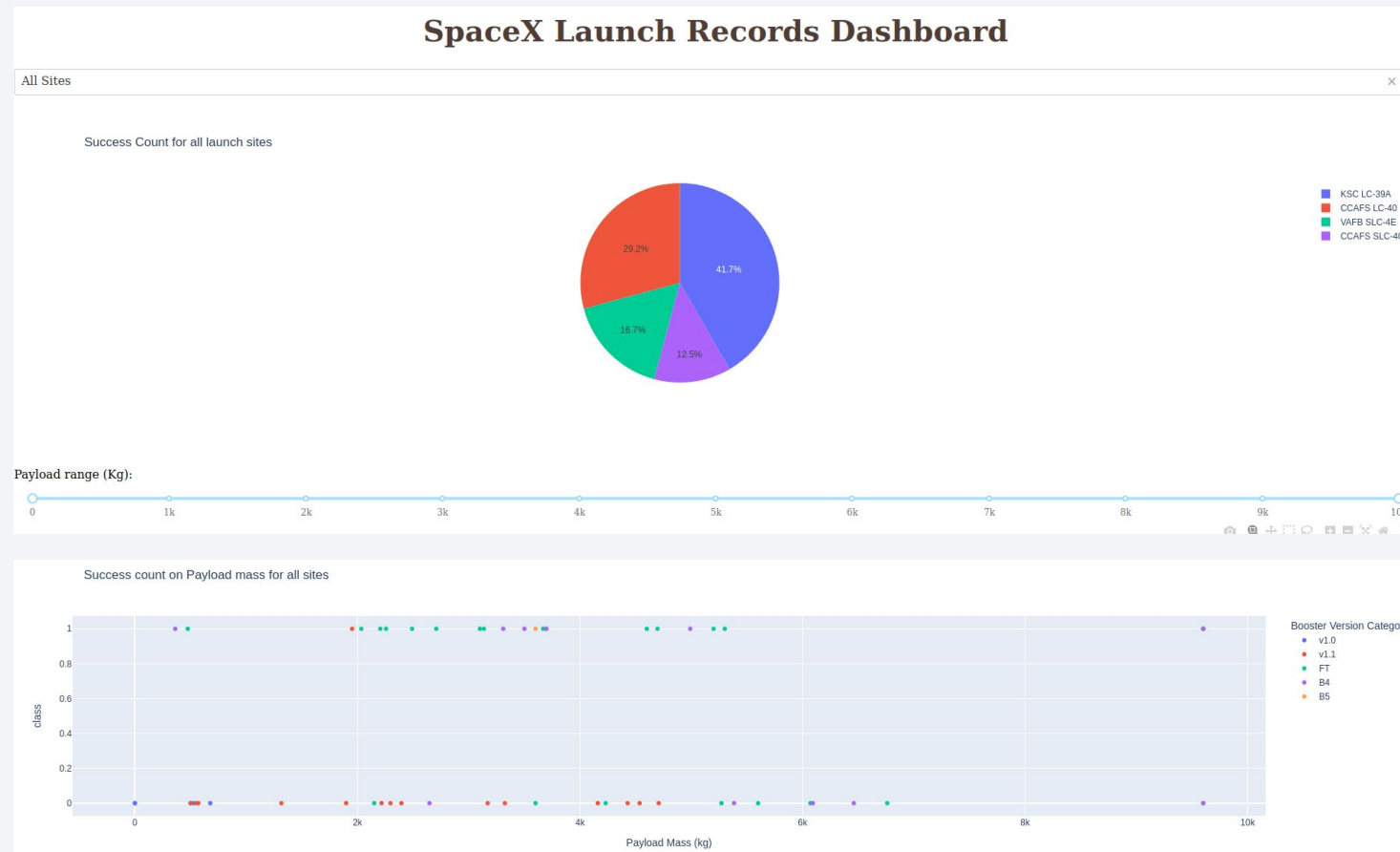
- The best model was found by comparing accuracy scores and looking for MAX
- https://github.com/skoopsy/ibm-ds-course/blob/main/a8_ml.ipynb

Results

- Exploratory data analysis results:
 - 4 launch sites
 - total payload = 45596 kg
 - avg payload = 2928.4kg
 - first successful landing outcome in 2015
 - landing success rate improved over time

Results

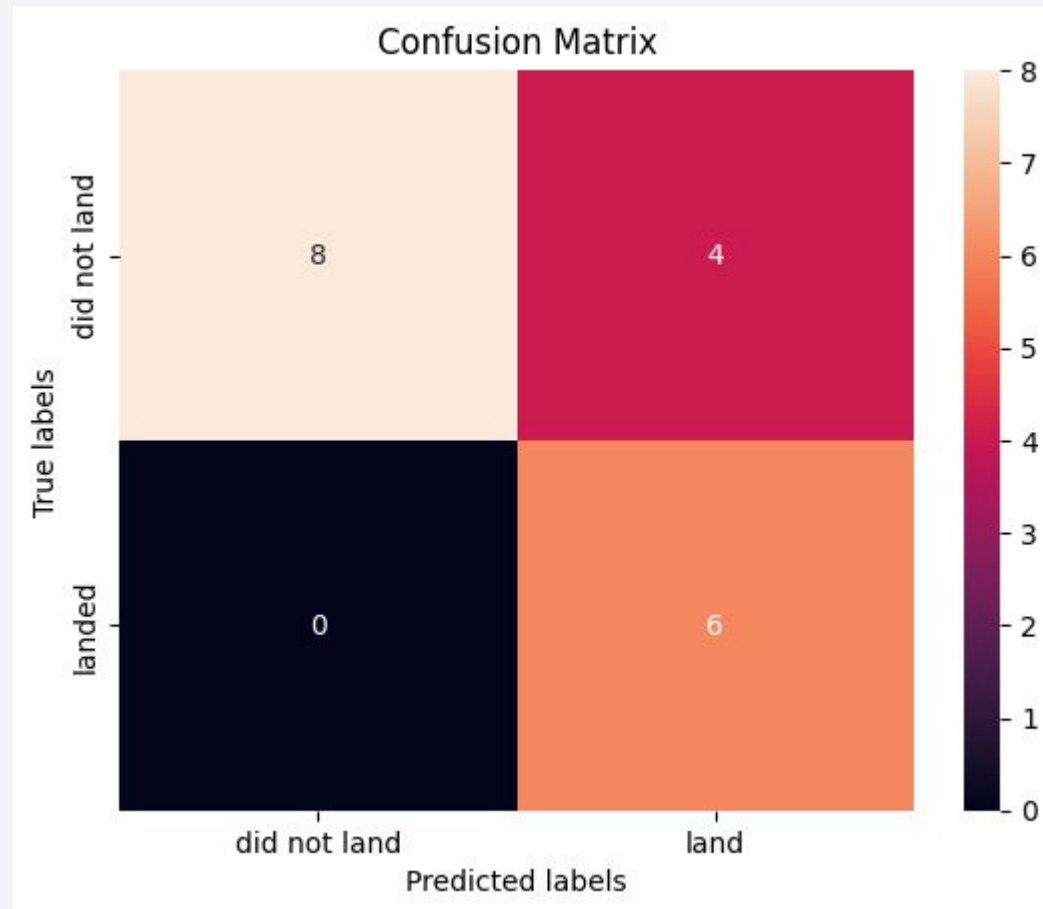
- Interactive analytics demo in screenshots (Plotly Dash):



