



IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

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Overview



Executive Summary



Introduction



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Appendix

Executive Summary

- Summary of methodologies
 - Data Collection through SpaceX API
 - Data Collection with Web Scraping
 - Data Wrangling
 - Exploratory Data Analysis with SQL
 - Exploratory Data Analysis with Data Visualization
 - Interactive Visual Analytics with Folium
 - Machine Learning Prediction
- Summary of all results
 - Display relationships between various features in the data
 - Visualize and analyze launch sites highlighted on interactive map
 - Predict Space X Falcon 9 First Stage Landing.

Introduction

- Project background and context

SpaceX advertises Falcon 9 rocket launches on its website, with a cost of 62 million dollars; other providers cost upward of 165 million dollars each, much of the savings is because SpaceX can reuse the first stage. Therefore if we can determine if the first stage will land, we can determine the cost of a launch. This information can be used if an alternate company wants to bid against SpaceX for a rocket launch. In this module, you will be provided with an overview of the problem and the tools you need to complete the course.

- Problems

- What factors determine a successful first-stage rocket landing?
- What relationship do these factors have with each other?
- Can we use those factors to predict whether or not a future launch will be successful?

Section 1

Methodology

Methodology

Executive Summary

- Data collection methodology:
 - Data was collected via the SpaceX API and web scraped from Wikipedia
- Perform data wrangling
 - One-hot encoding was applied to categorical features
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - Build, fit, & evaluate classification models

Data Collection Method

GET request via SpaceX API



Use **BeautifulSoup** to
web scrape Falcon 9 launches



Clean data

Convert to pandas data frame
using `.json_normalize()`

Extract as an HTML Table and parse it into
a pandas data frame

Fill in missing values

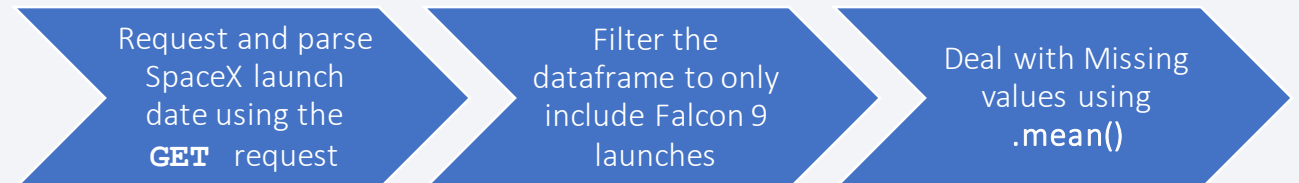
Data Collection – SpaceX API

`getBoosterVersion`

`getLaunchSite`

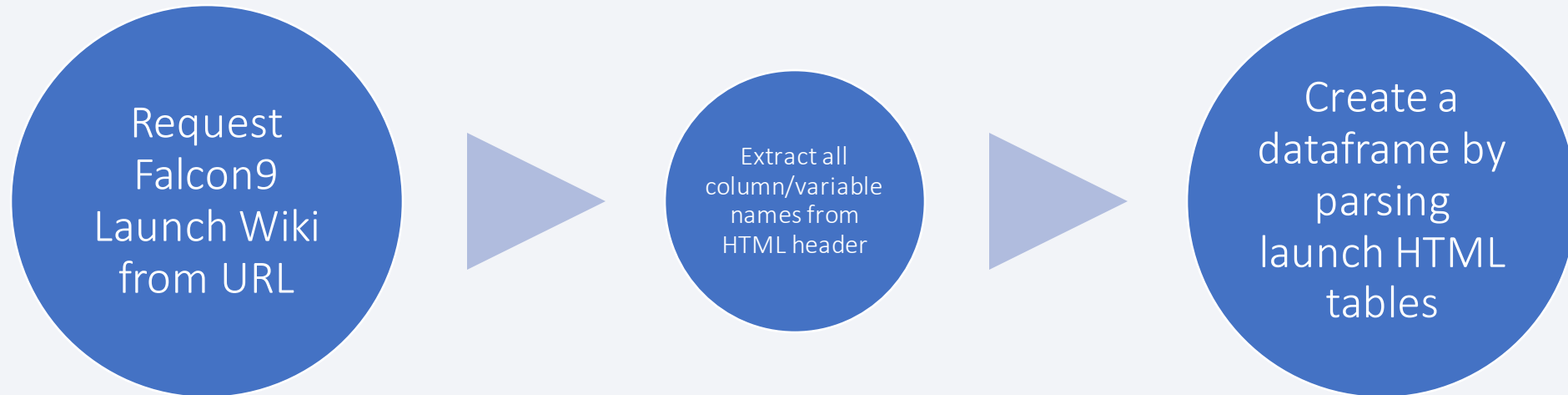
`getPayloadData`

`getCoredata`



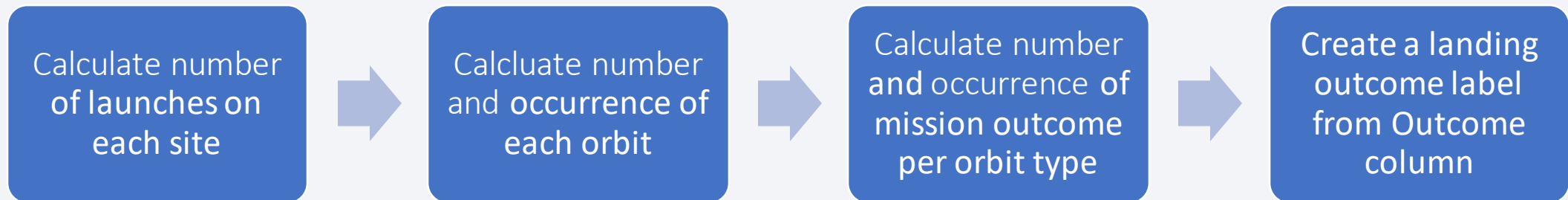
<https://github.com/phil-lee675/IBMDS0720EN/blob/main/Data%20Collection%20API.ipynb>

