

# Winning Space Race with Data Science

Tatarnikov Egor August 10, 2025



#### **Outline**

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

### **Executive Summary**

- Summary of Methodologies:
- Data collection via API and web scraping
- Data wrangling and cleaning
- Exploratory data analysis using SQL and visualizations
- Interactive geographic visual analytics with Folium
- Machine learning for prediction

#### **Summary of Results:**

- Insights from exploratory data analysis
- Interactive visual analytics demonstrated with screenshots
- Performance and outcomes of predictive models

#### Introduction

#### **Project Background:**

- Launch success depends on factors like launch site, payload, and booster type.
- Analyzing past data helps improve future missions.

#### **Problems to Solve:**

- Which sites have the best success rates?
- How does payload affect success?
- Which booster versions perform best?
- Can we predict launch outcomes?



# Methodology

#### **Executive Summary**

- Data collection methodology:
  - Data was collected from SpaceX launch records via APIs and web scraping
- Perform data wrangling
  - Data was cleaned, transformed, and analyzed using SQL and visualization tools to prepare it for modeling and insights
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Classification models were built using Logistic Regression, SVM, Decision Trees, and KNN. Hyperparameters were tuned with GridSearchCV, and models were evaluated using accuracy and confusion matrices on test data

#### **Data Collection**

The data was collected from public SpaceX launch records using APIs and web scraping techniques. It includes detailed information on launch sites, payloads, launch outcomes, and booster versions, compiled into structured datasets for analysis.

#### Sources:

- Space X API (https://api.spacexdata.com/v4/rockets/)
- Wikipedia (https://en.wikipedia.org/wiki/List\_of\_Falcon/\_9/\_and\_Falcon\_Heavy\_launches)

# Data Collection - SpaceX API

 Data collection with SpaceX REST API

#### • GitHub:

https://github.com/EgorTatarnikov/
Coursera IBM DS Final Project/bl
ob/main/11%20Data%20Collectio
n/jupyter-labs-spacex-datacollection-api.ipynb

Now let's start requesting rocket launch data from SpaceX API with the following URL:

In [6]: spacex\_url="https://api.spacexdata.com/v4/launches/past"

In [7]: response = requests.get(spacex\_url)

Check the content of the response

In [8]: print(response.content)

b'[{"fairings":{"reused":false, "recovery attempt":false, "recovered":false, "ships":[]}, "links":{"patch":{"small":"https://imag es2.imgbox.com/94/f2/NN6Ph45r\_o.png","large":"https://images2.imgbox.com/5b/02/QcxHUb5V\_o.png"},"reddit":{"campaign":null,"la unch":null, "media":null, "recovery":null}, "flickr": {"small":[], "original":[]}, "presskit":null, "webcast": "https://www.youtube.c om/watch?v=0a 00nJ Y88","youtube id":"0a 00nJ Y88","article":"https://www.space.com/2196-spacex-inaugural-falcon-1-rocket-los t-launch.html", "wikipedia": "https://en.wikipedia.org/wiki/DemoSat"}, "static fire date utc": "2006-03-17T00:00:00.000Z", "static \_fire\_date\_unix":1142553600,"net":false,"window":0,"rocket":"5e9d0d95eda69955f709d1eb","success":false,"failures":[{"time":3 3, "altitude":null, "reason": "merlin engine failure" }], "details": "Engine failure at 33 seconds and loss of vehicle", "crew": [], "ships":[], "capsules":[], "payloads":["5eb0e4b5b6c3bb0006eeb1e1"], "launchpad": "5e9e4502f5090995de566f86", "flight number": 1, "name": "FalconSat", "date utc": "2006-03-24T22:30:00.000Z", "date unix": 1143239400, "date local": "2006-03-25T10:30:00+12:00", "d ate\_precision":"hour","upcoming":false,"cores":[{"core":"5e9e289df35918033d3b2623","flight":1,"gridfins":false,"legs":false e, "reused":false, "landing attempt":false, "landing success":null, "landing type":null, "landpad":null}], "auto update":true, "tb d":false, "launch\_library\_id":null, "id":"5eb87cd9ffd86e000604b32a"},{"fairings":{"reused":false, "recovery\_attempt":false, "reco vered":false,"ships":[]},"links":{"patch":{"small":"https://images2.imgbox.com/f9/4a/ZboXReNb\_o.png","large":"https://images 2.imgbox.com/80/a2/bkWotCIS\_o.png"},"reddit":{"campaign":null,"launch":null,"media":null,"recovery":null},"flickr":{"small": [],"original":[]},"presskit":null,"webcast":"https://www.youtube.com/watch?v=Lk4zQ2wP-Nc","youtube id":"Lk4zQ2wP-Nc","articl e": "https://www.space.com/3590-spacex-falcon-1-rocket-fails-reach-orbit.html", "wikipedia": "https://en.wikipedia.org/wiki/Demo Sat"}, "static\_fire\_date\_utc":null, "static\_fire\_date\_unix":null, "net":false, "window":0, "rocket": "5e9d0d95eda69955f709d1eb", "su ccess":false, "failures":[{"time":301, "altitude":289, "reason": "harmonic oscillation leading to premature engine shutdown"}],"d etails": "Successful first stage burn and transition to second stage, maximum altitude 289 km, Premature engine shutdown at T+ 7 min 30 s, Failed to reach orbit, Failed to recover first stage", "crew":[], "ships":[], "capsules":[], "payloads":["5eb0e4b6b6c 3bb0006eeb1e2"], "launchpad": "5e9e4502f5090995de566f86", "flight number": 2, "name": "DemoSat", "date utc": "2007-03-21T01:10:00.000 Z", "date unix":1174439400, "date local":"2007-03-21T13:10:00+12:00", "date precision":"hour", "upcoming":false, "cores":[{"cor e":"5e9e289ef35918416a3b2624", "flight":1, "gridfins":false, "legs":false, "reused":false, "landing attempt":false, "landing succes s":null, "landing\_type":null, "landpad":null}], "auto\_update":true, "tbd":false, "launch\_library\_id":null, "id":"5eb87cdaffd86e0006 04b32b"},{"fairings":{"reused":false,"recovery\_attempt":false,"recovered":false,"ships":[]},"links":{"patch":{"small":"http s://images2.imgbox.com/6c/cb/na1tzhHs o.png","large":"https://images2.imgbox.com/4a/80/k1oAkY0k o.png"},"reddit":{"campaign":

