



IBM Developer  
SKILLS NETWORK

# Winning Space Race with Data Science

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# Outline

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

# Executive Summary

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- Summary of methodologies
  - Data collection using API and Web Scraping
  - Data Wrangling
  - Exploratory Data Analysis using SQL and visualization
  - Interactive Visual Analytics
  - Dashboarding
  - Predictive analytics
- Summary of all results
  - Identify the key features that affect the result of the launch
  - An interactive dashboard for real-time analysis has been developed
  - Produce a model capable of predicting the outcome of rocket launch with an accuracy of 83.8%

# Introduction

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- Project background and context

SpaceX advertises Falcon 9 rocket launches at a cost of 62 million dollars, compared to 165 million dollars for other providers

This cost saving is essentially due to the fact that SpaceX reuses its rocket first stage

Being able to predict the likelihood of a successful landing gives the advantage, to an alternate company, to estimate more precisely the launch costs, and apply a better bid strategies when competing against SpaceX

And this is the aim of this analysis

- Problems you want to find answers

- What are the factors that determine the successful landing of a rocket?
- Can we predict the likelihood of a successful landing?
- What is the maximum accuracy of the prediction?



Section 1

# Methodology

# Methodology

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## Executive Summary

- Data collection methodology:
  - Calling SpaceX REST API
  - Web Scraping on Wikipedia pages
- Perform data wrangling
  - Data Inspection, Data Cleaning, Encoding Categorical values, Data Normalization
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
  - Models from Scikit-Learn library
  - Find best parameters using *GridSearchCV*

# Data Collection

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We have 2 different data sources involving 2 different methods:

- **SpaceX REST API**

- Call the API using *requests* object
- Parse the JSON response to Pandas *DataFrame*

- **Web Scraping**

- *BeautifulSoup* is used to extract data from Wikipedia web page
- SpaceX launch records contained in HTML Table are parsed and converted to Pandas *DataFrame*

# Data Collection – SpaceX API

- Need to call multiple endpoints to get the whole data set
- Start requesting rocket launch historical data
- From *launchpad* , request for launch site name and location
- From *payloads* , request for payload weight and orbit
- From *cores* , request for rocket details
- Compile all into a single DataFrame

## API CALL

### 1. Call the endpoint

```
spacex_url="https://api.spacexdata.com/v4/launches/past"  
response = requests.get(spacex_url)
```

### 2. Extract JSON from HTTP reponse

```
json = response.json()
```

### 3. Parse JSON into DataFrame

```
dataframe = pd.json_normalize(json)
```

- Link to notebook: [https://github.com/zina-mahfarivo/spacex-capstone-project/blob/main/1\\_jupyter-labs-spacex-data-collection-api.ipynb](https://github.com/zina-mahfarivo/spacex-capstone-project/blob/main/1_jupyter-labs-spacex-data-collection-api.ipynb)



# Data Collection - Scraping

- Obtain page content by using HTTP get
  - Create BeautifulSoup object from HTML page
  - Select Launch HTML table
  - Iterate on rows
  - Extract data from each cell
  - Create DataFrame from all result lists
- 
- Link to notebook: [https://github.com/zina-mahfarivo/spacex-capstone-project/blob/main/2\\_jupyter-labs-webscraping.ipynb](https://github.com/zina-mahfarivo/spacex-capstone-project/blob/main/2_jupyter-labs-webscraping.ipynb)

## WEB SCRAPING

### 1. Get page content

```
static_url = "https://en.wikipedia.org/w/index.php"
            + "?title=List_of_Falcon_9_and_Falcon_Heavy_launches"
            + "&oldid=1027686922"
response = requests.get(static_url)
```

### 2. Create soup

```
soup = BeautifulSoup(response.content, "html.parser")
```

### 3. Iterate on rows

```
for table_number, table in enumerate(
    soup.find_all('table', "wikitable plainrowheaders collapsible")
):
```

### 4. Extract data from cell

```
launch_site = row[2].a.string
launch_dict["Launch site"].append(launch_site)
```

### 5. Create DataFrame

```
df = pd.DataFrame(
    { key:pd.Series(value) for key, value in launch_dict.items() }
)
```

# Data Wrangling

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- Clean data
  - Identify missing values
  - Replace missing *PayloadMass* by mean value

- Ensure Data Consistency
  - Check data types

- Explore Data
  - Nb of launches per site
  - Nb of occurrence of each orbit
  - Nb of occurrence of outcome

OUTCOME	COUNT	SITE	LANDING	CLASS
True ASDS	41	Drone Ship	Success	1
None None	19	Not Attempted	Not Attempted	0
True RTLS	14	Ground Pad	Success	1
False ASDS	6	Drone Ship	Fail	0
True Ocean	5	Ocean	Success	1
False Ocean	2	Ocean	Fail	0
None ASDS	2	Drone Ship	Not Attempted	0
False RTLS	1	Ground Pad	Fail	0

- Transform Data
  - From *Outcome*, extract/encode the *Class* of the launch

- Link to notebook: [https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/3\\_labs-jupyter-spacex-Data%20wrangling.ipynb](https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/3_labs-jupyter-spacex-Data%20wrangling.ipynb)

# EDA with Data Visualization

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## Plotted Charts

- Flight Number VS Launch Site: To verify the pattern between the 2 variables and the outcome
- Payload VS Launch Site: To visualize which site launch heavy payload
- Orbit VS Success Rate: To identify which orbits have high landing success rate
- Orbit VS Flight Number: To check the relationship between Flight Nb and orbit
- Orbit VS Payload: To verify to which orbit heavy/light payloads are sent
- Year VS Success Rate: To emphasize from which year success rate started to increase
- Link to notebook: <https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/4-jupyter-lab-data-visualization.ipynb>

# EDA with SQL

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- List of unique launch sites in the space mission
- List 5 records where launch sites begin with 'CCA'
- Total payload mass carried by boosters launched by NASA (CRS)
- Average payload mass carried by booster version F9 v1.1
- Date of first successful landing on ground pad
- Boosters with payload between 4000 and 6000 kg that successfully landed on drone ship
- Total number of successful and failure mission outcomes
- Names of boosters that carried the maximum payload mass
- Count of landing outcomes between 2 dates in descending order
- Link to notebook: [https://github.com/zina-mahfarivo/spacex-capstone-project/blob/main/5\\_jupyter-lab-eda-with-sql.ipynb](https://github.com/zina-mahfarivo/spacex-capstone-project/blob/main/5_jupyter-lab-eda-with-sql.ipynb)

