



IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

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05/09/2023



Outline

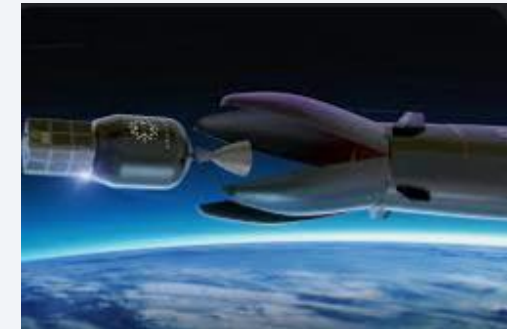
- Executive Summary
- Introduction
- Methodology
- Results
- Conclusion
- Appendix

Executive Summary

- Summary of methodologies:
 - Data collection from CVS, SQL and web (webscrapping).
 - Data wrangling: dealing with missing values, standardize values.
 - Exploratory Data Analysis (EDA):
 - Visualization methods like plotting scatter plots, bar charts, interactive charts.
 - Predictive analysis: Calssification.
- Summary of all results
 - Success of mission based on mass, launch site, orbit, etc.
 - From 2013 to 202 the missions' outcomes improved.

Introduction

- The commercial space age is here, companies are making space travel affordable for everyone: Virgin Galactic is providing suborbital spaceflights. Rocket Lab is a small satellite provider. Blue Origin manufactures sub-orbital and orbital reusable rockets.
- Perhaps the most successful is **SpaceX**.
 - SpaceX's accomplishments: Sending spacecraft to the International Space Station.
 - Differentiation: rocket launches are relatively inexpensive (reuse the first stage). SpaceX advertises Falcon 9 rocket launches cost 62 million dollars; other providers cost upwards of 165 million dollars each. Therefore, if we can determine if the first stage will land, we can determine the cost of a launch. SpaceX's Falcon 9 launch like regular rockets



Introduction

- Study will be based on another company called “SpaceY” and below are the problems to find answers to:
 - Determine the price of each lunch.
 - Determine if SpaceX will reuse the first stage.
 - Determine if Falcon’s 9 second stage will land.
 - Explain how to choose an optimal launch site.



Section 1

Methodology

Methodology

Executive Summary

- Data collection methodology:
 - Data was collected from an API and loaded into coding environment (Jupyter Notebook) using Python language and several libraries such as Pandas, Numpy, Matplotlib, Seaborn, Sci-kit Learn, etc.
- Perform data wrangling
 - Data was processed by the following general steps:
 - Identify percentage of missing values and fill them with mean values from rest of data.
 - Identify type of data for each occurrence.
 - Calculate several key information (i.e.: number of launches per site, number and occurrence of each orbit, etc.).

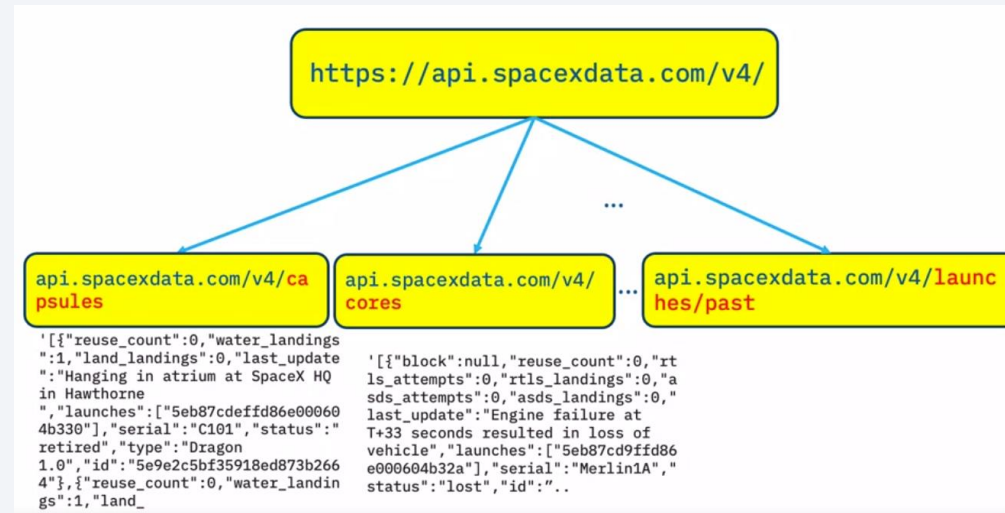
Methodology

Executive Summary

- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Standardize data in order to get valuable insights.
 - Split data into training and testing subsets.
 - Train and evaluate accuracy of different classification models.

Data Collection

- Data sets were collected via an endpoint or URL from REST API from SpaceX with launch data and from a specific web page via web-scraping methods.
- This endpoint is targeted via Python commands and loaded into the coding environment as a csv (comma separated values) file and then converted to a table that this language can manage.

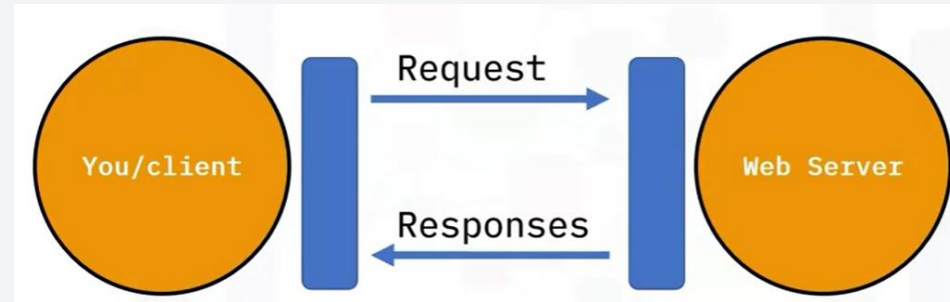


Data Collection – SpaceX API

- Basically a series of communication standards and functions from libraries are structured and performed to retrieve data from a server or database into the Jupyter notebook where the data can be filtered, structured and calculations can be made.
- Below is the link to the several steps for this study:

https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation

Basic diagram of REST API call.



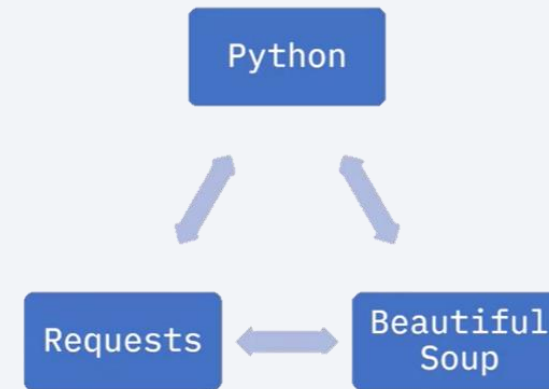
Basic diagram of database call.