# **Data Collection - Scraping**

 Web scraping process from Wikipedia

#### • GitHub:

https://github.com/EgorTatar nikov/Coursera IBM DS Fina I Project/blob/main/11%20 Data%20Collection/jupyterlabs-webscraping.ipynb

```
First, let's perform an HTTP GET method to request the Falcon9 Launch HTML page, as an HTTP response.
 # use requests.get() method with the provided static_url
  # assign the response to a object
  response = requests.get(static url)
 Create a BeautifulSoup object from the HTML response
# Use BeautifulSoup() to create a BeautifulSoup object from a response text content
  soup = BeautifulSoup(response.text, 'html.parser')
 Print the page title to verify if the BeautifulSoup object was created properly
 # Use soup.title attribute
  print(soup.title)
<title>List of Falcon 9 and Falcon Heavy launches - Wikipedia</title>
  Next, we want to collect all relevant column names from the HTML table header
 Let's try to find all tables on the wiki page first. If you need to refresh your memory about BeautifulSoup, please check the external
  reference link towards the end of this lab
  # Use the find all function in the BeautifulSoup object, with element type `table`
  # Assign the result to a list called `html tables`
  html tables = soup.find all('table')
  Starting from the third table is our target table contains the actual launch records.
  # Let's print the third table and check its content
   first launch table = html tables[2]
   print(first launch table)
```

# **Data Wrangling**

#### The data processing workflow included:

- Importing raw datasets
- Handling missing and inconsistent values
- Converting data types for analysis compatibility
- Filtering and selecting relevant features
- Aggregating and summarizing data where necessary
- · Preparing clean datasets ready for exploratory analysis and modeling
- GitHub:

https://github.com/EgorTatarnikov/Coursera IBM DS Final Project/blob/main/12%20Data%20Wrangling/labs-jupyter-spacex-Data%20wrangling.ipynb

#### **EDA** with Data Visualization

During the exploratory data analysis, several types of plots were used to better understand the data:

- Catplot: Used to visualize the relationship between categorical variables and numerical outcomes. This helped in comparing success rates across different launch sites and categories.
- **Barplot**: Employed to show the count or frequency of different categories, such as the number of launches per site or booster version. Barplots provided a clear overview of categorical distributions.
- Lineplot: Used to observe trends and changes over time or across ordered variables, such as payload mass versus launch success. This helped to identify patterns and correlations in continuous data.
- GitHub

  <u>URL:https://github.com/EgorTatarnikov/Coursera\_IBM\_DS\_Final\_Project/blob/main/22</u>

  <u>%20EDA%20with%20Visualization/edadataviz.ipynb</u>

  11

### **EDA** with SQL

- Counted launches by site
- Calculated success rates per site
- Filtered launches by payload range Joined launch data with booster info
- Analyzed launch trends over time
- Aggregated failure causes by site Identified top-performing sites
- Retrieved site locations for mapping
- GitHub URL:
   https://github.com/EgorTatarnikov/Coursera IBM DS Final Project/blob/main
   /21%20EDA%20with%20SQL/jupyter-labs-eda-sql-coursera sqllite.ipynb

### Build an Interactive Map with Folium

- Markers: Added to pinpoint exact launch site locations for easy identification on the map.
- Circles: Used to highlight launch sites with a visible radius, making them stand out and showing their proximity areas.
- Lines (Polylines): Drawn between launch sites and nearby points of interest (coastlines, cities, railways) to visualize distances and spatial relationships. These objects help reveal geographic patterns and assist in analyzing how location factors may affect launch success.
- GitHub URL:

https://github.com/EgorTatarnikov/Coursera IBM DS Final Project/blob/main/31% 20Interactive%20Visual%20Analytics%20with%20Folium%20(Maps)/lab\_jupyter\_launch\_site\_location.ipynb