# Build an Interactive Map with Folium

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- Circle - Used to indicate a launch site
- Marker - Combined with a Circle, is used to indicate a launch site
- Marker - Also used to indicate the outcome of each launch in each site
- MarkerCluster - Used to display a constellation of Markers
- MousePosition - Indicates the latitude and longitude of the mouse pointer
- Line - Used to indicate distance
- Link to notebook: [https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/6\\_jupyter-lab\\_launch-site-location.ipynb](https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/6_jupyter-lab_launch-site-location.ipynb)



# Build a Dashboard with Plotly Dash

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## INPUTS

- Drop-down list - Allows to select different launch sites
- Range slider - Allows to select a range of payload weight

## CHARTS

- Pie chart - Shows the proportion of success landing for all or the selected site
  - Scatter plot - Displays the correlation between the Payload and the Outcome
- 
- Link to notebook: [https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/spacex\\_dash\\_app.py](https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/spacex_dash_app.py)

# Predictive Analysis (Classification)

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- Algorithms: Logistic Regression, SVM, Decision Tree, KNN
  - Use `GridSearchCV` object to find the best parameters
  - Display confusion matrix
  - Calculate accuracy on training and test sets
  - Compare scores to identify the best model
- 
- Link to notebook: [https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/7\\_jupyter-lab\\_SpaceX\\_Machine\\_Learning\\_Prediction.ipynb](https://github.com/zina-mahefarivo/spacex-capstone-project/blob/main/7_jupyter-lab_SpaceX_Machine_Learning_Prediction.ipynb)

# Results

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- Exploratory data analysis results
- Interactive analytics demo in screenshots
- Predictive analysis results



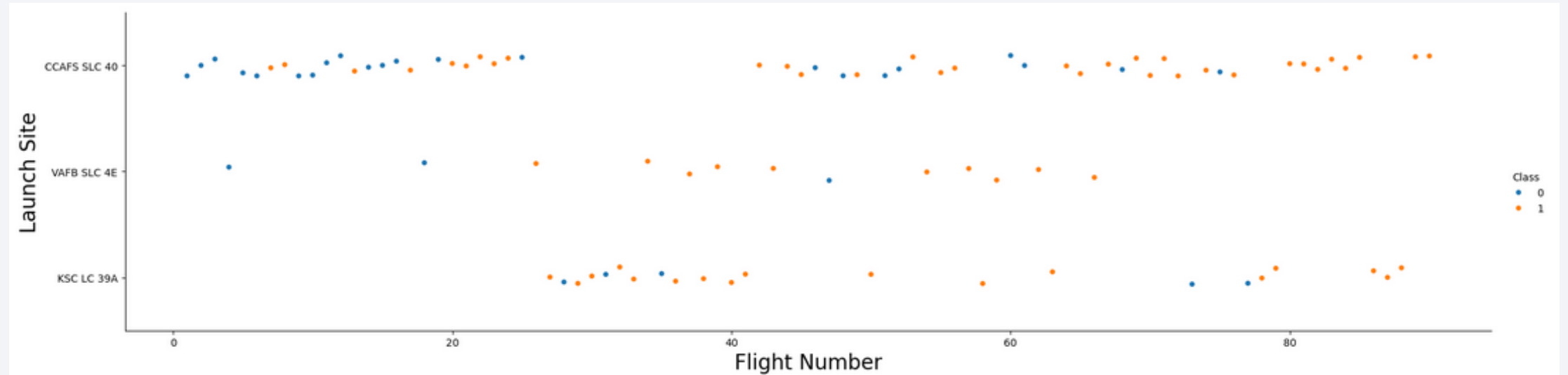
The background of the slide is an abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks in shades of red and cyan, creating a sense of motion and depth. A faint, light blue grid pattern is also visible, particularly in the lower-left quadrant. The overall effect is high-tech and digital.

Section 2

# Insights drawn from EDA



# Flight Number vs. Launch Site

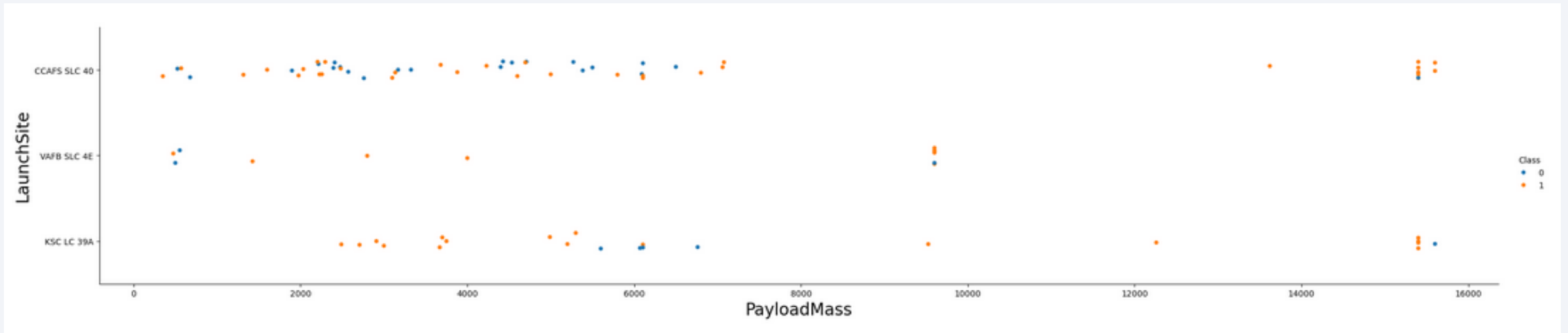


## Insights:

- There are more successful landing starting from Flight Number 25
- With more attempts, we have more successful outcome