Results

- Predictive analysis results:

Log Reg

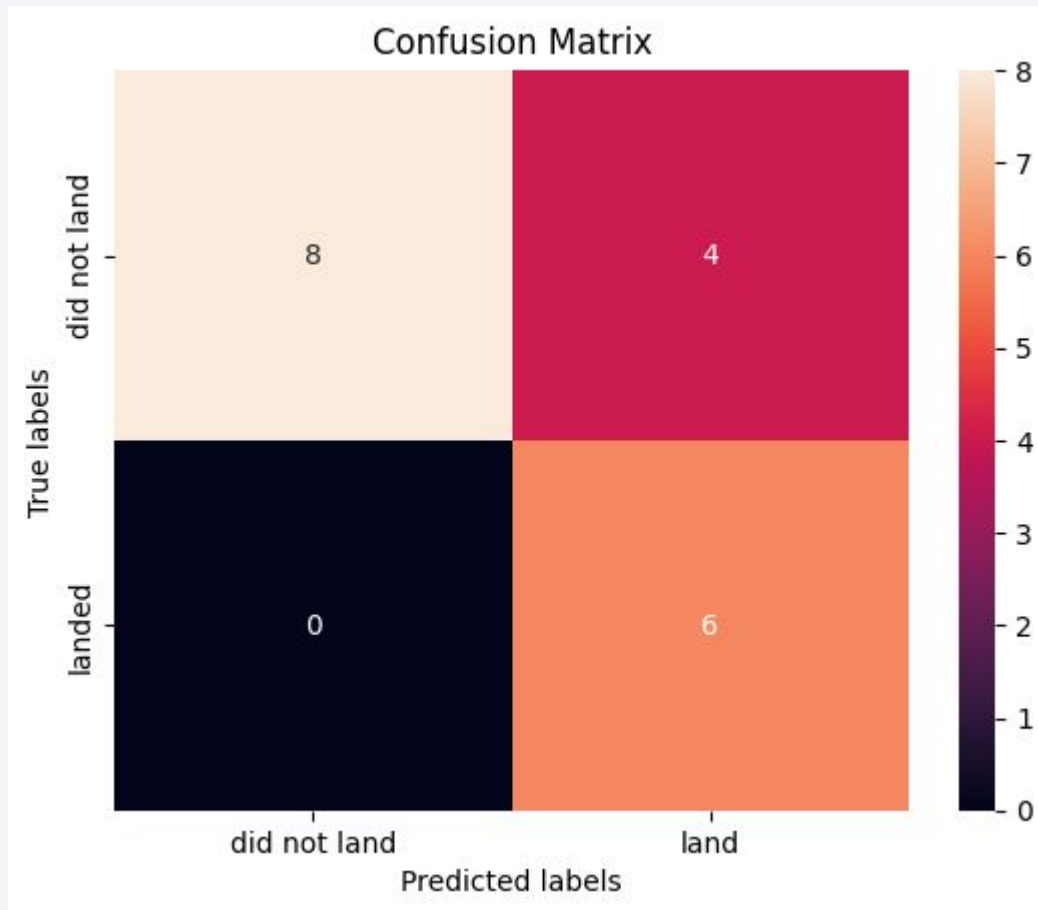


SVM

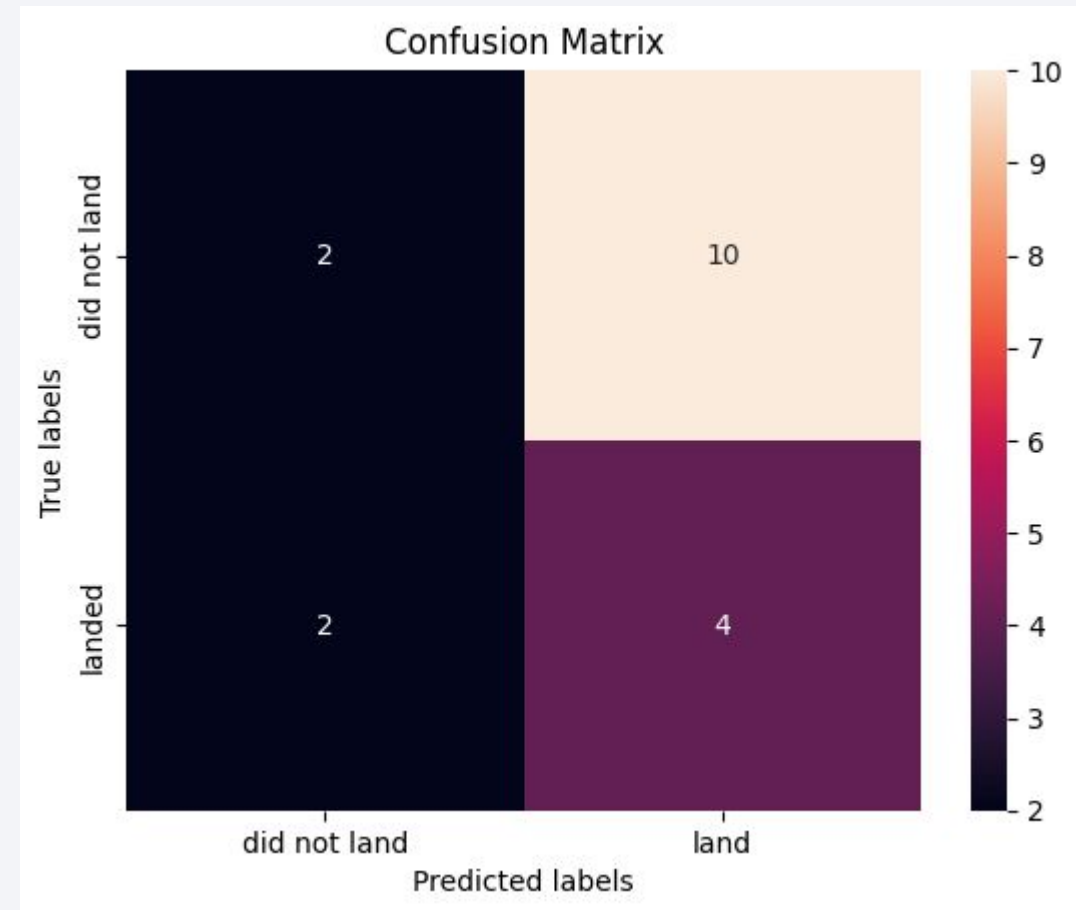
Results

- Predictive analysis results:

Tree



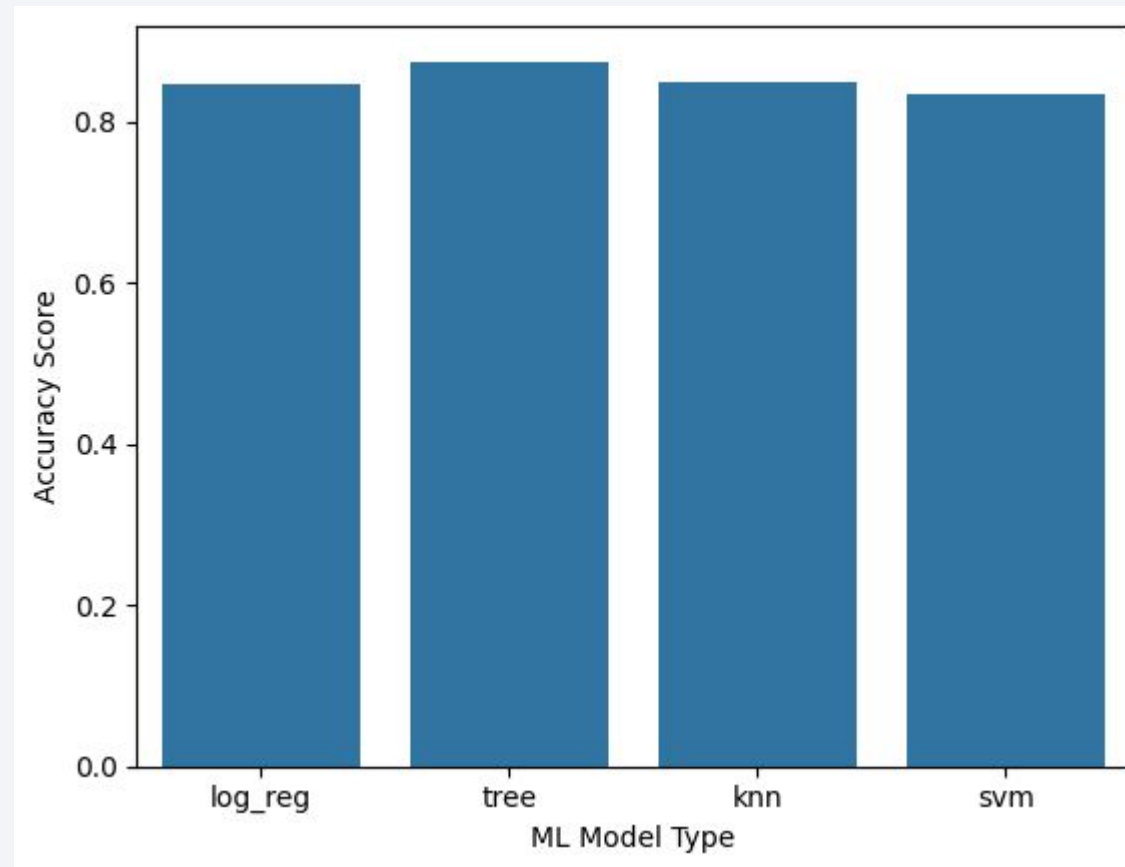
KNN



Results

- Predictive analysis results:

Accuracy Score comparison



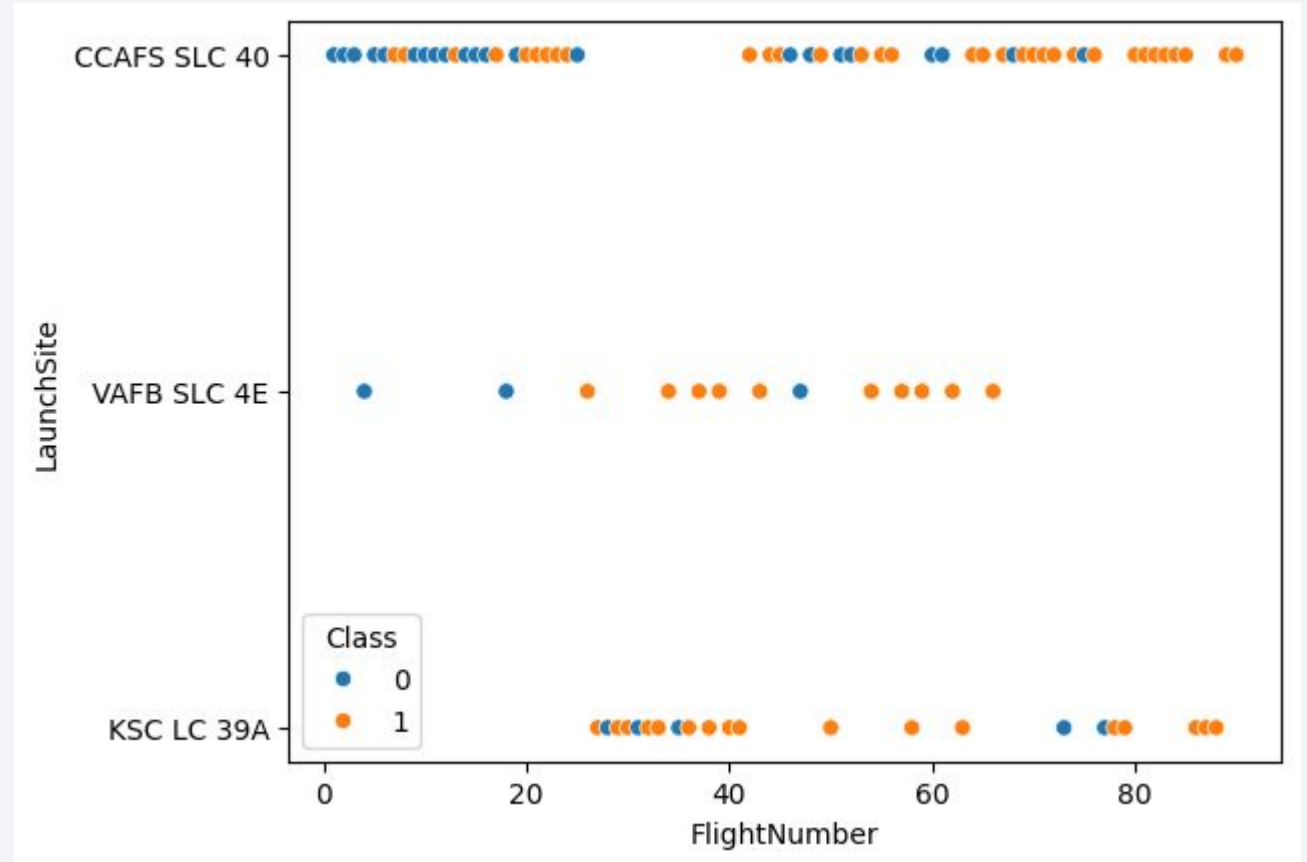
The background of the slide is an abstract composition. It features a solid blue area on the left side, which transitions into a dynamic pattern of diagonal streaks in shades of blue, red, and teal on the right. These streaks have a textured, almost woven appearance. Overlaid on this pattern is a faint, light blue grid that creates a sense of depth and structure.