Data Collection - Scraping

- Basically a series of communication standards and functions from libraries are structured and performed to retrieve data from a webpage (mainly tables) into the Jupyter notebook where the data can be filtered, structured and calculations can be made.
- Below is the link to the several steps for this study:

[https://github.com/JulianCarroVerdia/IBM
Data Science Capstone Presentation](https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation)



Data Wrangling

- The main techniques for processing data were:
 - Replacing missing values in columns with mean of the rest of the data from that column.
 - Create a new column or list with binary values: 0 for one specific case and 1 for the rest of the cases.
- Below is the link to the several steps for this study:

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EDA with Data Visualization

- The following charts were used to see if there is a correlation between those variables and with that information, estimate outcomes:
 - Flight Number vs. Payload Mass in a scatter plot and overlay the outcome of the launch.
 - Flight Number vs. Launch Site in a scatter plot and overlay the outcome of the launch.
 - Payload Mass vs. Launch Site in a scatter plot and overlay the outcome of the launch.
 - Flight Number vs. Orbit in a scatter plot and overlay the outcome of the launch.
 - Payload Mass vs. Orbit in a scatter plot and overlay the outcome of the launch.
 - Success rate over the years in a line plot.
- Below is the link to the several steps for this study:

[https://github.com/JulianCarroVerdia/IBM Data Science Capstone Presentation](https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation)

EDA with SQL

- Below is the summary of the SQL queries performed:
 - `select distinct "Launch_Site" from SPACEXTBL`
 - `SELECT * FROM SPACEXTBL WHERE "Launch_Site" LIKE 'CCA%' LIMIT 20`
 - `SELECT SUM("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE Customer="NASA (CRS)"`
 - `SELECT AVG("PAYLOAD_MASS__KG_") FROM SPACEXTBL WHERE "Booster_Version" LIKE 'F9 v1.1%'`
 - `SELECT MIN(Date) FROM SPACEXTBL WHERE "Landing _Outcome"="Success (ground pad)"`
 - `SELECT distinct "Booster_Version" FROM SPACEXTBL WHERE "Landing _Outcome"="Success (drone ship)" AND "PAYLOAD_MASS__KG_">4000 AND "PAYLOAD_MASS__KG_"<6000`
 - `SELECT "Mission_Outcome", count(*) as total_outcomes FROM SPACEXTBL GROUP BY "Mission_Outcome"`
 - `SELECT DISTINCT"Booster_Version" FROM SPACEXTBL where (select MAX(PAYLOAD_MASS__KG_) from SPACEXTBL)`

EDA with SQL

- SELECT

CASE

WHEN substr(Date, 4, 2) = '01' THEN 'January'

WHEN substr(Date, 4, 2) = '02' THEN 'February'

WHEN substr(Date, 4, 2) = '03' THEN 'March'

WHEN substr(Date, 4, 2) = '04' THEN 'April'

WHEN substr(Date, 4, 2) = '05' THEN 'May'

WHEN substr(Date, 4, 2) = '06' THEN 'June'

WHEN substr(Date, 4, 2) = '07' THEN 'July'

WHEN substr(Date, 4, 2) = '08' THEN 'August'

WHEN substr(Date, 4, 2) = '09' THEN 'September'

WHEN substr(Date, 4, 2) = '10' THEN 'October'

WHEN substr(Date, 4, 2) = '11' THEN 'November'

WHEN substr(Date, 4, 2) = '12' THEN 'December'

END as month_name,

EDA with SQL

```
"Landing _Outcome",
```

```
"Booster_Version",
```

```
"Launch_Site"
```

```
FROM SPACEXTBL
```

```
WHERE substr(Date,7,4)='2015' AND "Landing _Outcome" LIKE '%Failure%' AND "Landing _Outcome" LIKE '%(drone ship)%'
```

- ```
SELECT * COUNT(*) as successful_landings
```

```
FROM SPACEXTBL
```

```
WHERE "Landing _Outcome" LIKE '%Success%' AND "Date" BETWEEN '2010-06-04' AND '2017-03-20'
```

```
GROUP BY "Landing _Outcome"
```

```
ORDER BY successful_landings DESC
```

- Below is the link to the several steps for this study:

[https://github.com/JulianCarroVerdia/IBM\\_Data\\_Science\\_Capstone\\_Presentation](https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation)



# Build an Interactive Map with Folium

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- Below are the map objects such as markers, circles, lines, etc. created and added to the folium map:
  - Circle.
  - Marker.
  - Marker cluster.
  - Mouse position.
  - Polyline.
- These objects were added in order to get insights from the map from the launches such as on a mouse position get the coordinates quickly, having straightforward information over the map like success launch, and having distance to main railroads, coastlines, etc.
- Below is the link to the several steps for this study:

[https://github.com/JulianCarroVerdia/IBM Data Science Capstone Presentation](https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation)

# Build a Dashboard with Plotly Dash

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- Below is the summary of the plots/graphs and interactions added to a dashboard:
  - Launch site drop-down input component.
  - Pie chart with a callback function based on selected site from drop-down input component.
  - Range slider to select payload mass (kg).
  - Scatter plot success vs. payload.
- These graphs and interactions were added in order to enable stakeholders to explore and manipulate data in an interactive and real-time way. In this way, instead of presenting findings in static graphs, interactive data visualization can always tell a more appealing story.
- Below is the link to the several steps for this study:

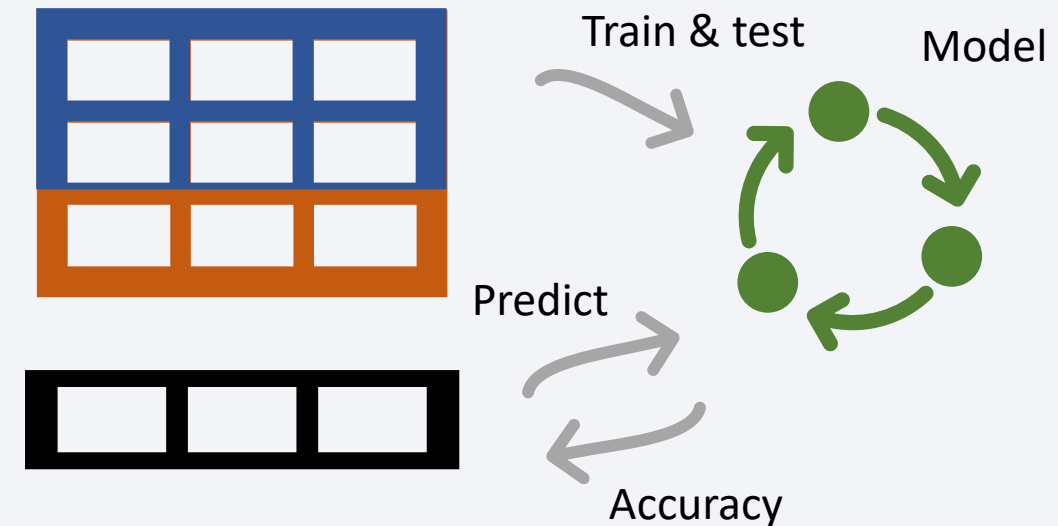
[https://github.com/JulianCarroVerdia/IBM\\_Data\\_Science\\_Capstone\\_Presentation](https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation)

# Predictive Analysis (Classification)

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- Below is the summary building, evaluating, improvement and finding the best performing classification model:
  - Creation of Numpy array from selected columns for performing model testing and evaluation.
  - Standardize data for the range to be consistent.
  - Split dataset into test and train subsets.
  - Create objects for each model to be fitted and getting the accuracy for each one.
- Below is the link to the several steps for this study:  
[https://github.com/JulianCarroVerdia/IBM Data Science Capstone Presentation](https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation)