### Build a Dashboard with Plotly Dash

- Dropdown for Launch Site Selection: Allows users to filter data by specific launch sites or view all sites together, enabling focused analysis.
- Pie Chart: Shows success vs. failure counts for selected launch sites to quickly visualize success rates.
- Payload Range Slider: Lets users select a payload mass range to explore its effect on launch outcomes.
- Scatter Plot: Displays correlation between payload mass and launch success, with points color-coded by booster version for deeper insight.
- GitHub URL: <a href="https://github.com/EgorTatarnikov/Coursera IBM DS Final Project/blob/main/32%20Plotly/spacex-dash-app.py">https://github.com/EgorTatarnikov/Coursera IBM DS Final Project/blob/main/32%20Plotly/spacex-dash-app.py</a>

# Predictive Analysis (Classification)

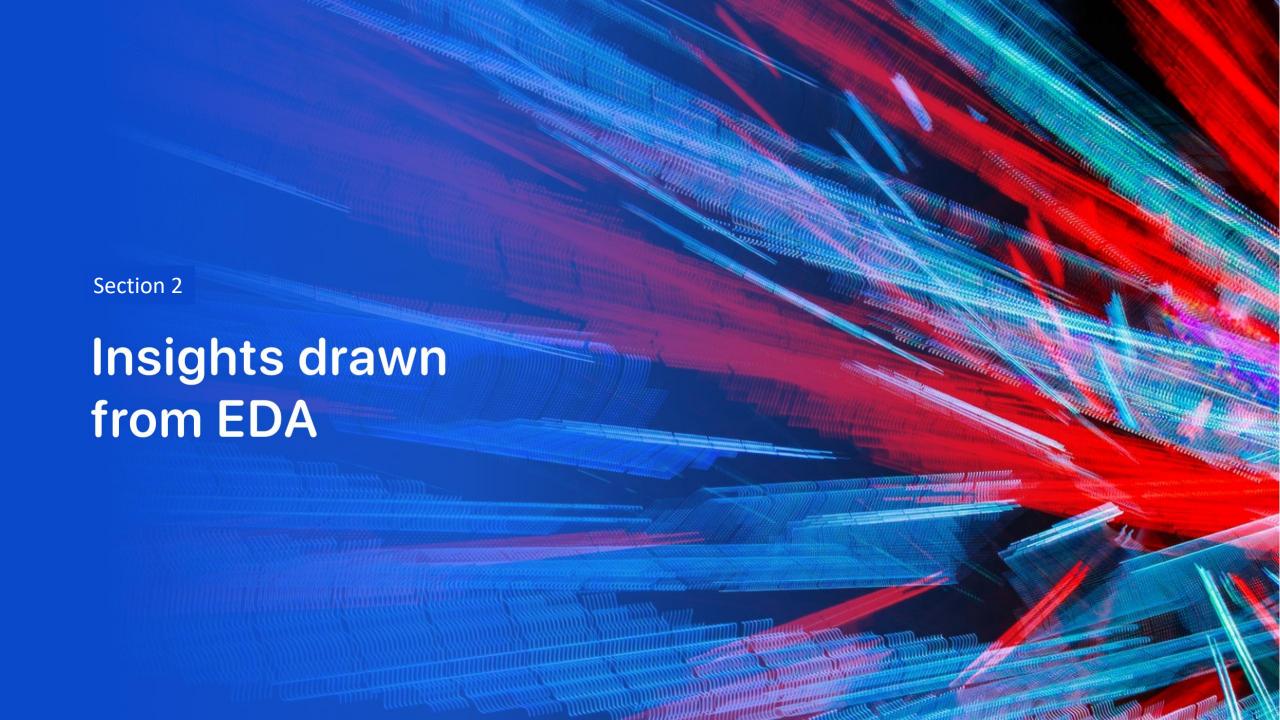
- Data Preparation: Cleaned and standardized input features; split data into training and testing sets.
- **Model Selection:** Tested multiple algorithms Logistic Regression, Support Vector Machine (SVM), Decision Tree, and K-Nearest Neighbors (KNN).
- Hyperparameter Tuning: Used GridSearchCV with cross-validation to find optimal hyperparameters for each model.
- Model Evaluation: Assessed models on test data using accuracy and confusion matrices to compare performance.
- **Best Model Identification:** Selected the model with the highest test accuracy and balanced classification results.
- Flowchart: Data Preprocessing → Train-Test Split → Model Training → Hyperparameter Tuning (GridSearchCV) → Model Evaluation → Best Model Selection
- GitHub URL:

  <a href="https://github.com/EgorTatarnikov/Coursera">https://github.com/EgorTatarnikov/Coursera</a> IBM DS Final Project/blob/main/41%20ML/Spa