Section 2

Insights drawn from EDA

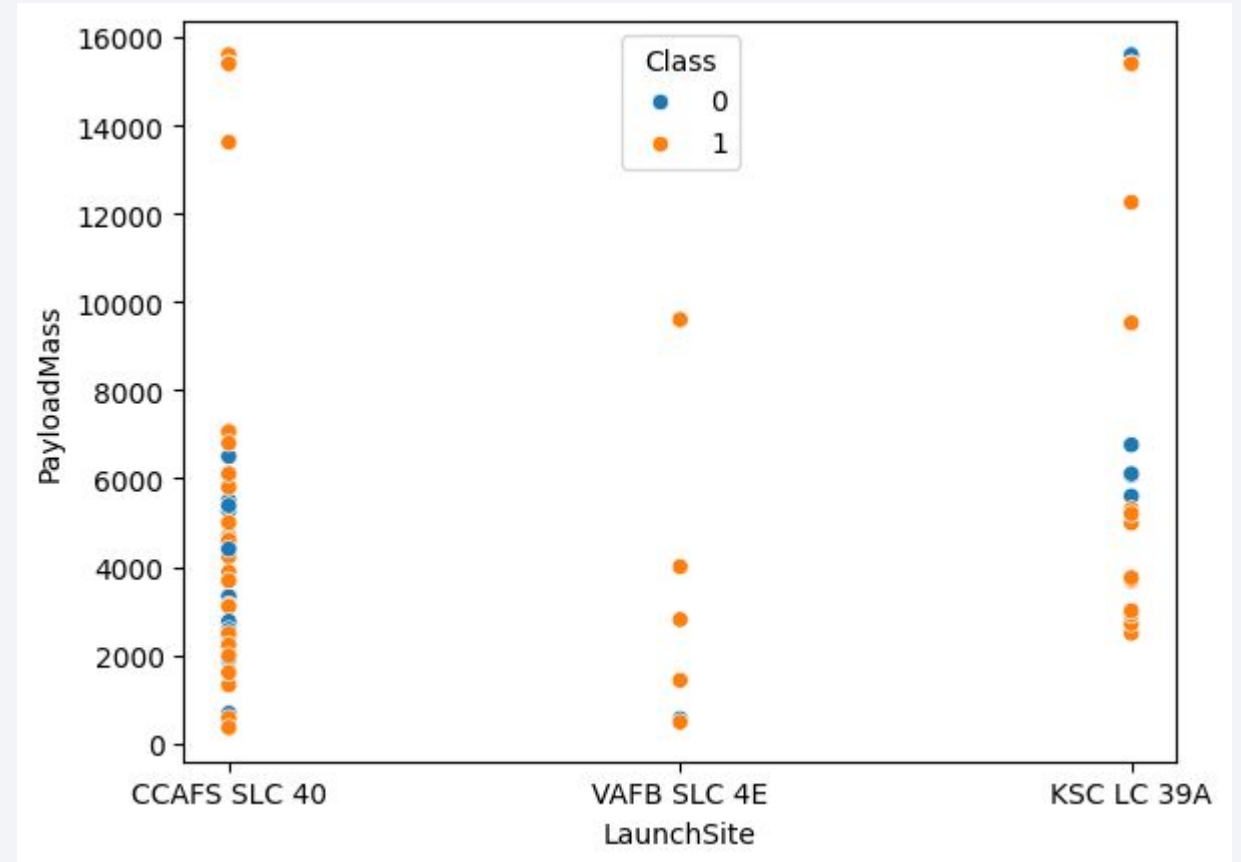
Flight Number vs. Launch Site

- higher flight numbers are more successful
- unclear if due to launch site or improvement of company processes



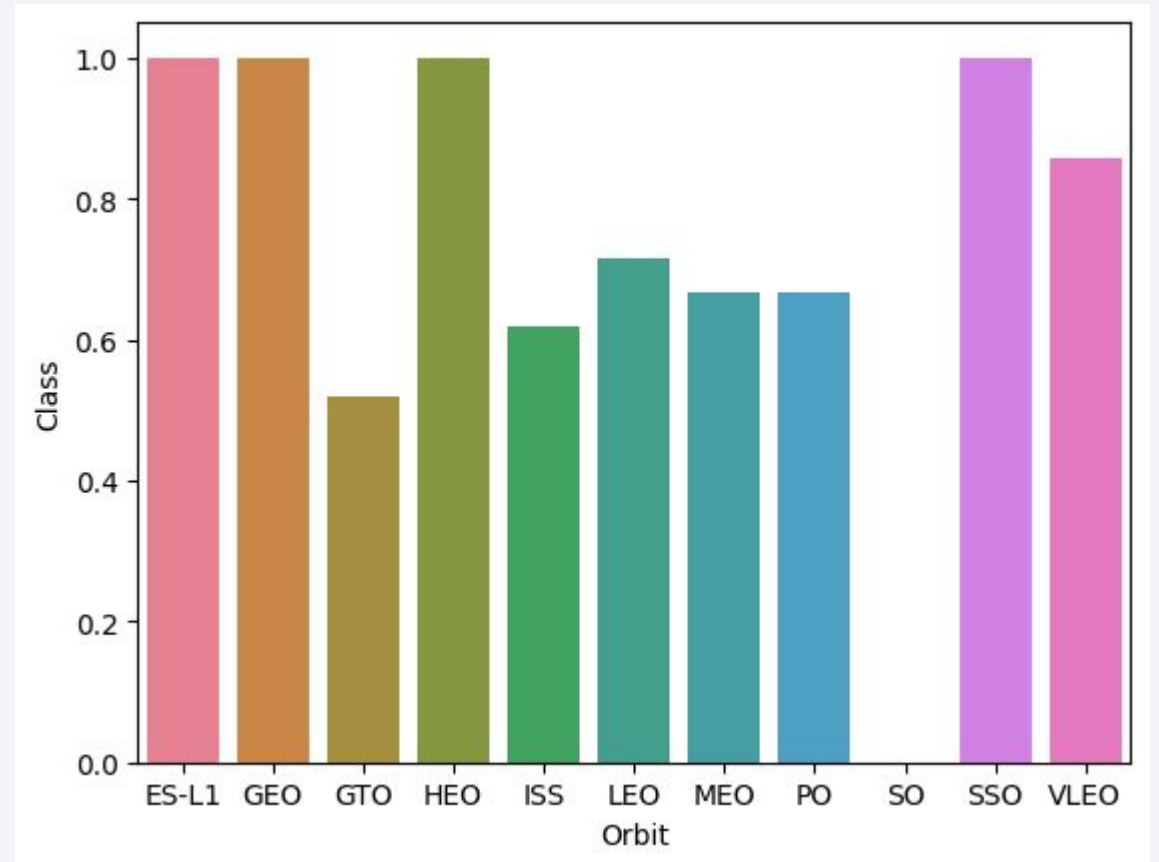
Payload vs. Launch Site

- Launch site VAFB SLC 4E has a high success rate across all payload sizes
- CCAFS SLC 40 has no failures at high payloads
- ksc lc 39a shows failures across the full payload range