Dataset Split: test & train



# Results

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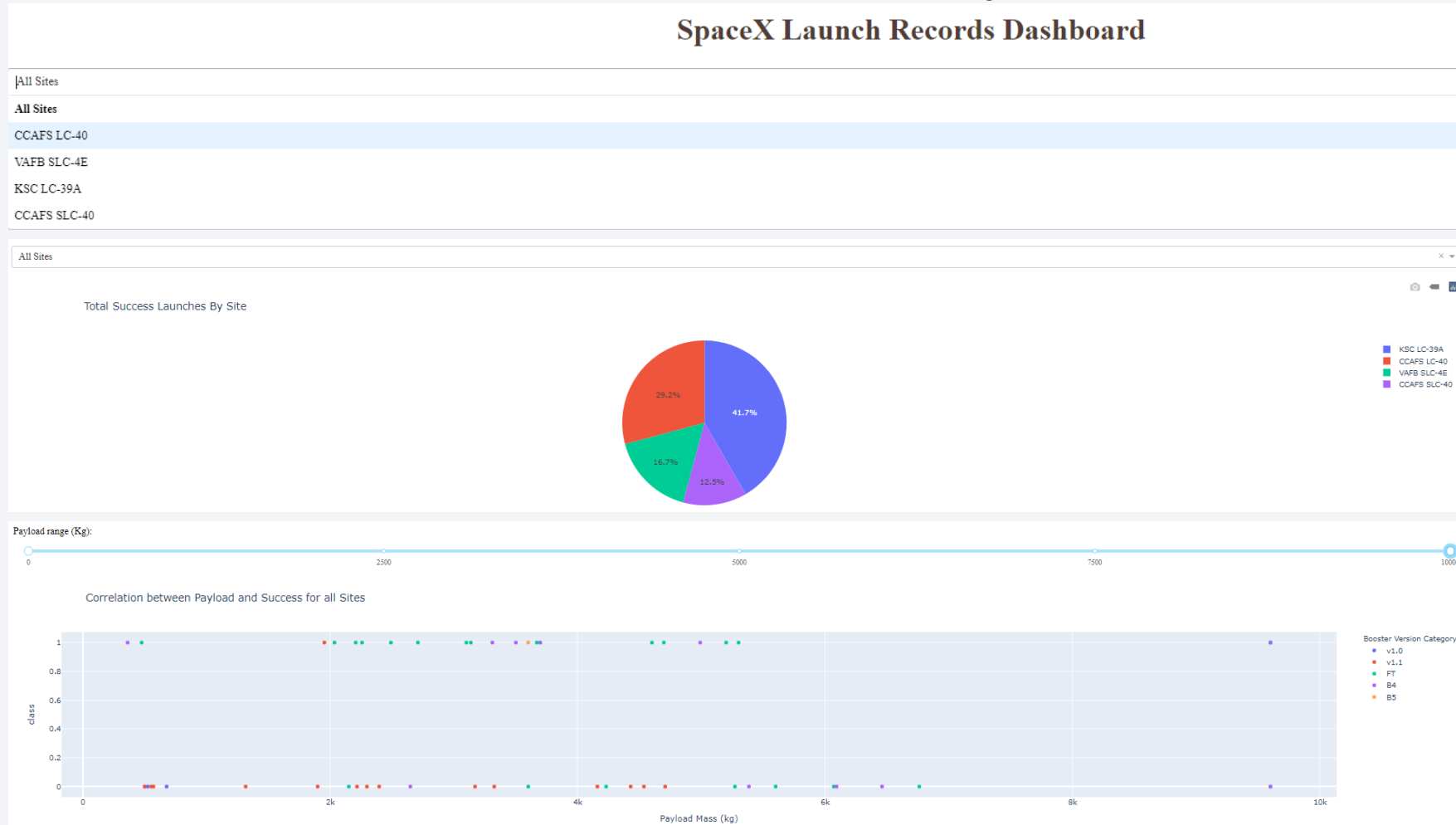
- By analyzing the information in the exploratory data analysis, we see the following results:
  - As the flight number increases, the first stage is more likely to land successfully. The payload mass is also important; it seems the more massive the payload, the less likely the first stage will return.
  - For the VAFB-SLC launch site there are no rockets launched for heavy payload mass (greater than 10000) and generally speaking for this heavy payload mass there is a high success rate.
  - With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS. However for GTO we cannot distinguish this well as both positive landing rate and negative landing (unsuccessful mission) are both there here.
  - The success rate since 2013 kept increasing till 2020.
  - There was a very high success rate for all missions. See table →

| Mission_Outcome                  | total_outcomes |
|----------------------------------|----------------|
| Failure (in flight)              | 1              |
| Success                          | 98             |
| Success                          | 1              |
| Success (payload status unclear) | 1              |



# Results

- Below are some screenshots from the interactive analytics:



# Results

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- Below are the predictive analysis results:
  - Overall all methods testes for making a predictive analysis had a relatively high accuracy (>83%), except for the decision tree methos which perfomed with a 72% of accuracy.



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. A faint, light blue grid pattern is also visible, particularly in the lower-left quadrant. The overall effect is dynamic and technological.

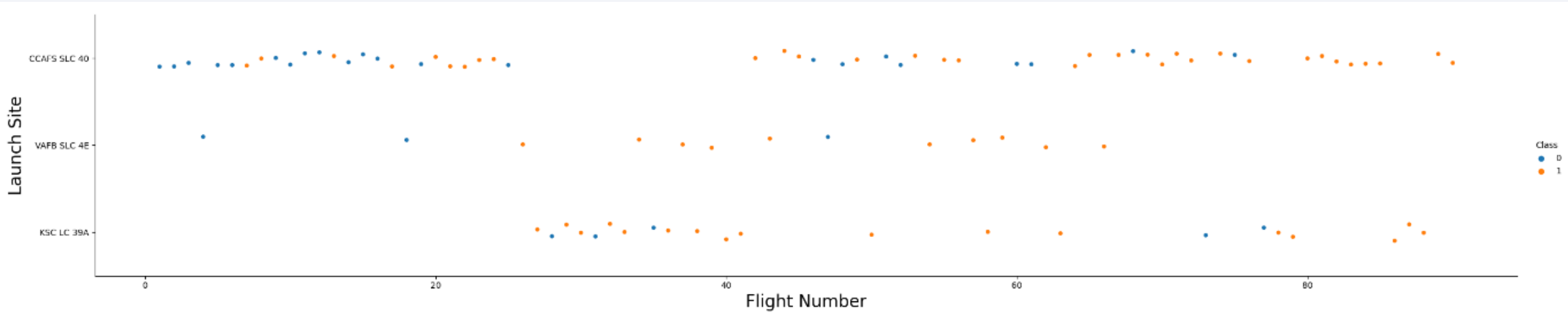
Section 2

# Insights drawn from EDA



# Flight Number vs. Launch Site

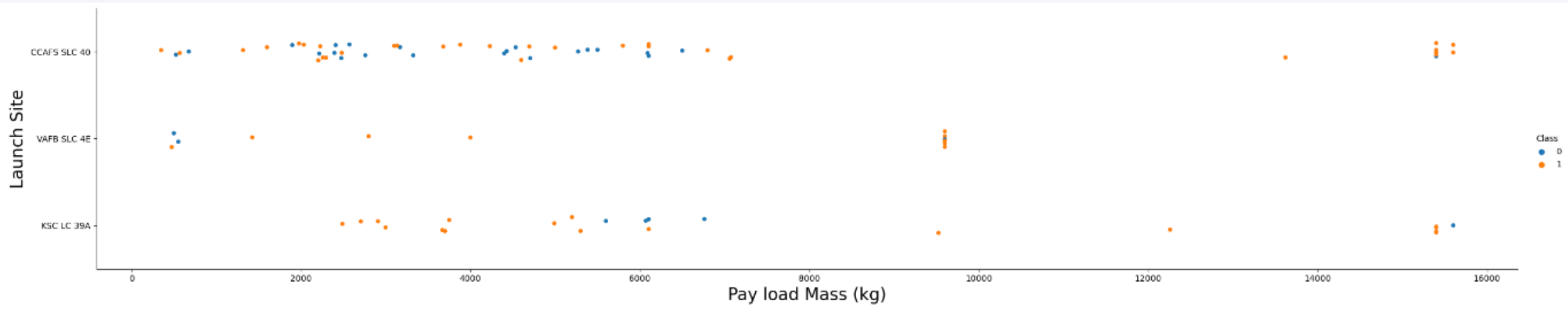
- As it can be seen from the plot, the highest succes rate was from the launch site VAFB SLC 4E, followed by the KSC LC 39A and last CCAFS SLC 40. (Class 0 = failure, class 1 = success).