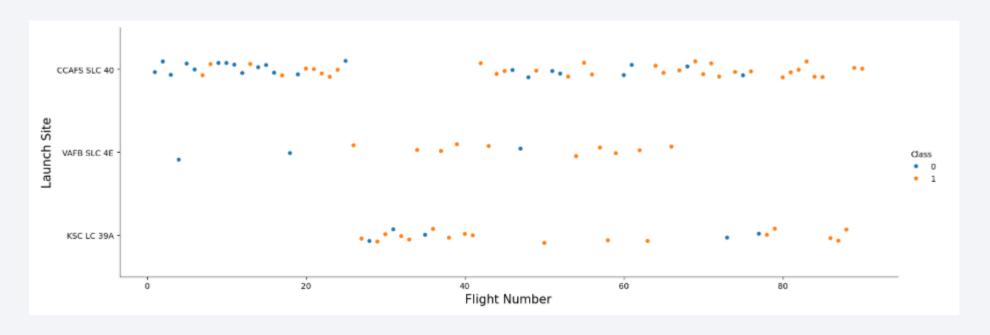
  <a href="mailto:cex-mailto:c

#### Results

- SpaceX operates from four distinct launch sites.
- Initial launches were conducted for SpaceX and NASA missions.
- The average payload capacity of the Falcon 9 v1.1 booster is approximately 2,928 kg.
- The first successful booster landing occurred in 2015, five years after the inaugural launch.
- Several Falcon 9 booster versions successfully landed on drone ships, often carrying payloads above the average.
- Nearly all missions were successful overall.
- Two booster versions, F9 v1.1 B1012 and B1015, failed drone ship landings in 2015.
- Landing success rates improved steadily over time.

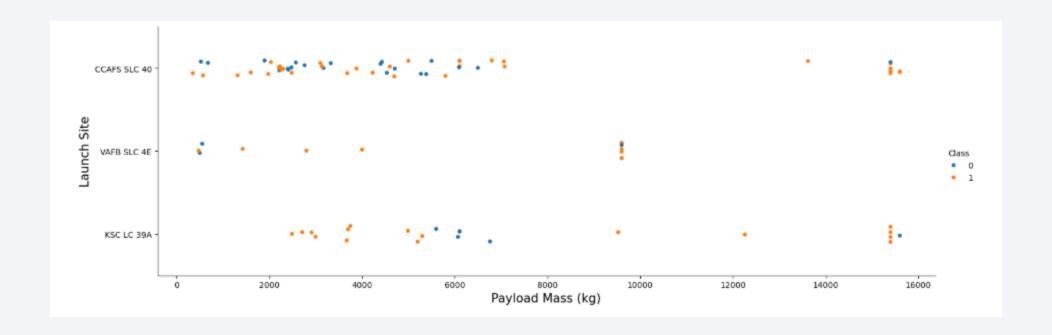


### Flight Number vs. Launch Site



- The most successful launch site at present is CCAF5 SLC-40, with the highest number of recent successful launches.
- VAFB SLC-4E ranks second, followed by KSC LC-39A in third place.
- Overall, the success rate has shown consistent improvement over time.

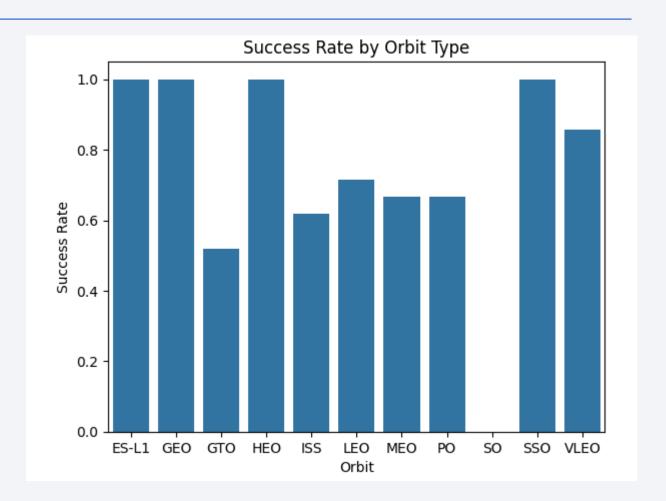
### Payload vs. Launch Site



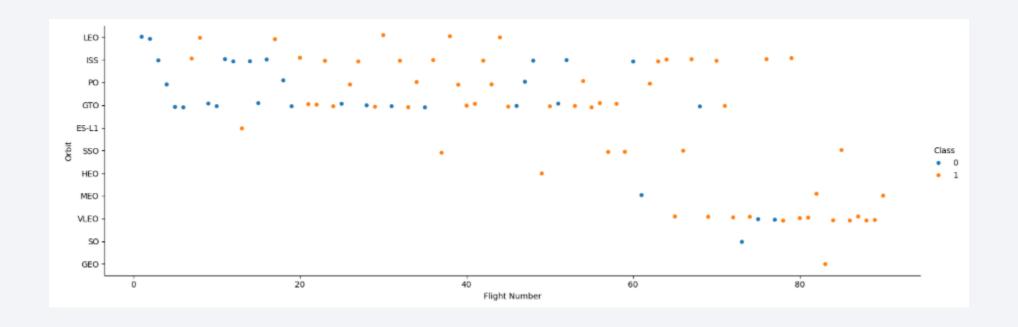
- Payloads exceeding 9,000 kg (roughly the weight of a school bus) show a very high success rate.
- Payloads above 12,000 kg appear to be launched exclusively from CCAFS SLC-40 and KSC LC-39A.