# Payload vs. Launch Site

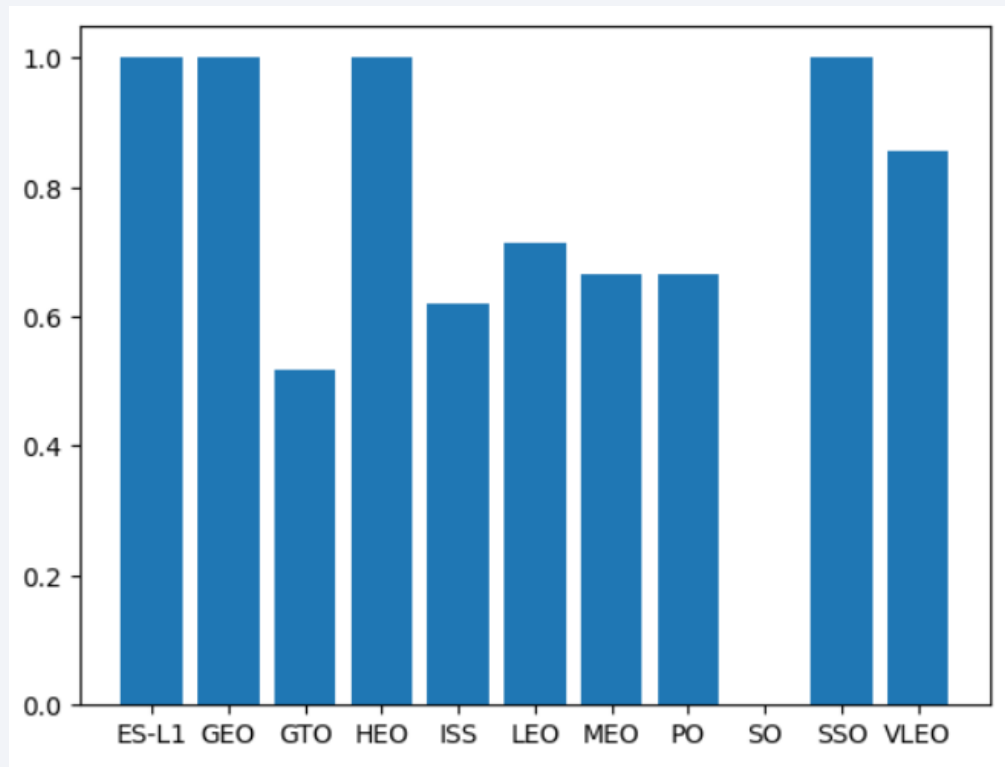


## Insights:

- No heavy launches has been sent from **VAFB SLC 4E**
- There are more failed landing attempts on **CCAFS SLC 40** than on other sites

# Success Rate vs. Orbit Type

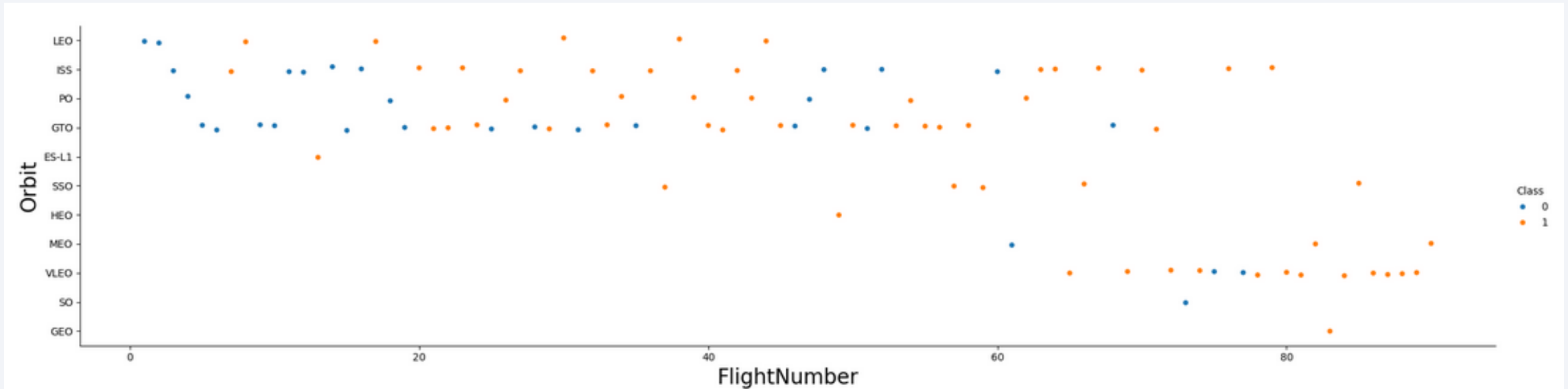
---



## Insights:

- Rockets sent to **ES-L1**, **GEO**, **HEO**, **SSO** and **VLEO** have high success rate
- No rockets sent to **SO** have landed successfully

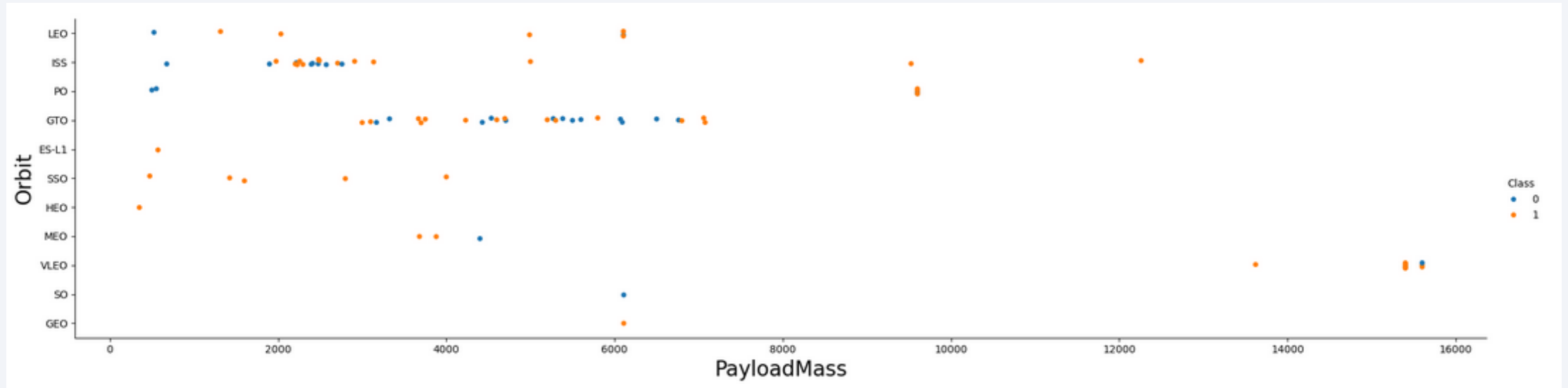
# Flight Number vs. Orbit Type



## Insights:

- For LEO orbit, the outcome appears related to the Flight Number
- No rockets sent to **SO** have landed successfully