# Payload vs. Launch Site

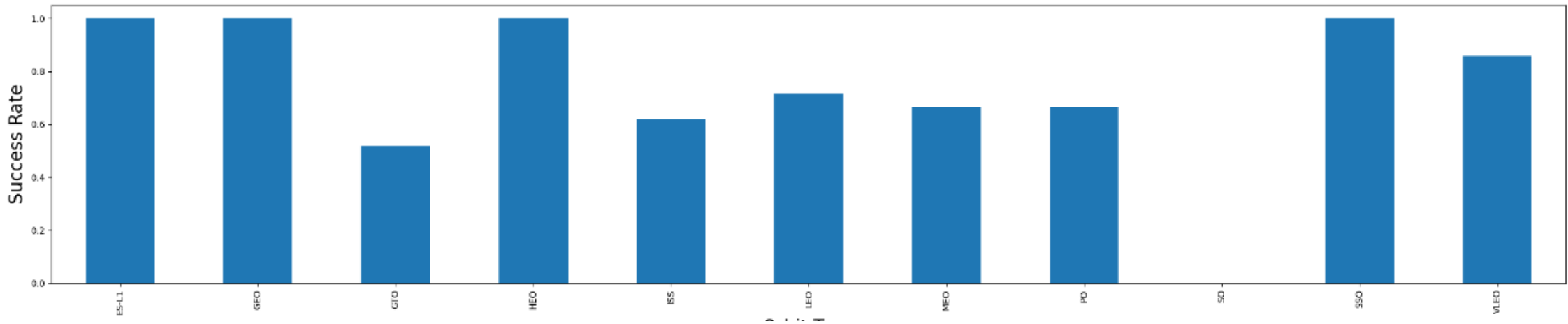
- As it can be seen from the plot, generally speaking for big payload mass the success rate was bigger. (Class 0 = failure, class 1 = success).



# Success Rate vs. Orbit Type

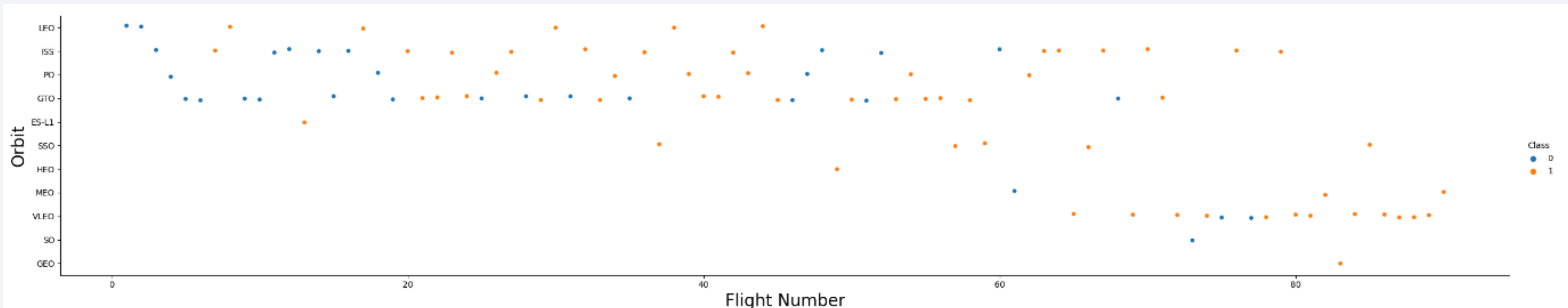
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- As it can be seen from the plot, the orbits ES-L1, GFO, HEO, SSO and VLEO had the best success rate.



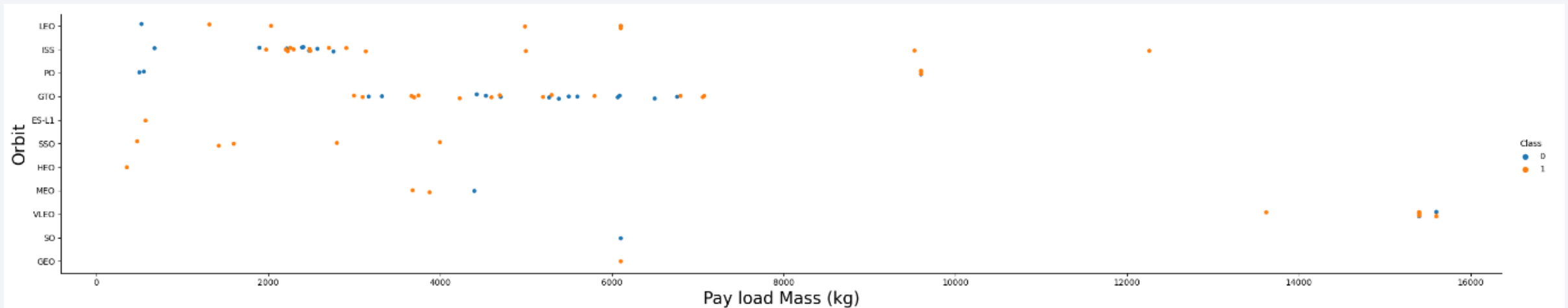
# Flight Number vs. Orbit Type

- As it can be seen from the plot, the LEO orbit the Success appears related to the number of flights; on the other hand, there seems to be no relationship between flight number when in GTO orbit.



# Payload vs. Orbit Type

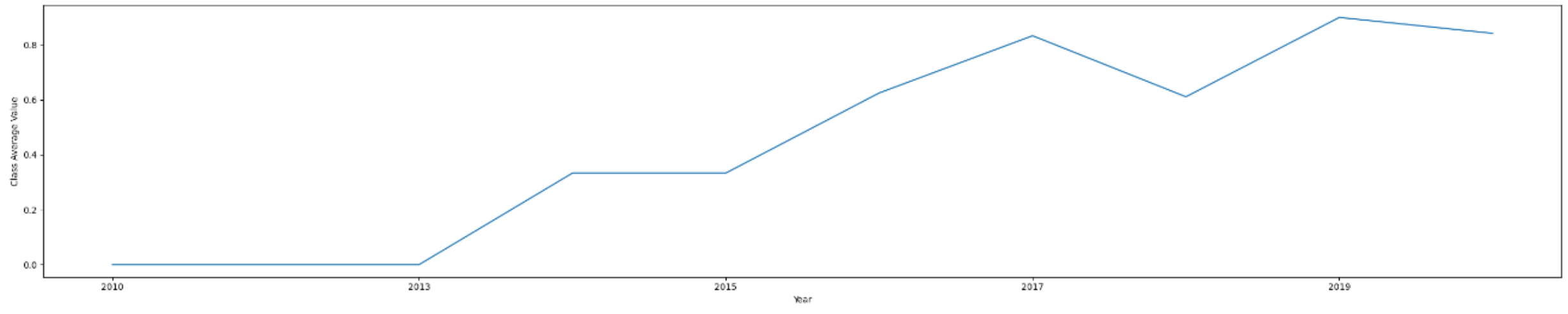
- As it can be seen from the plot, with heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS. However for GTO we cannot distinguish this well as both positive landing rate and negative landing (unsuccessful mission) are both there here.