Success Rate vs. Orbit Type

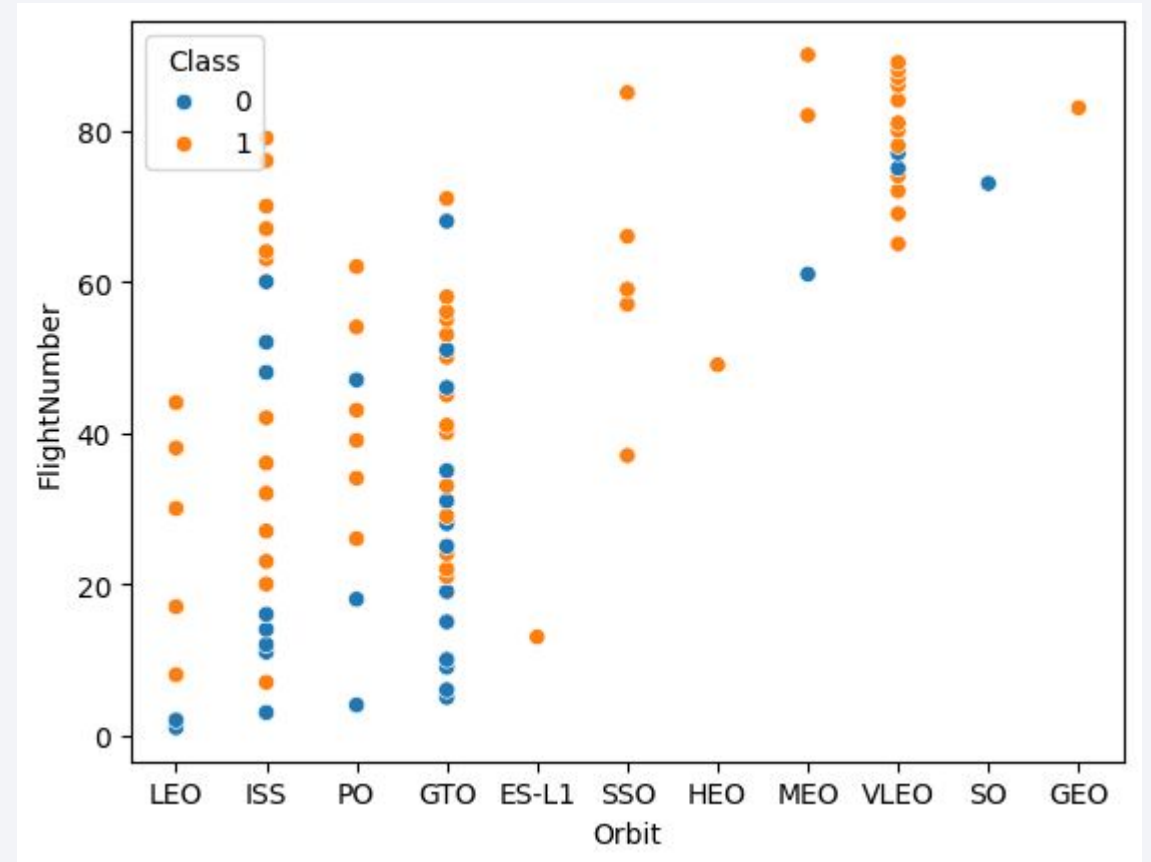
- Orbit types: ES-L1, GEO, HEO, SSO have a 100% success rate



Flight Number vs. Orbit Type

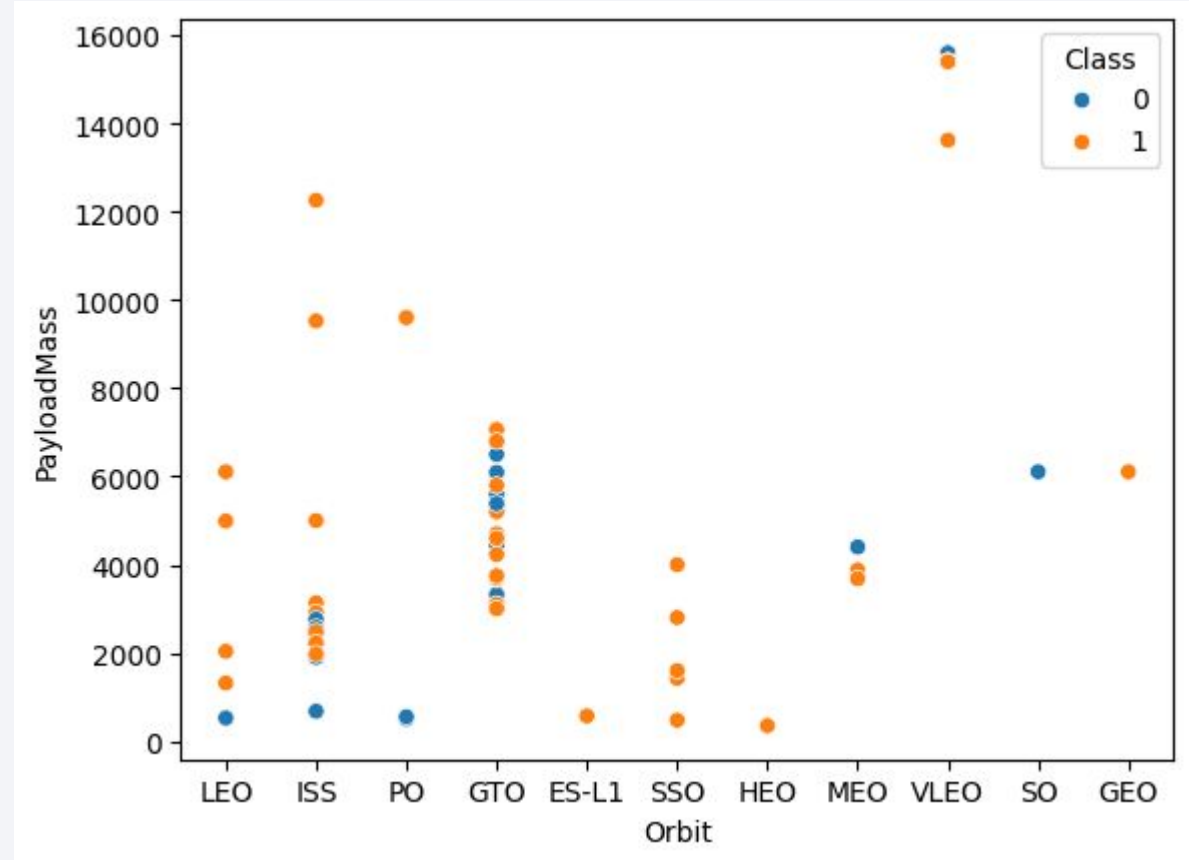
Orbits SSO to GEO were later in the launch chronology.