# Payload vs. Orbit Type

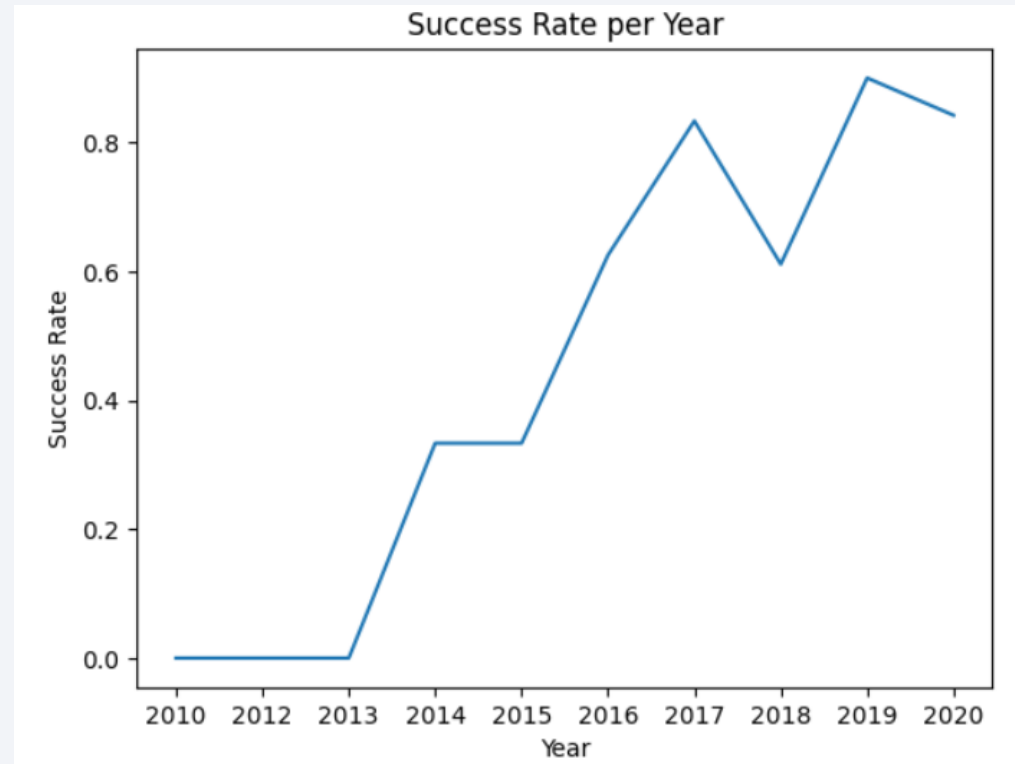


## Insights:

- For heavy payloads, there is more successful landing for **ISS**, **PO**, and **VLEO**
- Successful and failed landing are mixed for **GTO**

# Launch Success Yearly Trend

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## Insights:

- The landing success rate kept increasing since 2013 till 2020



# All Launch Site Names

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```
In [19]: %sql select distinct Launch_Site from SPACEXTABLE
* sqlite:///my_data1.db
Done.
Out[19]: Launch_Site
         CCAFS LC-40
         VAFB SLC-4E
         KSC LC-39A
         CCAFS SLC-40
```

- Use of the keyword **DISTINCT** to return unique Launch Site Name
- There is only 4 different launch Site used by SpaceX

# Launch Site Names Begin with 'CCA'

```
In [18]: %%sql
select * from SPACE_TABLE
where Launch_Site like 'CCA%'
limit 5;

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

Out[18]:

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 attachment
2012-10-08	0:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attachment
2013-03-01	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attachment

- `where Launch_Site like 'CCA%'` indicates that Launch Site must start with 'CCA'
- `limit 5;` restricts the results to 5 records only

# Total Payload Mass

---

In [23]:

```
%%sql  
  
select sum(PAYLOAD_MASS__KG_) as TOTAL_PAYLOAD_MASS  
from SPACEXTABLE  
where Customer = 'NASA (CRS)';
```

\* sqlite:///my\_data1.db

Done.

Out[23]:

**TOTAL\_PAYLOAD\_MASS**

45596

- The total payload carried by boosters from NASA is 45596 kg

# Average Payload Mass by F9 v1.1

---

```
In [26]: %%sql

select avg(PAYLOAD_MASS__KG_) as AVERAGE_PAYLOAD_MASS
from SPACEXTABLE
where Booster_Version = 'F9 v1.1'

* sqlite:///my_data1.db
Done.
Out[26]: AVERAGE_PAYLOAD_MASS
          2928.4
```

- Boosters version F9 v1.1 carry in average a payload of 2928.4 kg

# First Successful Ground Landing Date

---

```
In [30]: %%sql
          select min(Date) as FIRST_SUCCESS_LANDING
          from SPACEXTABLE
          where Landing_Outcome = 'Success (ground pad)'

* sqlite:///my_data1.db
Done.
Out[30]: FIRST_SUCCESS_LANDING
          2015-12-22
```

- The first successful Ground Landing was on 22 December 2015



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

---

```
In [33]: %%sql

select Booster_Version
from SPACEXTABLE
where Landing_Outcome = 'Success (drone ship)'
and PAYLOAD_MASS_KG_ between 4000 and 6000

* sqlite:///my_data1.db
Done.
Out[33]: Booster_Version
         F9 FT B1022
         F9 FT B1026
         F9 FT B1021.2
         F9 FT B1031.2
```

- There are 4 different booster versions that have successfully landed on drone ship with payload mass greater than 4000 but less than 6000

# Total Number of Successful and Failure Mission Outcomes

---

In [34]:

```
%%sql  
  
select Mission_Outcome, count(Mission_Outcome)  
from SPACEXTABLE  
group by Mission_Outcome
```

\* sqlite:///my\_data1.db

Done.

Out[34]:

Mission_Outcome	count(Mission_Outcome)
Failure (in flight)	1
Success	98
Success	1
Success (payload status unclear)	1

- Only 1 Failure Mission Outcome has been found in the dataset

# Boosters Carried Maximum Payload

In [37]:

```
%%sql

select Booster_Version
from SPACEXTABLE
where PAYLOAD_MASS_KG_ = (select max(PAYLOAD_MASS_KG_) from SPACEXTABLE)
```

\* sqlite:///my\_data1.db  
Done.