# Success Rate vs. Orbit Type

• ES-L1, GEO, HEO, SSO, VLEO had the most success rate.

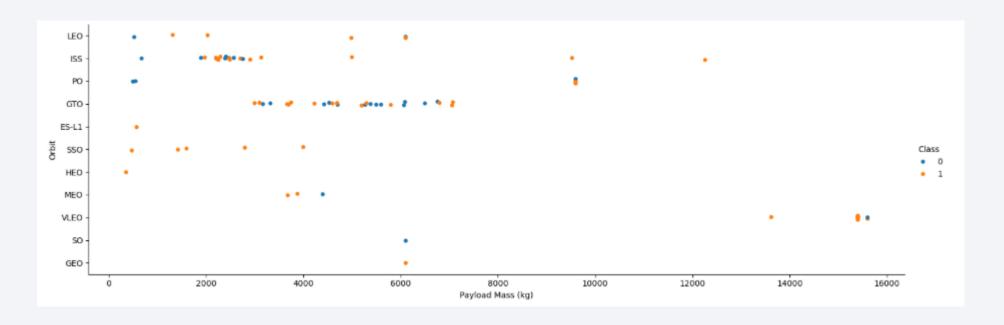


# Flight Number vs. Orbit Type



- For LEO orbits, success rates are linked to the number of flights,
- For GTO orbits, there is no apparent correlation between flight number and success.

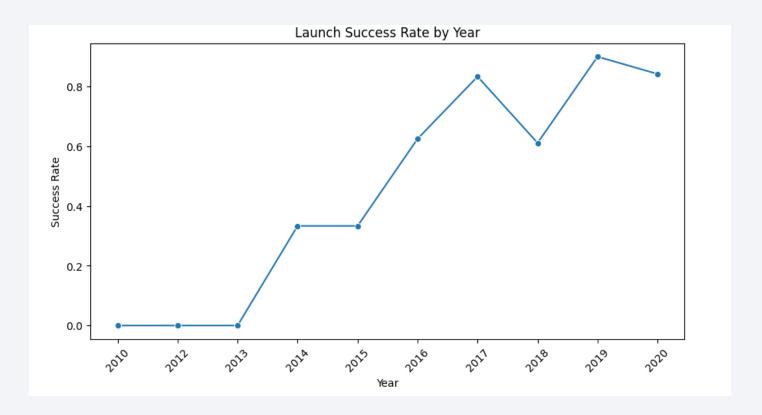
### Payload vs. Orbit Type



- There is no clear relationship between payload and success rate for GTO orbits.
- The ISS orbit accommodates the widest payload range and maintains a high success rate.
- Launches to SO and GEO orbits are relatively rare.

# Launch Success Yearly Trend

- The success rate began rising in 2013 and continued to improve through 2020.
- The initial three years appear to have been a period of adjustments and technological advancements.



#### All Launch Site Names

• There are four launch sites from which SpaceX launches rockets.

```
In [11]: %sql SELECT DISTINCT Launch_Site FROM SPACEXTABLE;

* sqlite://my_data1.db
Done.

Out[11]: 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`

In [13]:	SELEC	<pre>%%sql SELECT * FROM SPACEXTABLE WHERE Launch_Site LIKE 'CCA%' LIMIT 5;</pre>									
	* sqlite:///my_data1.db Done.										
Out[13]:	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	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**

The total payload carried by boosters from NASA

# Average Payload Mass by F9 v1.1

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

# First Successful Ground Landing Date

• The first successful landing outcome on ground pad

```
In [16]:  

**Sql
SELECT MIN(Date) AS FirstSuccessLandingDate
FROM SPACEXTABLE
WHERE Landing_Outcome LIKE '*Success*' AND Landing_Outcome LIKE '*ground pad*';

* sqlite:///my_data1.db
Done.

Out[16]:  

FirstSuccessLandingDate

2015-12-22
```

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

 List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

```
In [17]:

**Ssql

SELECT DISTINCT Booster_Version
FROM SPACEXTABLE
WHERE Landing_Outcome LIKE '*Success*' AND Landing_Outcome LIKE '**drone ship*'
AND Payload_Mass__kg_ > 4000 AND Payload_Mass__kg_ < 6000;

* sqlite:///my_data1.db
Done.