Less failures at higher flight numbers



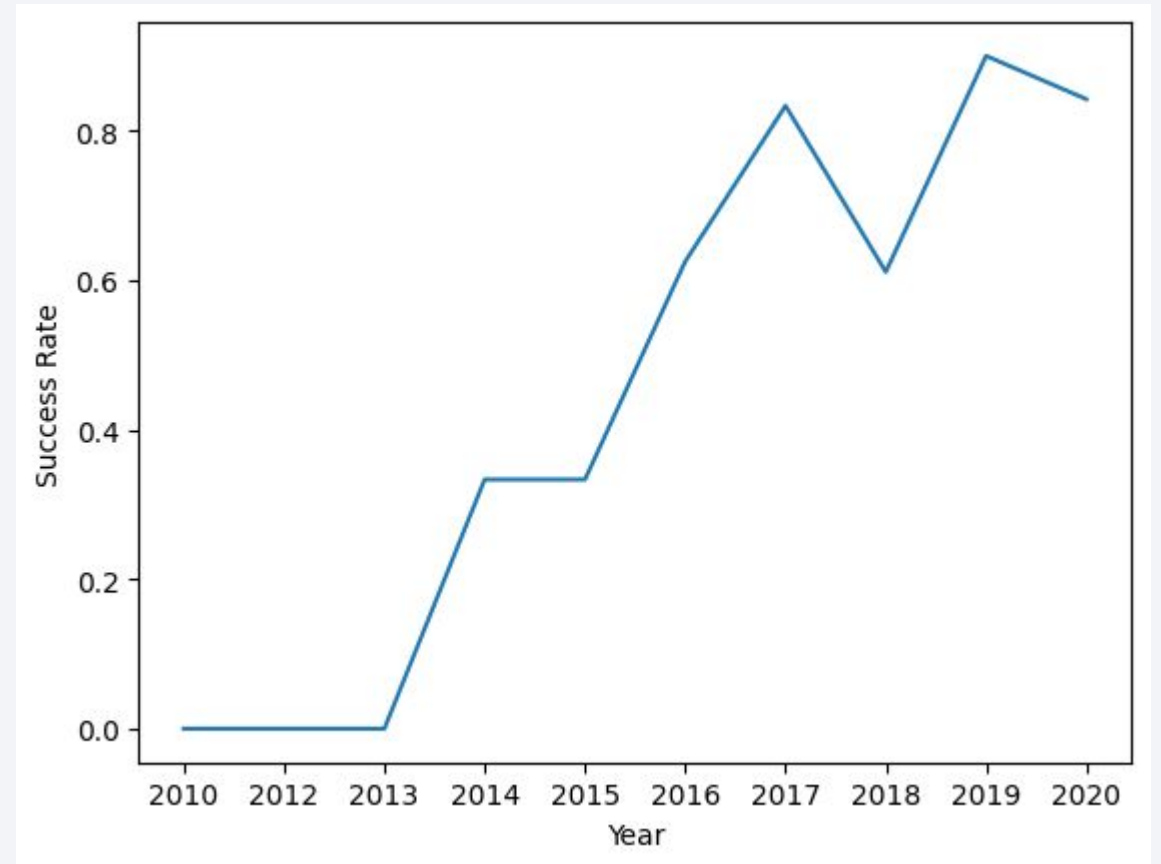
Payload vs. Orbit Type

- ISS orbit has a large range of pay loads and a high success rate (class 1)
- Launch sites SO and GEO do not have many launches



Launch Success Yearly Trend

- first 3 years were not successful
- success rate has mostly increased since 2013



All Launch Site Names

- Names of the unique launch sites:

```
%sql SELECT DISTINCT LAUNCH SITE from SPACEXTABLE;
```

Launch_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

Launch Site Names Begin with 'CCA'

- 5 records where launch sites begin with `CCA` (Cape Canaveral)

- ```
%sql SELECT LAUNCH_SITE from SPACEXTBL where (LAUNCH_SITE) LIKE 'CCA%' LIMIT 5;
```

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

# Total Payload Mass

---

- Total payload carried by boosters from NASA:

- ```
%sql select sum(PAYLOAD_MASS__KG_) from SPACEXTBL where customer='NASA (CRS)';
```

```
sum(PAYLOAD_MASS__KG_)
```

```
45596
```

Average Payload Mass by F9 v1.1

- Average payload mass carried by booster version F9 v1.1:

```
%sql SELECT AVG(PAYLOAD_MASS_KG) from SPACEXTBL WHERE BOOSTER_VERSION='F9 v1.1';
```

```
AVG(PAYLOAD_MASS_KG)
2928.4
```

First Successful Ground Landing Date

- Dates of the first successful landing outcome on ground pad:

- ```
%sql SELECT MIN (DATE) FROM SPACEXTBL WHERE LANDING_OUTCOME LIKE 'Success (ground pad)';
```

**MIN (DATE)**

2015-12-22



## Successful Drone Ship Landing with Payload between 4000 and 6000

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- Names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

```
%sql SELECT BOOSTER_VERSION from SPACEXTBL where LANDING OUTCOME='Success (drone ship)' and PAYLOAD MASS KG between
```

| Booster_Version |
|-----------------|
|-----------------|

|             |
|-------------|
| F9 FT B1022 |
|-------------|

|             |
|-------------|
| F9 FT B1026 |
|-------------|

|               |
|---------------|
| F9 FT B1021.2 |
|---------------|

|               |
|---------------|
| F9 FT B1031.2 |
|---------------|

-

# Total Number of Successful and Failure Mission Outcomes

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- Number of successful and failure mission outcomes:

- ```
%sql select count(MISSION_OUTCOME), mission_outcome from SPACEXTBL GROUP BY MISSION_OUTCOME like "Success%";
```

count(MISSION_OUTCOME)	Mission_Outcome
1	Failure (in flight)
100	Success

Boosters Carried Maximum Payload

- Maximum payload boosters

- ```
%sql select booster_version from SPACEXTBL where payload_mass__kg_=(select MAX(payload_mass__kg_) from SPACEXTBL)
```

| 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   |

# 2015 Launch Records

---

- List the failed landing\_outcomes in drone ship, their booster versions, and launch site names for in year 2015

```
%sql select LANDING_OUTCOME, booster version, launch site, date from SPACEXTBL where date like "2015%" and
LANDING_OUTCOME="Failure (drone ship)"
```

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

- 2 failures in time frame

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

---

- Rank 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

```
%sql select LANDING OUTCOME, count(*) from SPACEXTBL where date between "2010-06-04" and "2017-03-20"
```

| Outcome              | Count |
|----------------------|-------|
| NO ATTEMPT           | 10    |
| Failure (Drone)      | 5     |
| Success (Drone)      | 5     |
| Controlled (Ocean)   | 3     |
| Success (Ground Pad) | 3     |
| Failure (Parachute)  | 2     |
| Uncontrolled (Ocean) | 2     |
| Precluded (Drone)    | 1     |

A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The image is a composite of a dark blue sky with stars and a view of the Earth's surface from space. The Earth's surface is mostly dark, with a thin layer of atmosphere visible along the horizon. The city lights are concentrated in the lower right portion of the image, showing a dense network of urban areas. The text "Section 3" is overlaid on the left side of the image.

Section 3

# Launch Sites Proximities Analysis



# Folium: USA Launch Site Overview

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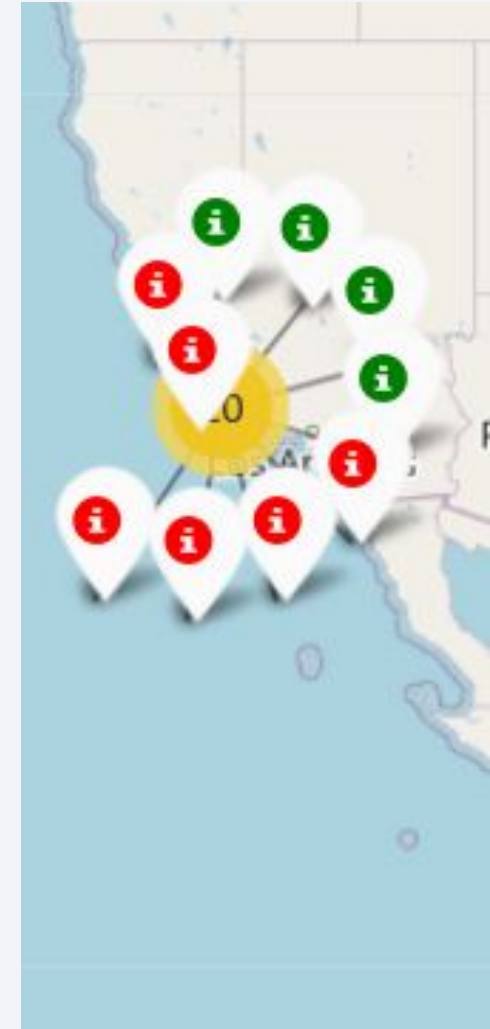
- Zoomed out folium map shows 2 launch sites across the us in California and Florida.
- The yellow circle indicates a mixture of failed and successful launches