Out[37]:

**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

- There are 12 different Boosters that carried the maximum payload

# 2015 Launch Records

---

```
In [41]: %%sql

select substr(Date, 6,2) as month, Landing_Outcome, Booster_Version
from SPACEXTABLE
where substr(Date,0,5)='2015'
and Landing_Outcome = 'Failure (drone ship)'
-- limit = 1

* sqlite:///my_data1.db
Done.
Out[41]:
```

month	Landing_Outcome	Booster_Version
01	Failure (drone ship)	F9 v1.1 B1012
04	Failure (drone ship)	F9 v1.1 B1015

- There are 2 failed landing\_outcomes on drone ship in year 2015  
First was on **January** and the second in **April**

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

```
In [45]: %%sql
select Landing_Outcome as OUTCOME, count(Landing_Outcome) as COUNT
from SPACEXTABLE
where Date between '2010-06-04' and '2017-03-20'
group by Landing_Outcome
order by 2 desc

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

Out[45]:

OUTCOME	COUNT
No attempt	10
Success (drone ship)	5
Failure (drone ship)	5
Success (ground pad)	3
Controlled (ocean)	3
Uncontrolled (ocean)	2
Failure (parachute)	2
Precluded (drone ship)	1

- The most frequent outcome is **No attempt**
- The least recurring outcome is **Precluded on Drone Ship**

A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The background is a deep blue gradient.

Section 3

# Launch Sites Proximities Analysis

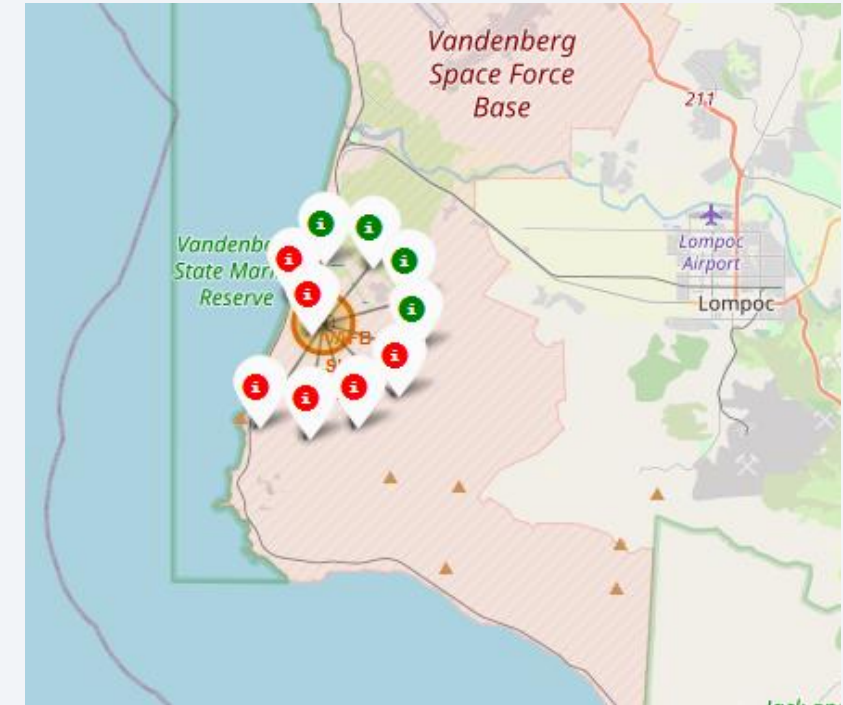
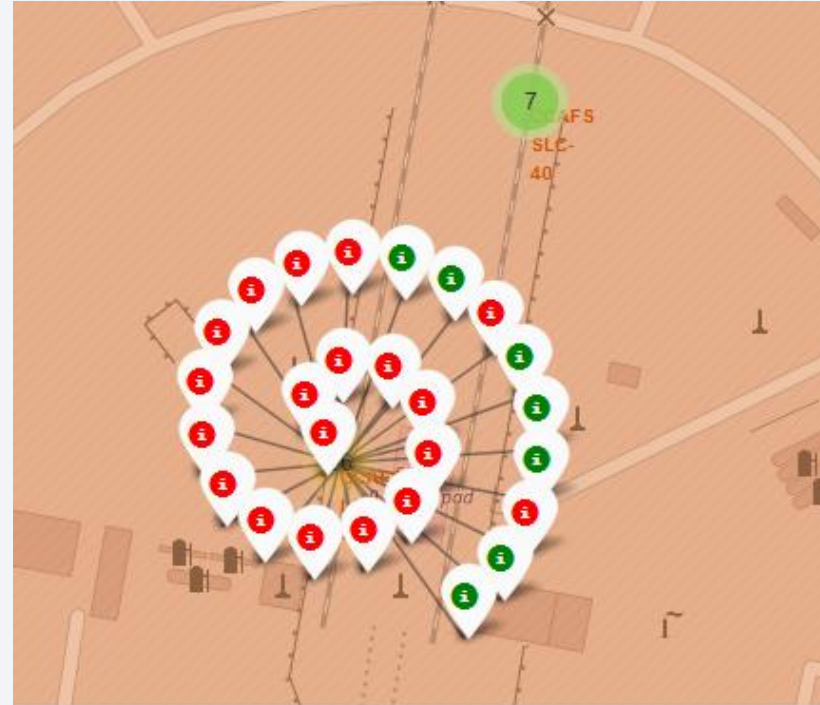
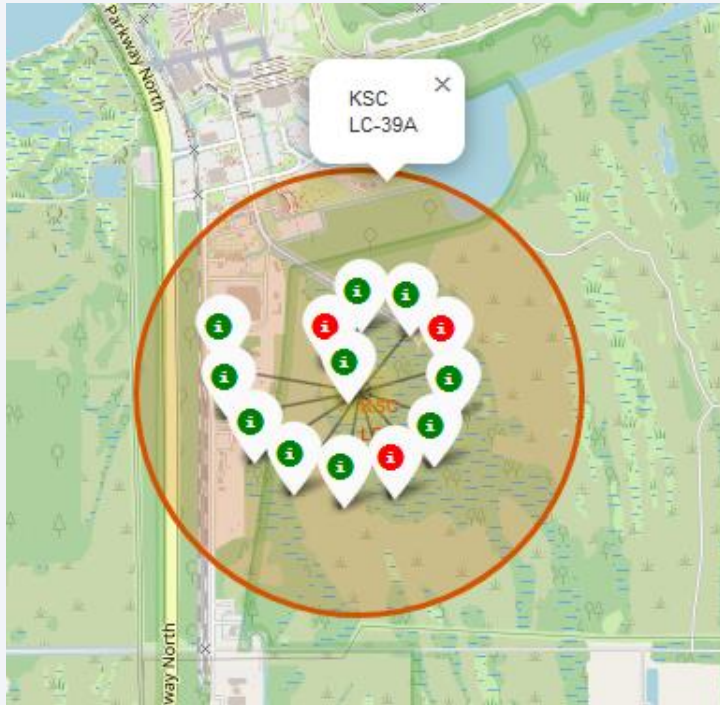
# SpaceX Launch Sites



- All launch sites are located on American coasts
- Only VAFB SLC 4E is located on west coast. All the others are on east coast



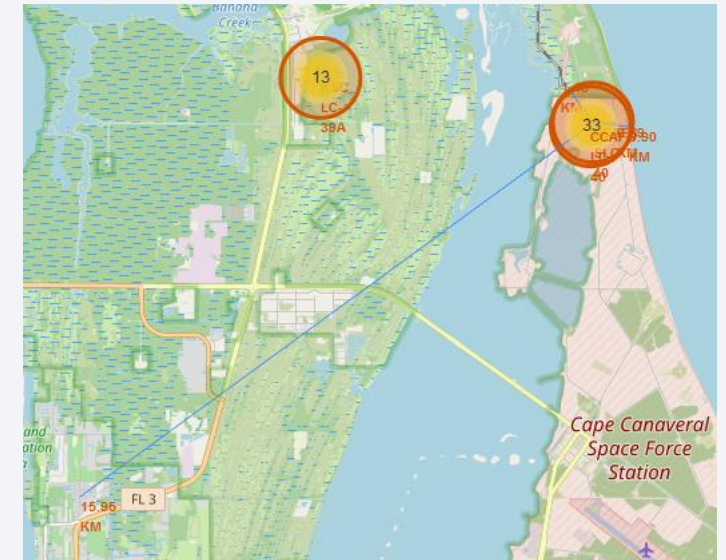
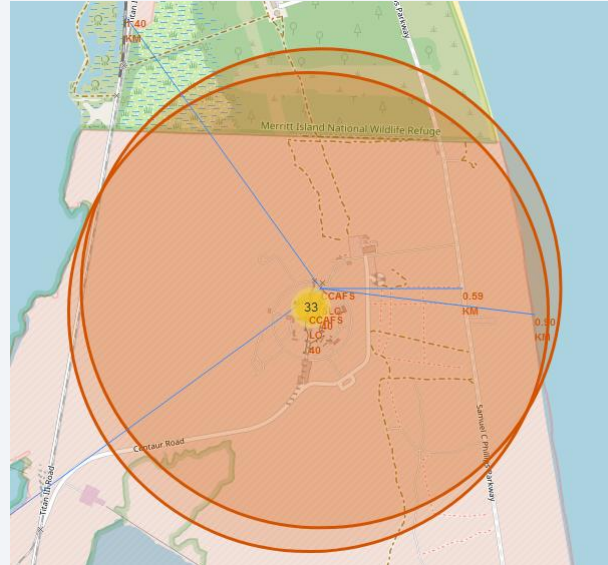
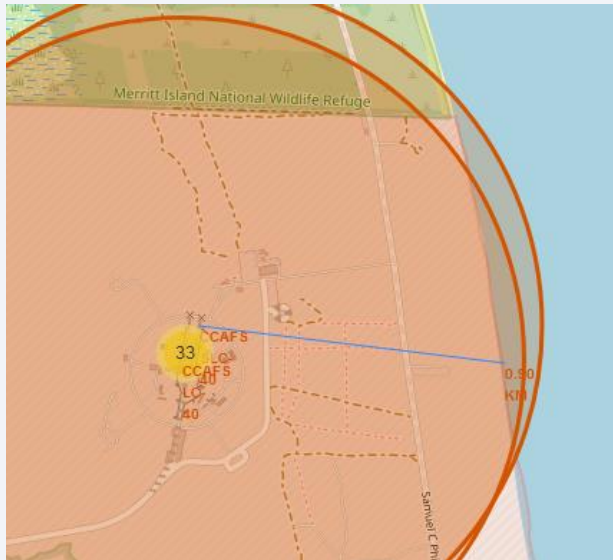
# Landing outcome for each Launch Site



- A successful landing is represented by a green Icon, And Red Icon for a failure
- Launch sites with high success rates can be easily identified



# Launch Site Proximities



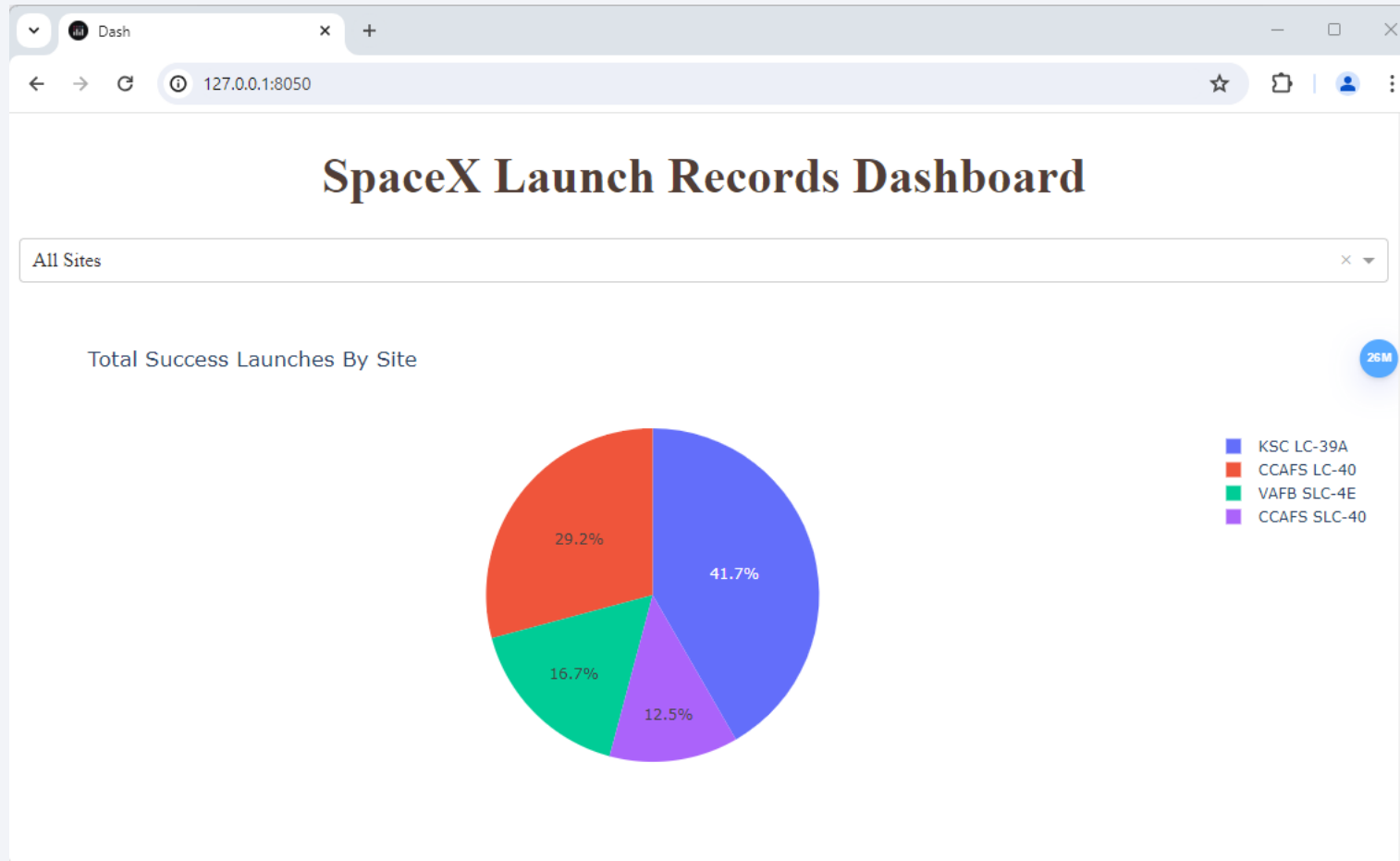
- Space launch sites are located near the equator