# Launch Success Yearly Trend

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- As it can be seen from the plot, the success rate since 2013 kept increasing till 2020



# All Launch Site Names

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- The unique launch site in the space missions were retrieved with a SQL query to get these unique values, where the column Launch\_Site is selected and a function named DISTINCT is applied to that column to return only unique values.
- Query: %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'

- This query is used to retrieve all columns from SQL database and filter the column Launch\_Site where the names begin with CCA by applying LIKE 'CCA%'.
- Query: %sql SELECT \* FROM SPACEXTBL WHERE "Launch\_Site" LIKE 'CCA%' LIMIT 5

| Date       | Time (UTC) | Booster_Version | Launch_Site | Payload                                                       | PAYLOAD_MASS_KG_ | Orbit     | Customer        | Mission_Outcome | Landing_Outcome     |
|------------|------------|-----------------|-------------|---------------------------------------------------------------|------------------|-----------|-----------------|-----------------|---------------------|
| 04-06-2010 | 18:45:00   | F9 v1.0 B0003   | CCAFS LC-40 | Dragon Spacecraft Qualification Unit                          | 0                | LEO       | SpaceX          | Success         | Failure (parachute) |
| 08-12-2010 | 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) |
| 22-05-2012 | 07:44:00   | F9 v1.0 B0005   | CCAFS LC-40 | Dragon demo flight C2                                         | 525              | LEO (ISS) | NASA (COTS)     | Success         | No attempt          |
| 08-10-2012 | 00:35:00   | F9 v1.0 B0006   | CCAFS LC-40 | SpaceX CRS-1                                                  | 500              | LEO (ISS) | NASA (CRS)      | Success         | No attempt          |
| 01-03-2013 | 15:10:00   | F9 v1.0 B0007   | CCAFS LC-40 | SpaceX CRS-2                                                  | 677              | LEO (ISS) | NASA (CRS)      | Success         | No attempt          |



# Total Payload Mass

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- The query uses SUM function to sum all values from the column PAYLOAD\_MASS\_\_KG\_ where customer = NASA (CRS).
- Query: %sql SELECT SUM("PAYLOAD\_MASS\_\_KG\_") FROM SPACEXTBL WHERE Customer="NASA (CRS)"

| SUM("PAYLOAD_MASS__KG_") |
|--------------------------|
| 45596                    |

# Average Payload Mass by F9 v1.1

---

- The query uses AVG function to calculate the average of all values from the column PAYLOAD\_MASS\_KG\_ where Booster\_Version has the characters F9 v.1.1.
- Query: %sql SELECT AVG("PAYLOAD\_MASS\_KG\_") FROM SPACEXTBL WHERE "Booster\_Version" LIKE 'F9 v1.1%'

```
AVG("PAYLOAD_MASS_KG_")
```

```
2534.66666666666665
```

# First Successful Ground Landing Date

---

- The query uses MIN function to calculate the minimum value from the column Date (first date) where Outcome = Success (ground pad).
- Query: `%sql SELECT MIN(Date) FROM SPACEXTBL WHERE "Landing_Outcome"="Success (ground pad)"`

| <b>MIN(Date)</b> |
|------------------|
| 01-05-2017       |

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

---

- The query uses DISTINCT function retrieve unique values from Booster\_Version column, with Landing\_Outcome=Success and Payload between 4000 and 6000.
- Query: %sql SELECT distinct "Booster\_Version" FROM SPACEXTBL WHERE "Landing\_Outcome"="Success (drone ship)" AND "PAYLOAD\_MASS\_\_KG\_">4000 AND "PAYLOAD\_MASS\_\_KG\_"<6000

| 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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- The query selects the Mission outcome column and adds a column where it fills with the count of each case of mission outcome.
- Query: %sql SELECT "Mission\_Outcome", count(\*) as total\_outcomes FROM SPACEXTBL GROUP BY "Mission\_Outcome"

| Mission_Outcome                  | total_outcomes |
|----------------------------------|----------------|
| Failure (in flight)              | 1              |
| Success                          | 98             |
| Success                          | 1              |
| Success (payload status unclear) | 1              |

# Boosters Carried Maximum Payload

- The query selects unique values from Booster Version column and then uses a sub-query to select the max Payload Mass. (The table is very long, so a portion of it was included).
- Query: %sql SELECT DISTINCT "Booster\_Version" FROM SPACEXTBL where (select MAX(PAYLOAD\_MASS\_\_KG\_) from SPACEXTBL)

| Booster_Version |
|-----------------|
| F9 v1.0 B0003   |
| F9 v1.0 B0004   |
| F9 v1.0 B0005   |
| F9 v1.0 B0006   |
| F9 v1.0 B0007   |
| F9 v1.1 B1003   |
| F9 v1.1         |
| F9 v1.1 B1011   |
| F9 v1.1 B1010   |
| F9 v1.1 B1012   |
| F9 v1.1 B1013   |
| F9 v1.1 B1014   |

# 2015 Launch Records

- The query uses a CASE function to translate number-type month date into letter-type month date, and filter: 2015 year failure launches, their booster version and launch sites.
- Query:

```
%%sql
SELECT
 CASE
 WHEN substr(Date, 4, 2) = '01' THEN 'January'
 WHEN substr(Date, 4, 2) = '02' THEN 'February'
 WHEN substr(Date, 4, 2) = '03' THEN 'March'
 WHEN substr(Date, 4, 2) = '04' THEN 'April'
 WHEN substr(Date, 4, 2) = '05' THEN 'May'
 WHEN substr(Date, 4, 2) = '06' THEN 'June'
 WHEN substr(Date, 4, 2) = '07' THEN 'July'
 WHEN substr(Date, 4, 2) = '08' THEN 'August'
 WHEN substr(Date, 4, 2) = '09' THEN 'September'
 WHEN substr(Date, 4, 2) = '10' THEN 'October'
 WHEN substr(Date, 4, 2) = '11' THEN 'November'
 WHEN substr(Date, 4, 2) = '12' THEN 'December'
 END as month_name,
 "Landing_Outcome",
 "Booster_Version",
 "Launch_Site"
FROM SPACEXTBL
WHERE substr(Date,7,4)='2015' AND "Landing_Outcome" LIKE '%Failure%' AND "Landing_Outcome" LIKE '%(drone ship)%'
```

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



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

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- The query selects the Landing\_Outcome column and counts the cases for the specified date range, groups it by Landing\_Outcome case and orders it by what was counted.
- Query: % %sql

```
SELECT "Landing _Outcome", COUNT(*) as 'Quantity' FROM SPACEXTBL WHERE
DATE BETWEEN '04-06-2010' AND '20-03-2017' GROUP BY "Landing _Outcome"
ORDER BY "Quantity"
```

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

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.