# Folium: California Launch Site

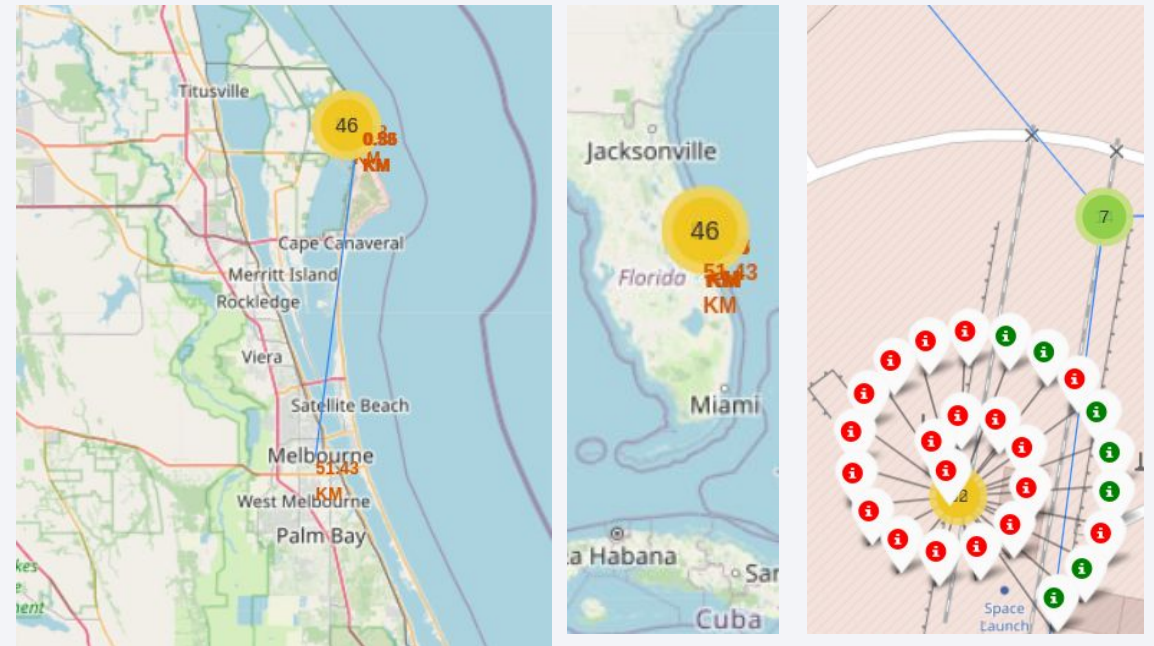
---

- More failed (red) than successful (green) launches can be seen at this site when zoomed in



# Folium: Florida Launch Site

- Images show progressively increased magnification
- Lines depict distances from major infrastructure



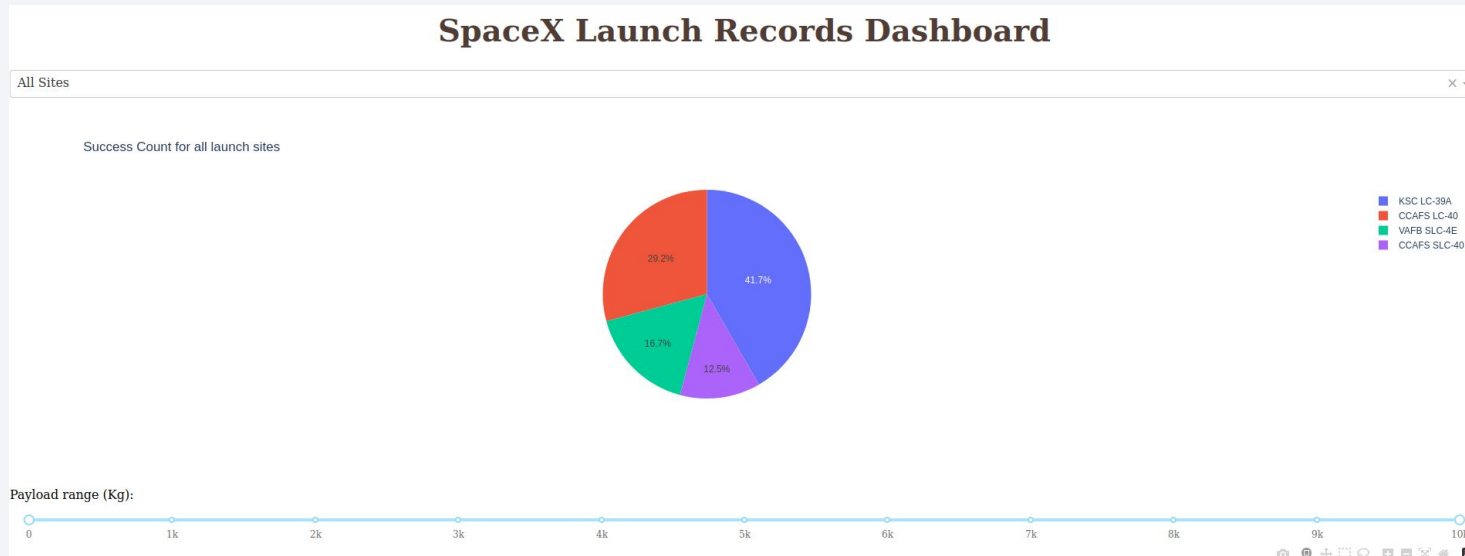


Section 4

# Build a Dashboard with Plotly Dash



# Dash: Success Count Pie Chart



- Explain the important elements and findings on the screenshot

# Dash: Payload slider

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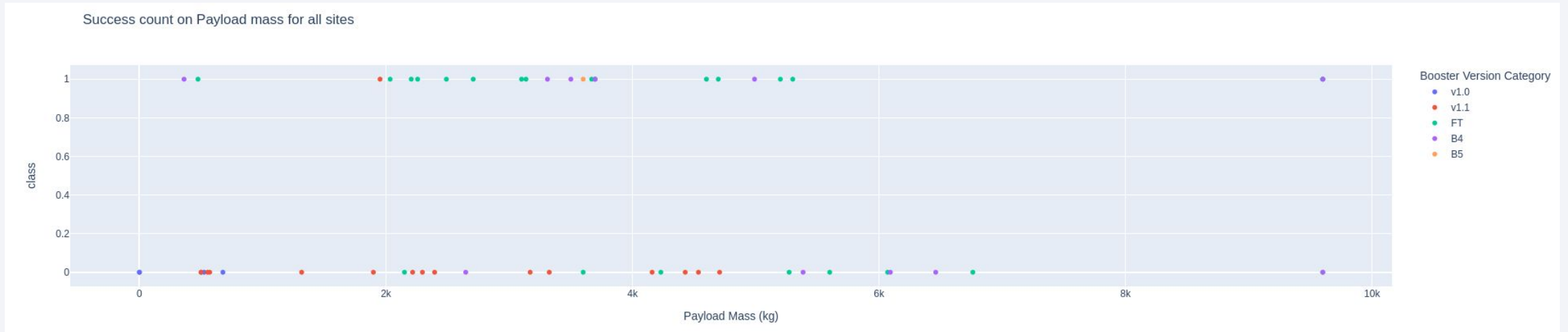
Payload range (Kg):