- Launch sites are in close proximity to railways to facilitate material (heavy cargo) transportation
- Launch sites are in close proximity to highways to facilitate people and material transportation
- Launch sites are in close proximity to coastline as water is safety buffer in case of launch failure
- Launch sites are not in close proximity to cities to ensure the safety of the population



Section 4

# Build a Dashboard with Plotly Dash

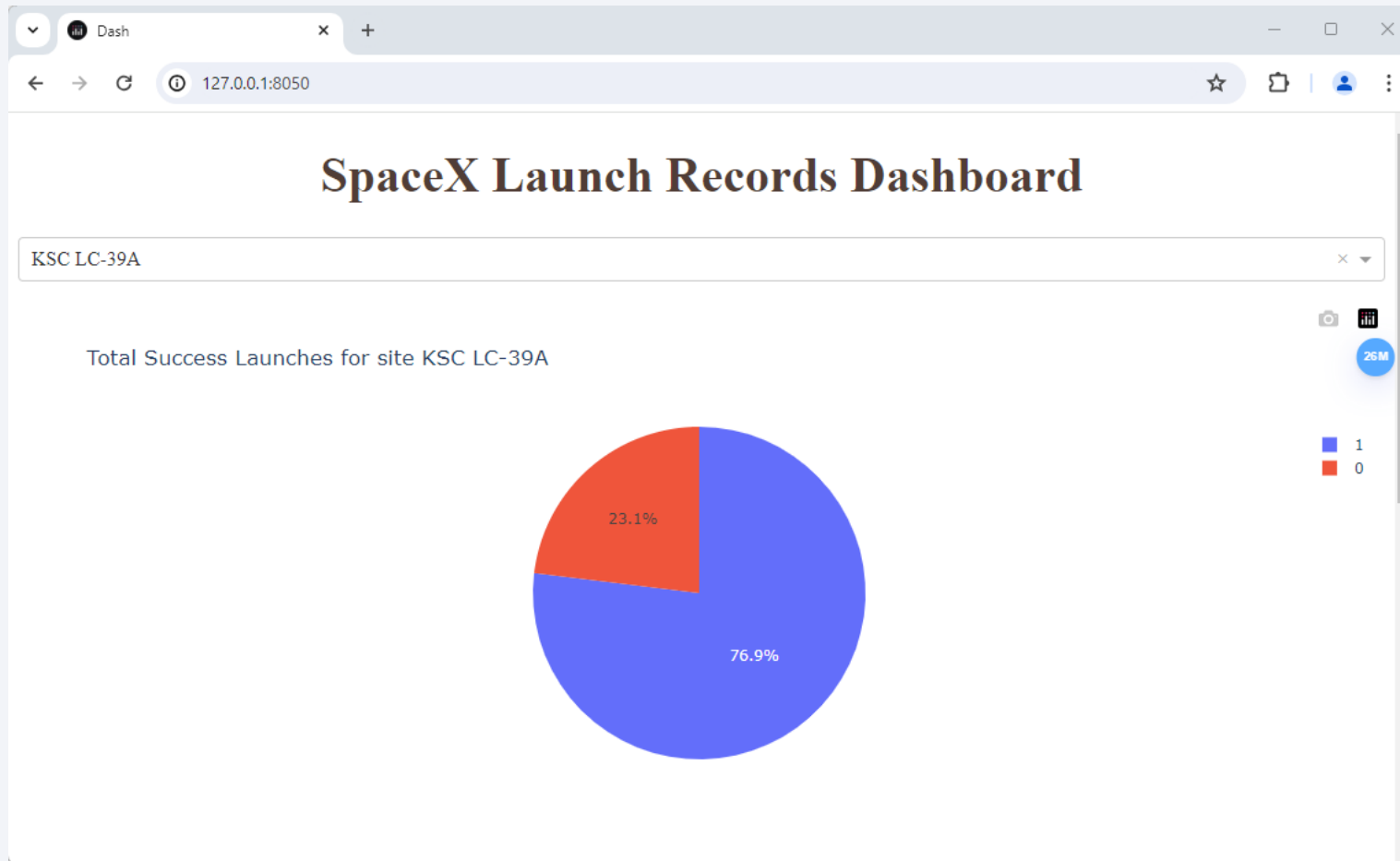
# Dashboard – Success Launches by Site



- The dropdown list allows to select launch Site
- KSC LC-39A has the highest success rate
- CCAFS SLC-40 has the lowest success rate



# Dashboard - Site with Highest Success Rate



- The proportion of success landing for KSC LC-39A is **76.9%**
- The ratio of failed landing for KSC LC-39A is **23.1%**
- KSC LC-39A is selected in the dropdown list

# Dashboard – Correlation between Payload and Success



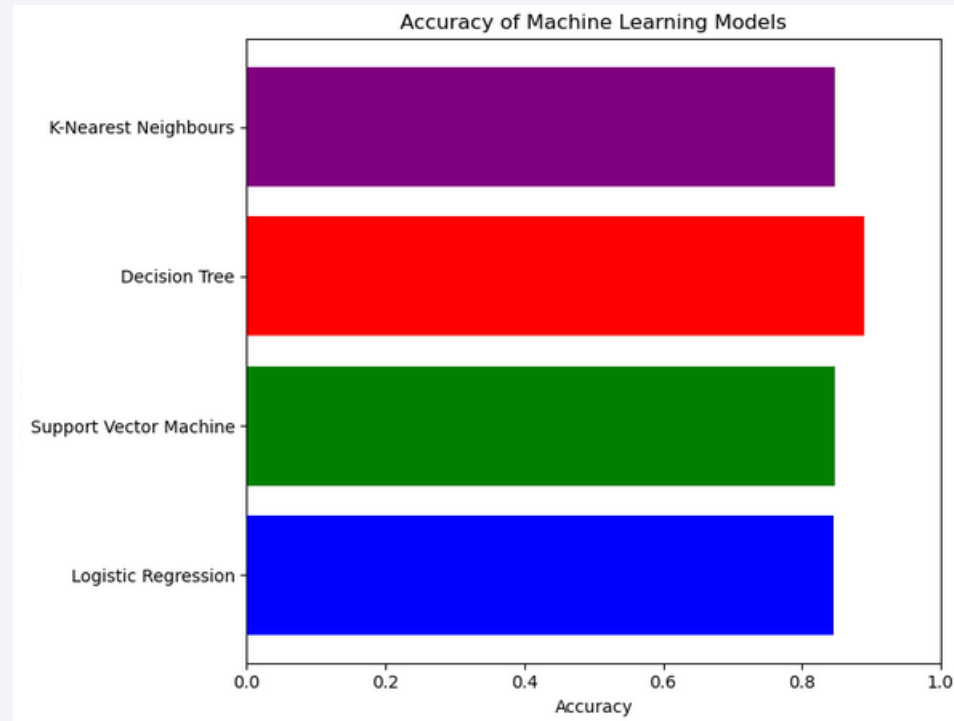
- For light payload, the boosters B4, FT and V1.1 provide a better success rate
- For medium payload, the boosters B4, FT provide a better results