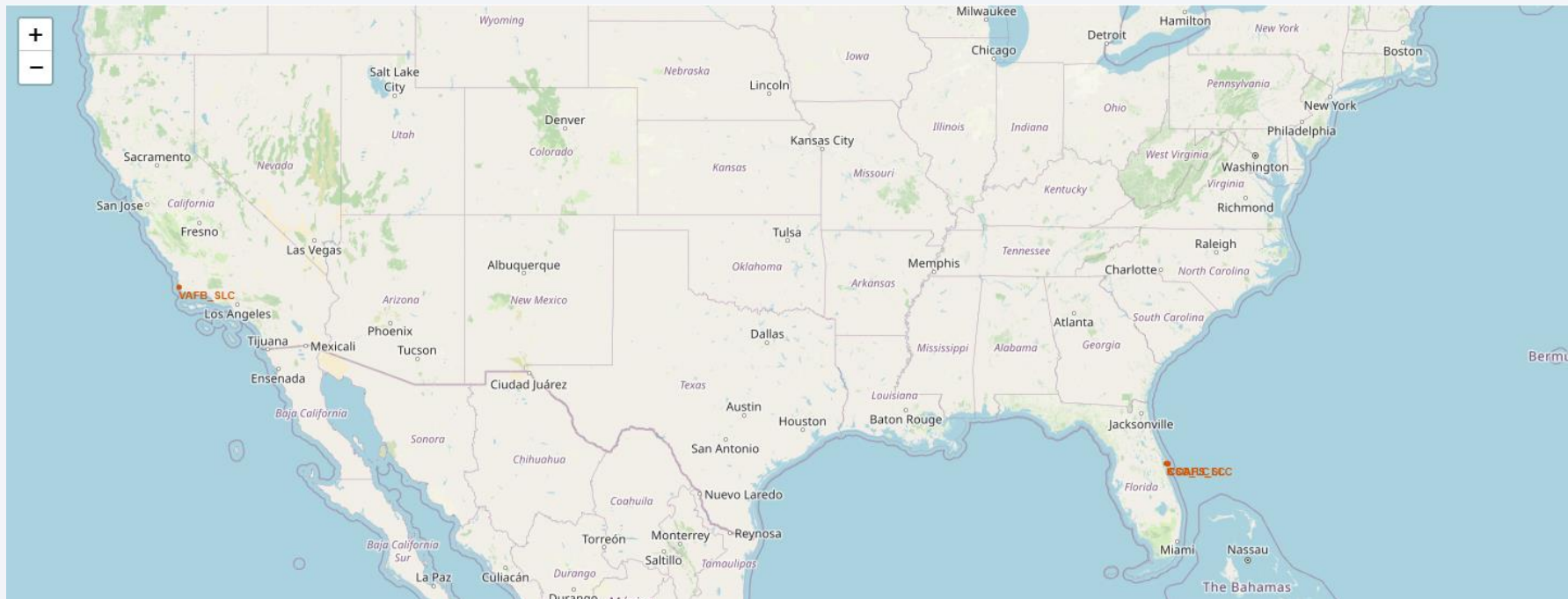
Section 3

# Launch Sites Proximities Analysis

# Map with markers and circles.

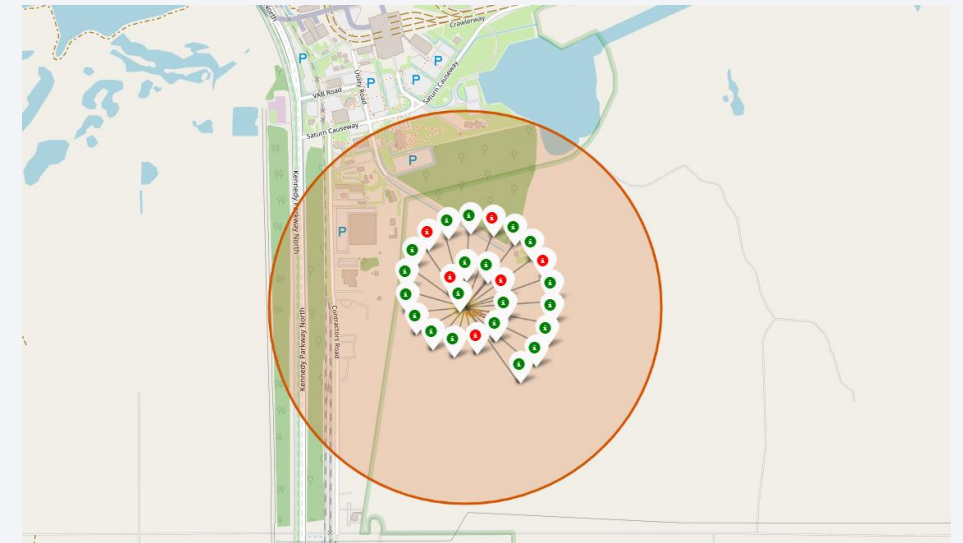
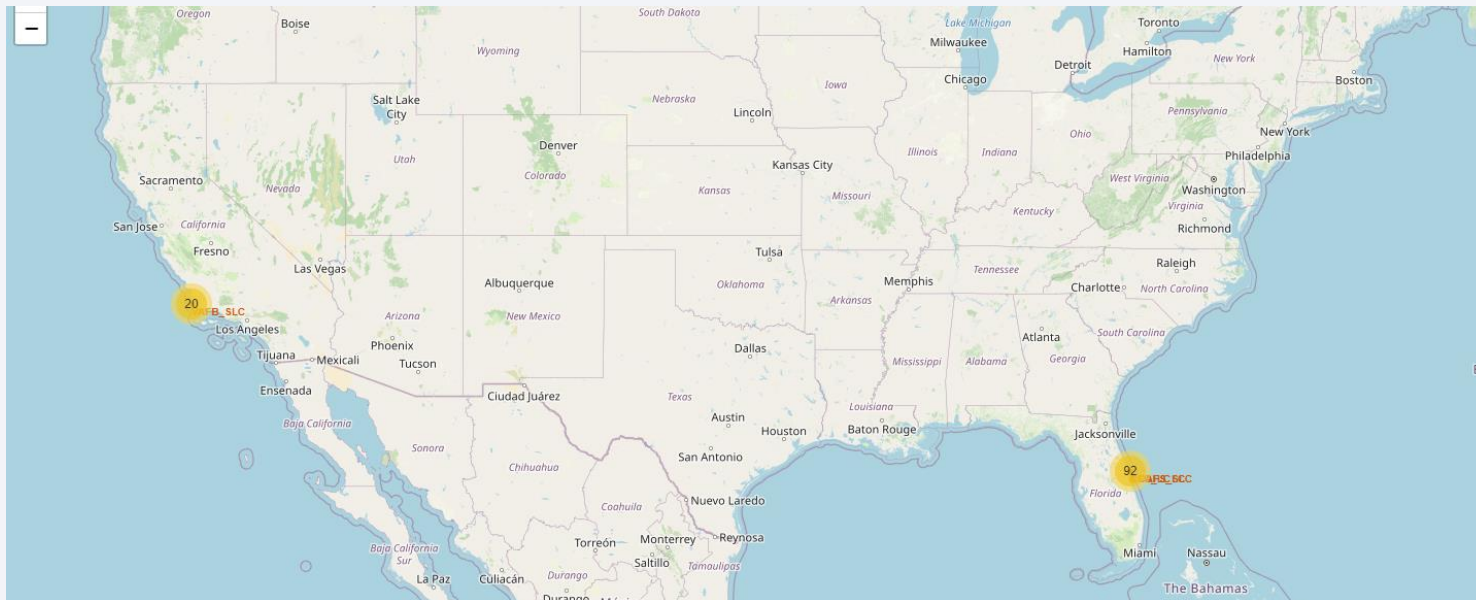
---

- In the following screenshot of the map, a first layer of circles and markers were added to it. This is important in order to detect how close each launch site is from each other.



# Map with markers, circles with success outcome.

- In the following screenshot of the map, an additional layer was added to display the quantity of successful launch outcomes. This is important in order to detect areas where successful outcomes were positive.