Out[17]:

Booster_Version

F9 FT B1022

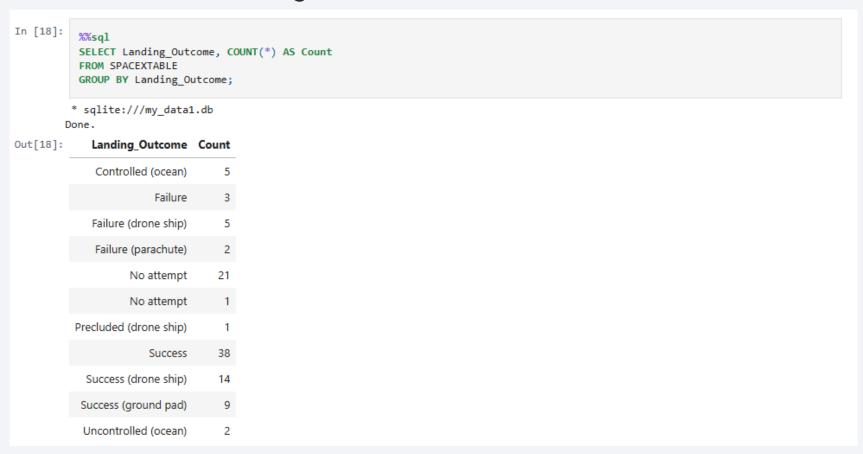
F9 FT B1021.2

F9 FT B1021.2

F9 FT B1031.2
```

#### Total Number of Successful and Failure Mission Outcomes

Total number of landing outcomes



# **Boosters Carried Maximum Payload**

• The names of the boosters which have carried the maximum payload mass

```
In [19]:
          %%sql
          SELECT Booster Version, Payload Mass kg
           FROM SPACEXTABLE
          WHERE Payload_Mass__kg_ = (SELECT MAX(Payload_Mass__kg_) FROM SPACEXTABLE);
         * sqlite:///my data1.db
Out[19]: Booster Version PAYLOAD MASS KG
             F9 B5 B1048.4
                                         15600
             F9 B5 B1049.4
                                         15600
             F9 B5 B1051.3
                                         15600
             F9 B5 B1056.4
                                         15600
            F9 B5 B1048.5
                                         15600
             F9 B5 B1051.4
                                         15600
             F9 B5 B1049.5
                                         15600
             F9 B5 B1060.2
                                         15600
             F9 B5 B1058.3
                                         15600
             F9 B5 B1051.6
                                         15600
             F9 B5 B1060.3
                                         15600
             F9 B5 B1049.7
                                         15600
```

#### 2015 Launch Records

• Failed landing outcomes in drone ship in 2015

```
In [20]:
          %%sql
           SELECT
              substr(Date, 6, 2) AS Month,
              Landing Outcome,
              Booster Version,
              Launch Site
           FROM SPACEXTABLE
          WHERE substr(Date, 1, 4) = '2015'
          AND Landing Outcome LIKE '%Failure%'
          AND Landing Outcome LIKE '%drone ship%';
         * sqlite:///my data1.db
Out[20]: Month Landing_Outcome Booster_Version Launch_Site
             01 Failure (drone ship)
                                    F9 v1.1 B1012 CCAFS LC-40
             04 Failure (drone ship)
                                    F9 v1.1 B1015 CCAFS LC-40
```

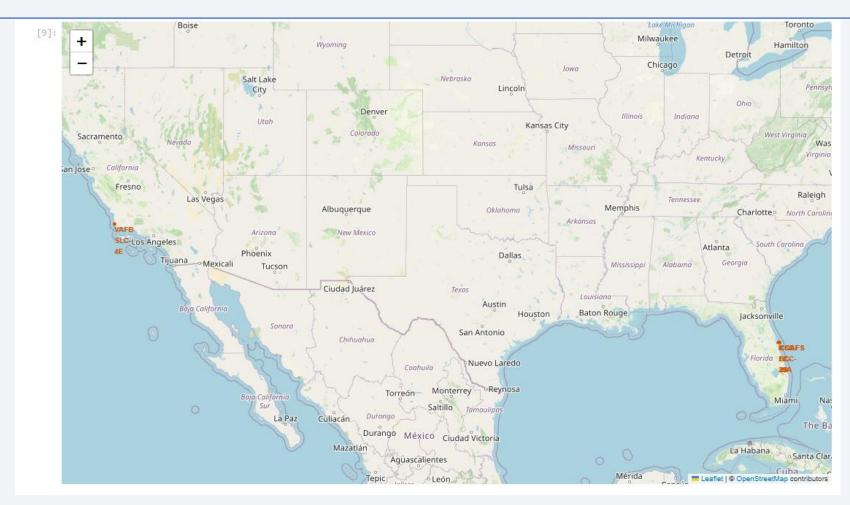
#### Rank Landing Outcomes Between 2010-06-04 and 2017-03-20