- For heavy payload, only B4 provides a good success rate
- The 1<sup>st</sup> stage has higher chance to land if they carried medium payload

Section 5

# Predictive Analysis (Classification)

# Classification Accuracy

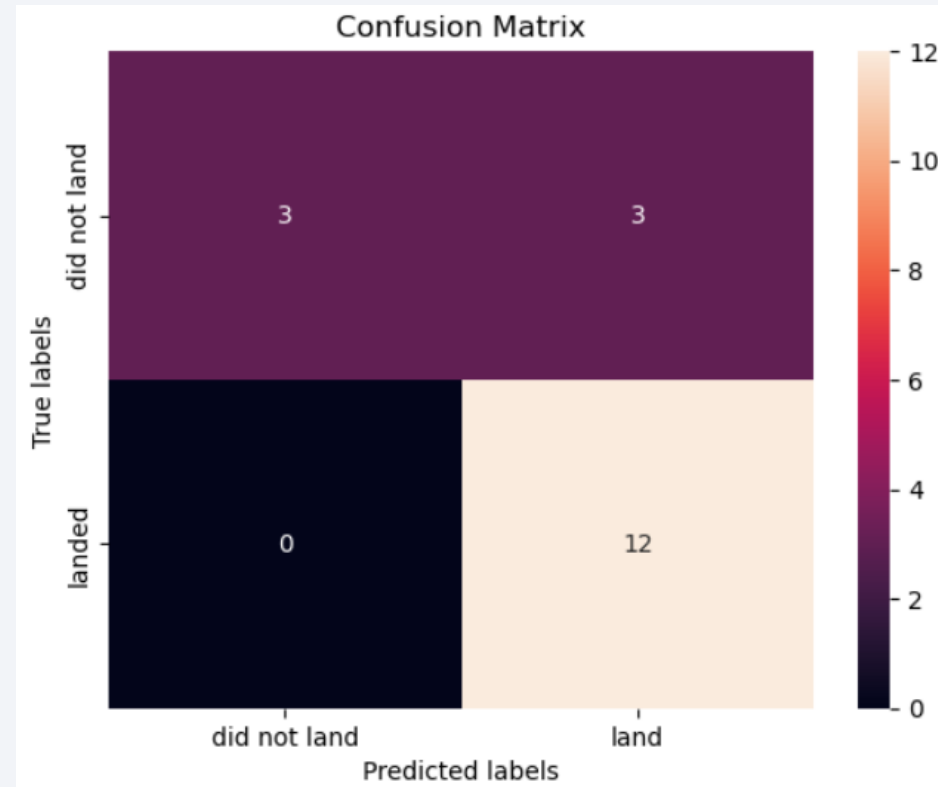
---



- The Decision Tree model has the highest classification accuracy

# Confusion Matrix

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- The major problem is **false positives**
- 3 unsuccessful landing predicted as successful by the classification model



# Conclusions

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- The success rate kept increasing since 2013 till 2020
- Rockets sent to orbits ES-L1, GEO, HEO, SSO and VLEO have high success rate
- KSC LC-39A has the highest success rate among launch sites
- The booster version B4 provides good results for light, medium, and heavy payloads
- The Decision Tree classifier performs slightly better on training set, but all models perform the same on unseen data (test set)
- The accuracy of the classification model is 83.33%

Thank you!