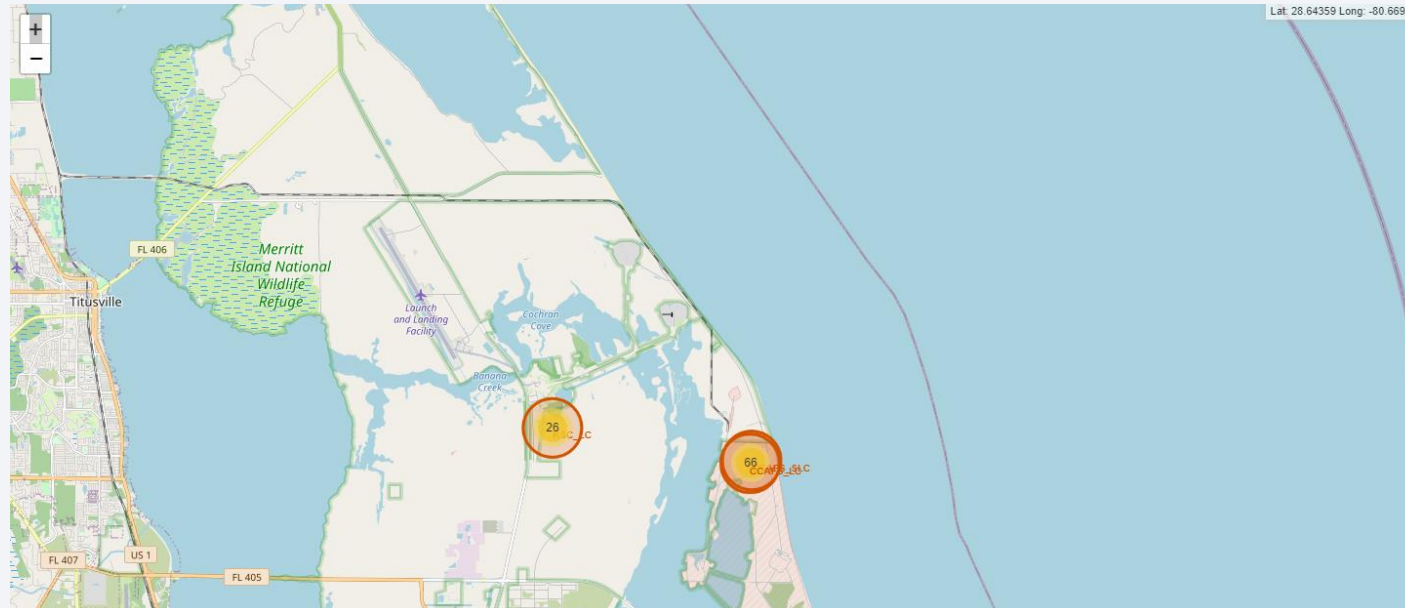




## Map with markers, circles with success outcome and coordinates.

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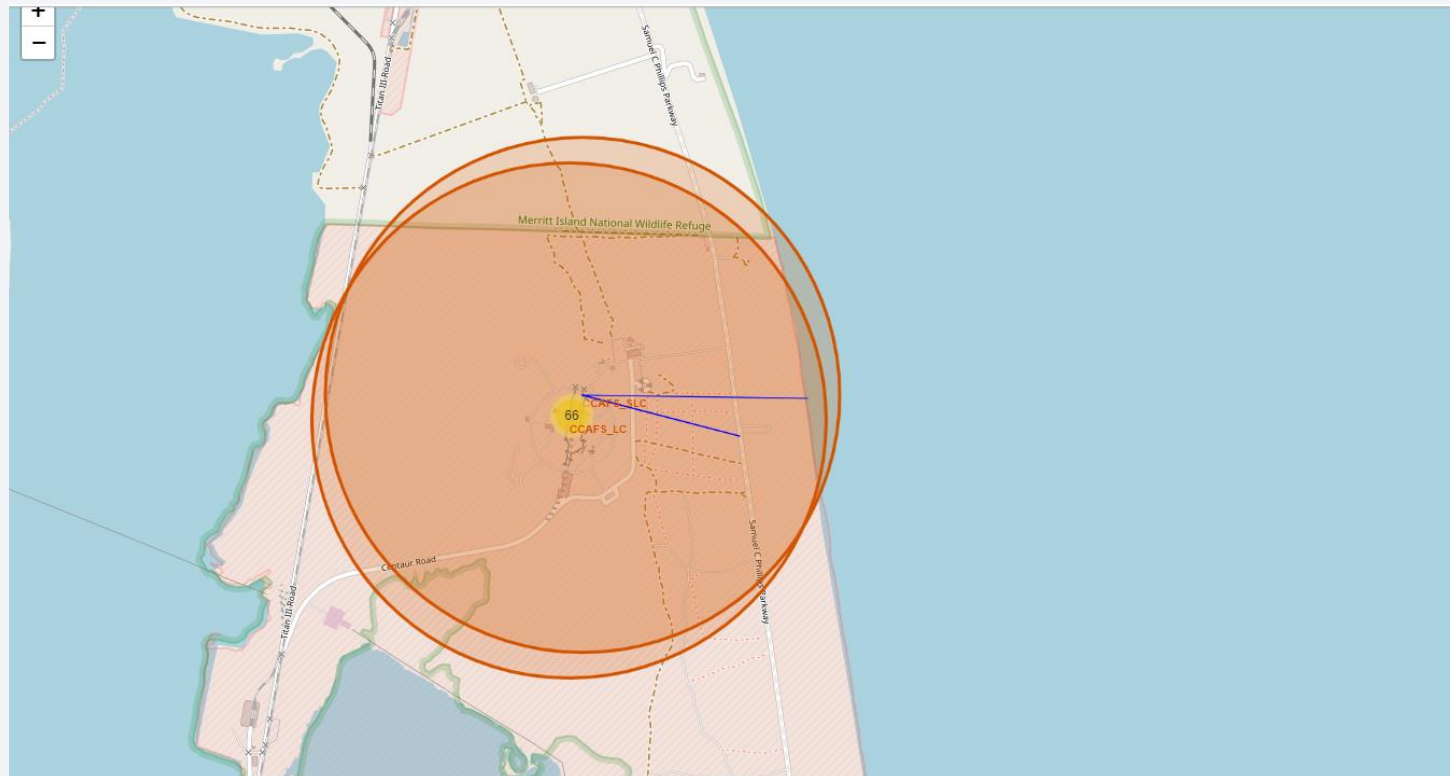
- This map is the same as the one in the previous slide but it has a functionality that it indicates the coordinates as you move the mouse over the map for giving the instantaneous coordinate ion a specific point if the map.



# Map with distance lines

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- The last map has all the functionalities from previous maps but lines with distances are drawn over it that show how long a line is from one point to another to gain insightful information whether if a launch site is close to a road, railroad, shore line, etc.