```
In [21]:
           SELECT Landing_Outcome, COUNT(*) AS Count
           FROM SPACEXTABLE
           WHERE Date BETWEEN '2010-06-04' AND '2017-03-20'
           GROUP BY Landing_Outcome
           ORDER BY Count DESC;
          * sqlite:///my_data1.db
         Done.
Out[21]:
             Landing_Outcome Count
                    No attempt
                                    10
            Success (drone ship)
             Failure (drone ship)
                                    5
           Success (ground pad)
              Controlled (ocean)
            Uncontrolled (ocean)
              Failure (parachute)
          Precluded (drone ship)
```

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



# SpaceX launch sites



• Location of four launch sites from which SpaceX launches rockets.

#### Markers of launch sites





West coast East coast

#### Landmarks of launch sites



Vandenberg Space Launch Complex



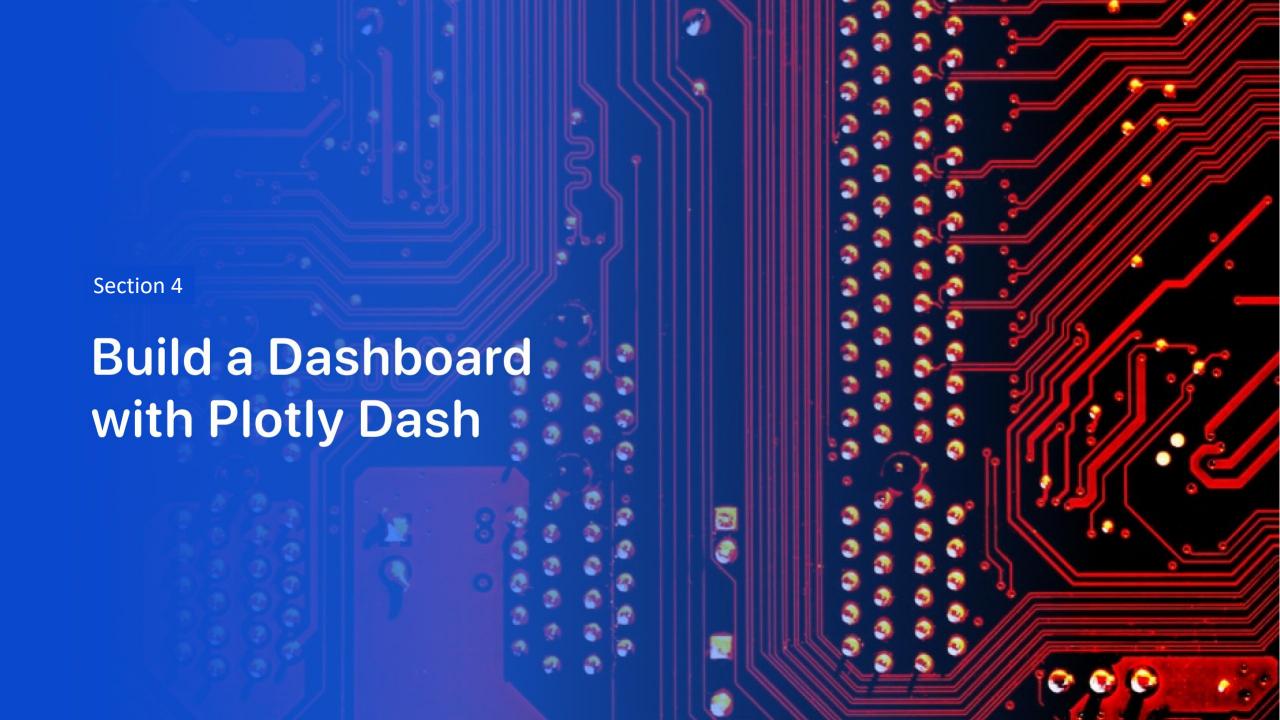
Lompoc Airport



Cape Canaveral Air Force Station

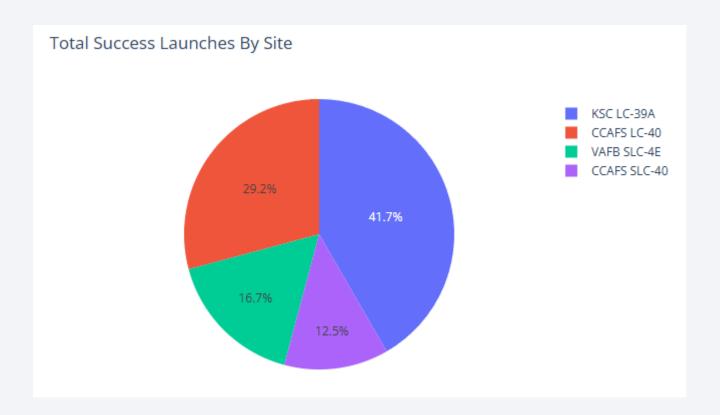


Cape Canaveral SFS Skid Strip



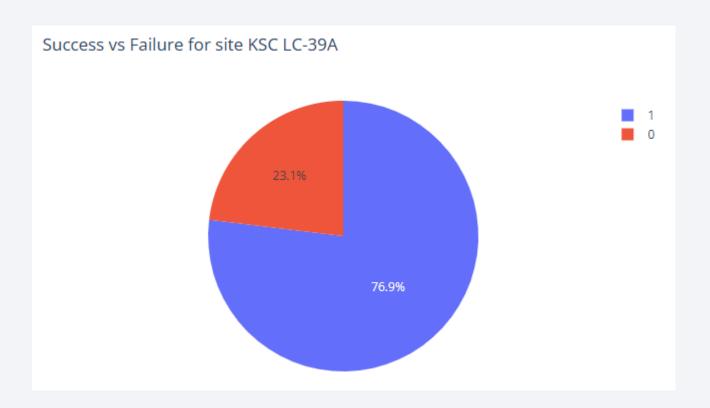
# Successful Launches by Site

 The largest number of successful launches from KSC LC-39A

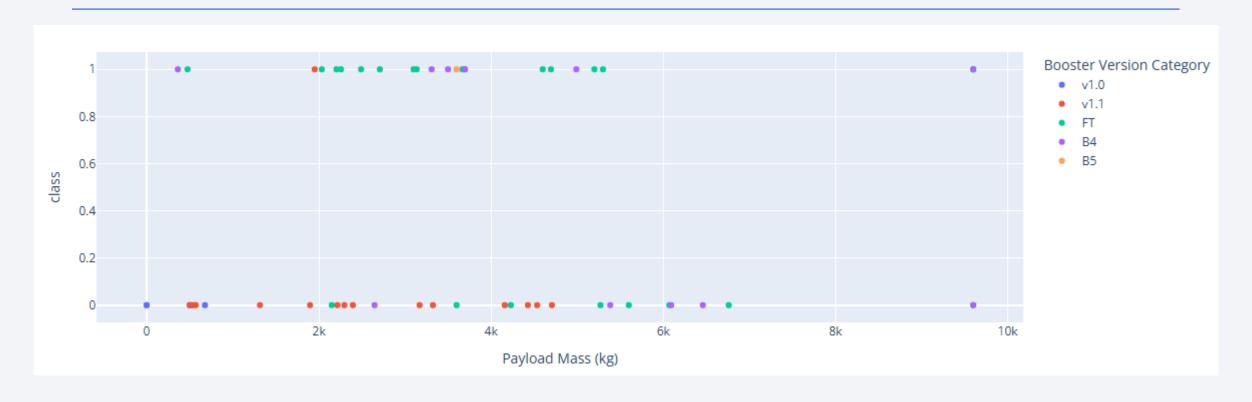