- A payload slider was created to give insights into how payload was correlated with success rate



# Dash: Payload vs Launch Outcome



- The FT booster ( Green ) shows a prominent success rate at more payloads under 6k than any other booster

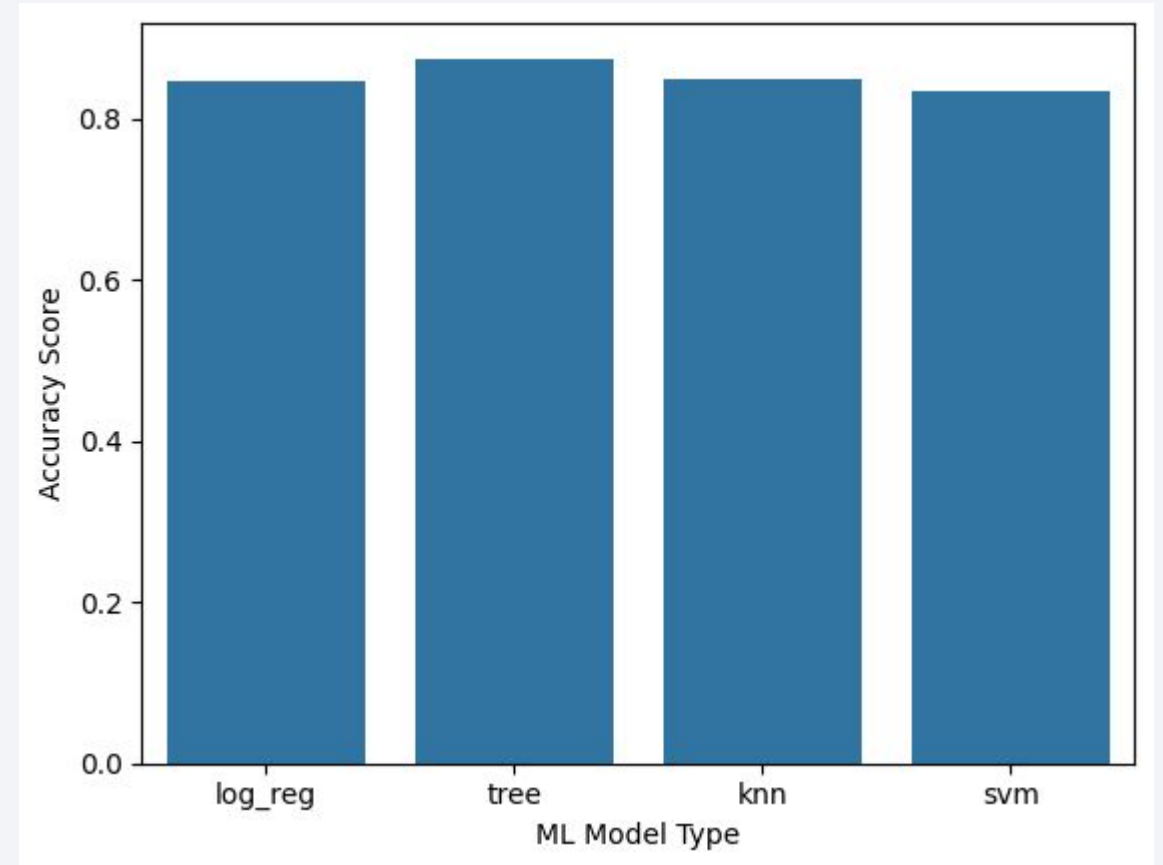
Section 5

# Predictive Analysis (Classification)

# Classification Accuracy

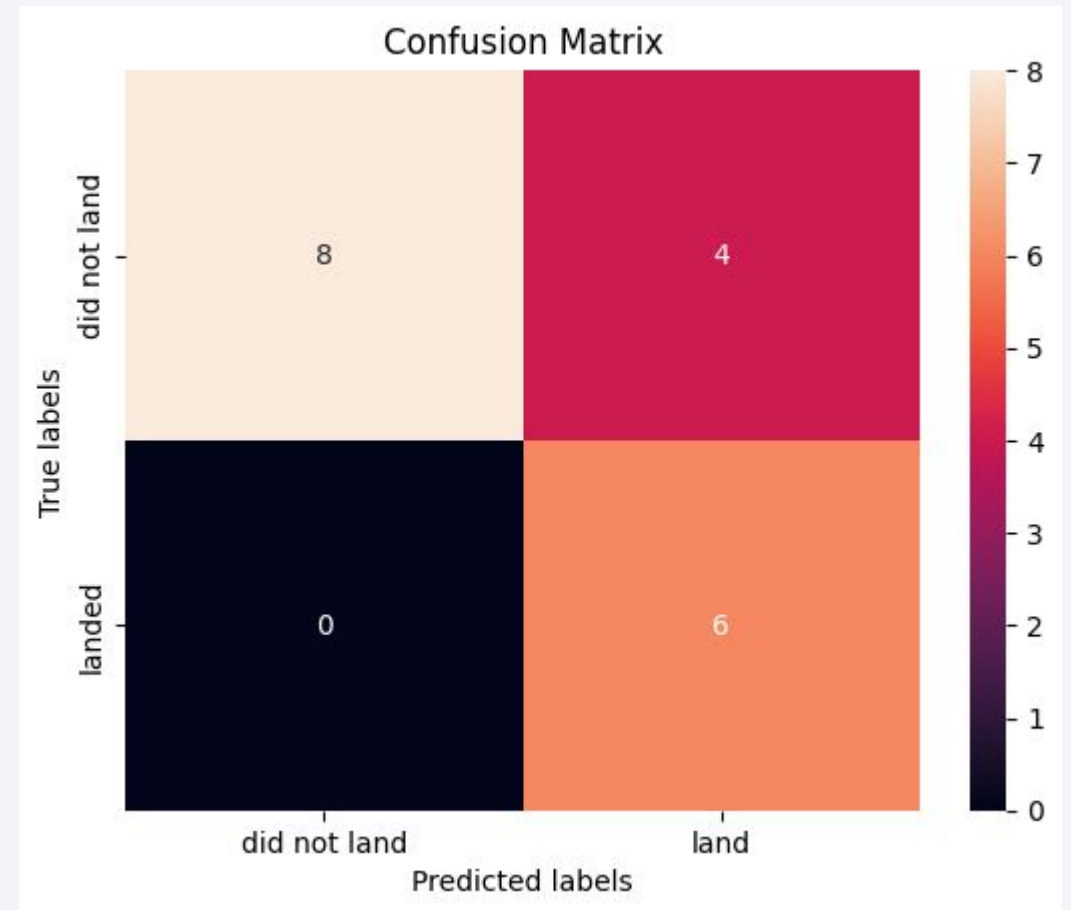
---

- Comparing accuracy scores from each model via sklearn `.score()` on the `xy` test split data, Decision Tree Classifier produces the highest accuracy score
- Tree is highest with ~87%



# Confusion Matrix

- Decision tree classifier confusion matrix revealing 0 false positives and 4 false negatives.



# Conclusions

---

- Most successful launch site is KSC LC-39A
- Higher payload launches have less failures
- Successful landings have increased over time
- A decision tree classifier showed the strongest metrics for predicting landing outcome.
- ...



Thank you!