Section 4

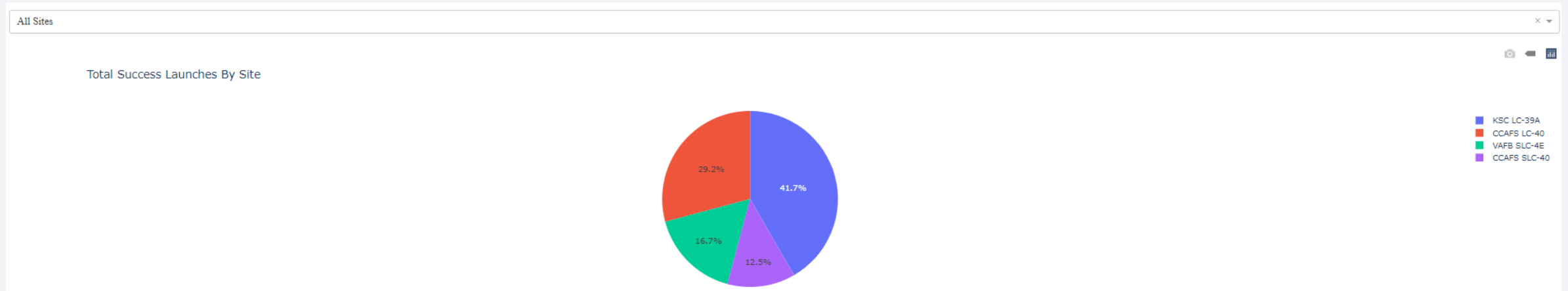
# Build a Dashboard with Plotly Dash



# Success Launch for all sites

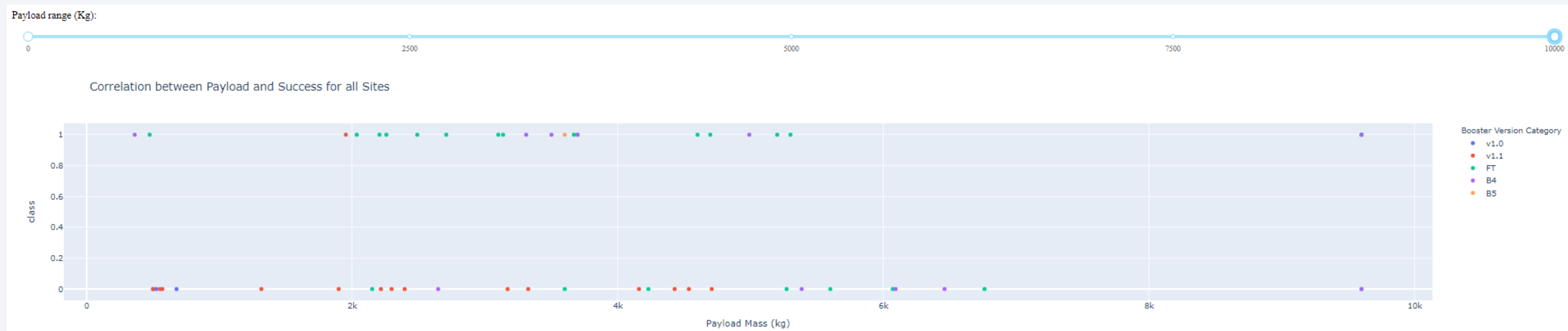
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- This pie chart clearly shows which site has the highest and lowest success rates, indicating their percentages.



# Payload vs. Launch Outcome scatter plot

- This scatter plot shows the correlation between Payload mass and launch outcome. The range slider is very useful because it can filter the payload mass range for a desired one and get instant insightful information.





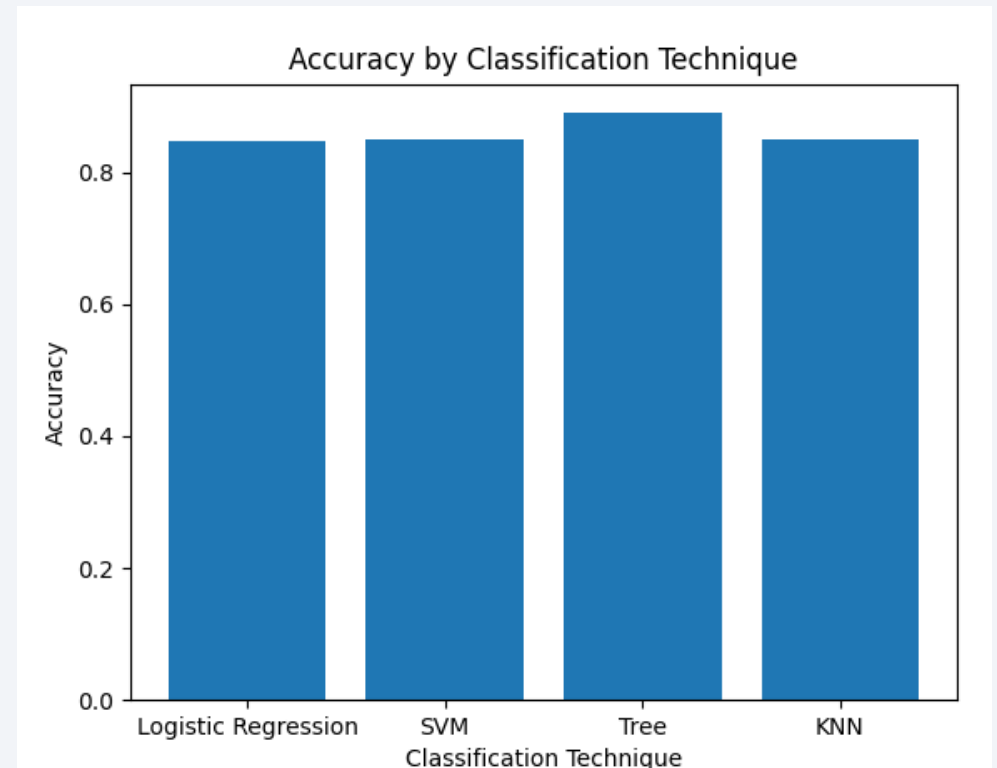
Section 5

# Predictive Analysis (Classification)

# Classification Accuracy

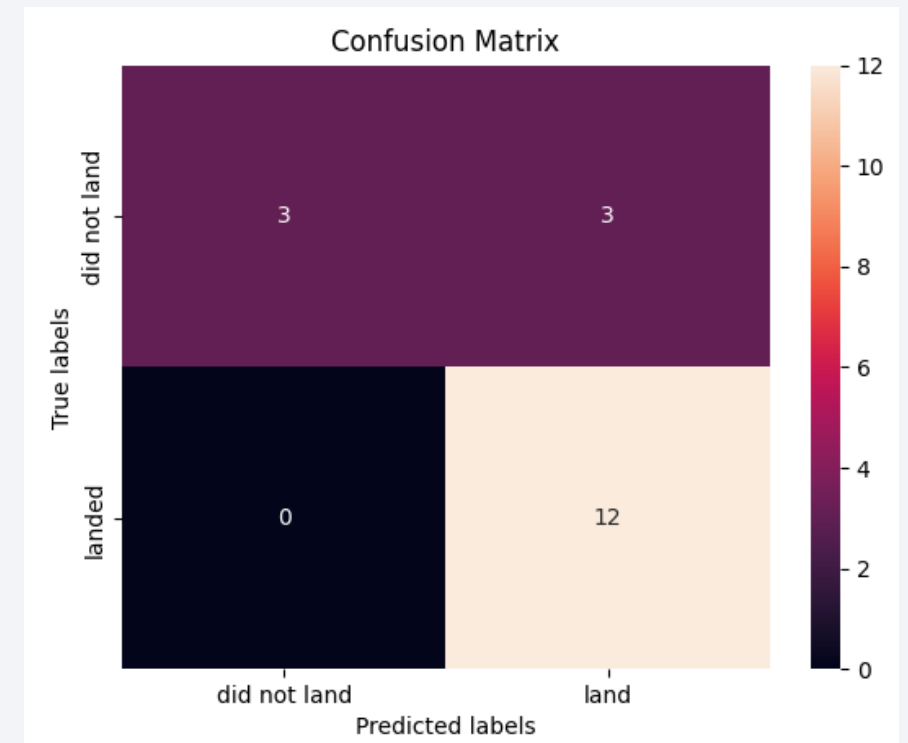
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- By comparing the accuracy of each model, we found out that the tree classification technique is the most suitable one, even though the others are not far from it.



# Confusion Matrix

- The confusion matrix for the decision tree technique is shown and it reveals that for the actual landed occurrences, the model predicted all the cases and for the not landed actual occurrences the model predicted half of them.



# Conclusions

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- As the flight number increases, the first stage is more likely to land successfully. The payload mass is also important; it seems the more massive the payload, the less likely the first stage will return.
- For the VAFB-SLC launch site there are no rockets launched for heavy payload mass(greater than 10000) and generally speaking for this heavy payload mass there is a high success rate.
- With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS. However for GTO we cannot distinguish this well as both positive landing rate and negative landing(unsuccesful mission) are both there here.
- The success rate since 2013 kept increasing till 2020.
- There was a very high success rate for all missions.
- The decision tree technique was the best technique for the model.





# Appendix

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- Below is the link to the several steps for this study:

[https://github.com/JulianCarroVerdia/IBM Data Science Capstone Presentation](https://github.com/JulianCarroVerdia/IBM_Data_Science_Capstone_Presentation)

Thank you!