#### Launches from KSC LC-39A

• KSC LC-39A has 76.9% launch success rate



# Payload and Success Correlation

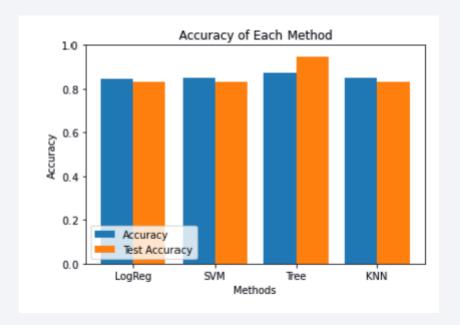


Most successful: payloads < 6,000 kg with FT boosters



# Classification Accuracy

- Four models were trained.
- Decision Tree classifier showed the highest accuracy.

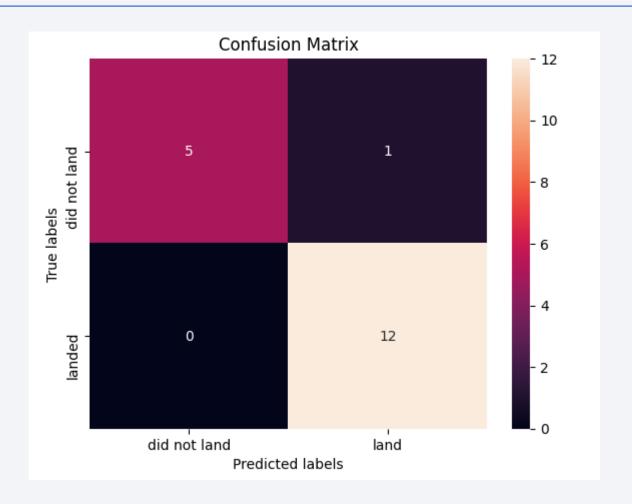


#### **Confusion Matrix**

#### There are:

- 12 TP
- 5 TN
- 1 FP
- 0 FN

Good result



#### **Conclusions**

- Multiple data sources were analyzed, enabling progressive refinement of insights.
- The most reliable launch site is CCAFS SLC-40, with the highest recent success rates.
- Payloads under 6,000 kg combined with FT boosters show the greatest success.
- Launch success rates have steadily improved since 2013, reflecting advancements in technology and processes.
- Certain orbits, such as ISS and LEO, show higher success rates compared to less common orbits like SO and GEO.
- Among the tested models, the Decision Tree classifier demonstrated the best accuracy for predicting successful landings, making it a valuable tool for operational planning.

# **Appendix**

 Additional materials on GitHub: <u>https://github.com/EgorTatarnikov/Coursera\_IBM\_DS\_Final\_Project</u>