Data Collection - Scraping



<https://github.com/phil-lee675/IBMDS0720EN/blob/main/Data%20Collection%20with%20Web%20Scraping.ipynb>

Data Wrangling



<https://github.com/phil-lee675/IBMDS0720EN/blob/main/Data%20Wrangling.ipynb>

EDA with Data Visualization

We want to visualize the relationships between the following variables:

Flight Number vs Launch Site

Payload vs Launch Site

Success rate of each Orbit

Flight Number and Orbit type

Payload and Orbit type

Launch Success yearly trend

<https://github.com/phil-lee675/IBMDS0720EN/blob/main/EDA%20with%20Visualization.ipynb>

EDA with SQL

With SQL query commands, we want to display or list the following:

- All Launch Site Names
- Launch Site Names that begin with 'KSC'
- Total Payload Mass carried by boosters launched by NASA (CRS)
- Average Payload Mass carried by Booster F9 v1.1
- The date where the first successful landing via drone ship was achieved.
- The names of the successful boosters that have a payload mass between 4000 & 6000
- Total number of successful and failure mission outcomes
- Names of boosters which have carried max payload
- Records of successful landings for 2017
- Ranking of successful landings between the June 4th 2010 and March 3rd 2020.

<https://github.com/phil-lee675/IBMDS0720EN/blob/main/Exploratory%20Data%20Analysis%20with%20SQL.ipynb>

Build an Interactive Map with Folium

With the following markings we can find the answer to the following:

- Are launch sites in close proximity to railways?
- Are launch sites in close proximity to highways?
- Are launch sites in close proximity to coastline?
- Do launch sites keep certain distance away from cities?

Mark all launch sites on a map

- Mark the success/failed launches for each site

Calculate distances between a launch site to its proximities

- Have it follow mouse position on map

Mark the closest railway

- Draw a line between marker and launch site

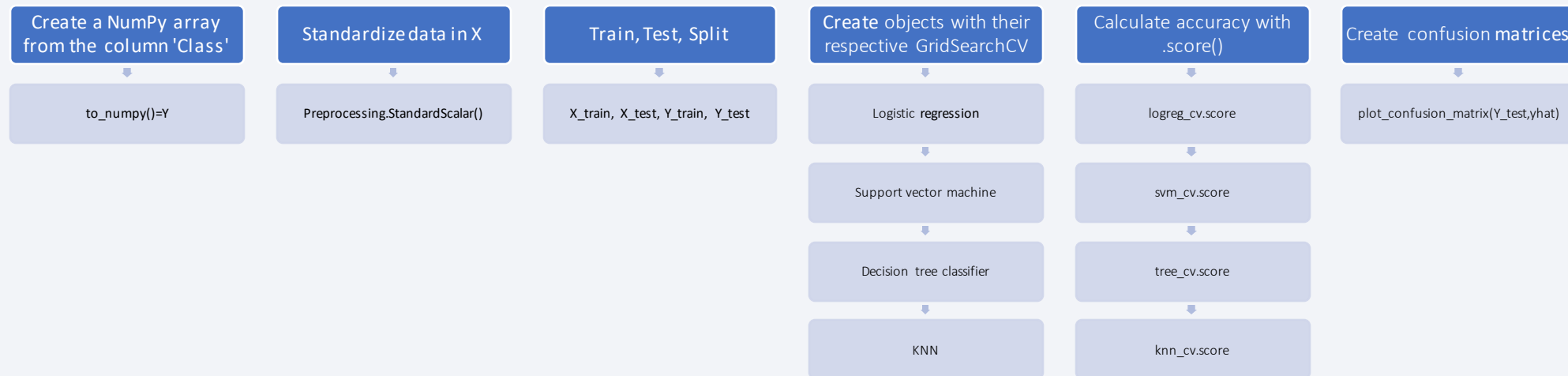
<https://github.com/phil-lee675/IBMDS0720EN/blob/main/Launch%20Site%20Locations%20Analysis%20with%20Folium.ipynb>

Build a Dashboard with Plotly Dash

- Added a dropdown list to enable Launch Site selection
- Added a pie chart to show the total successful launches count for all sites
- Added a slider to select payload range
- Added a scatter chart to show the correlation between payload and launch success

https://github.com/phil-lee675/IBMDS0720EN/blob/main/PlotlyDash/spacex_dash_app.py

Predictive Analysis (Classification)



<https://github.com/phil-lee675/IBMDS0720EN/blob/main/Machine%20Learning%20Prediction.ipynb>

Results

Exploratory data analysis results

Interactive analytics demo in
screenshots

Predictive analysis results

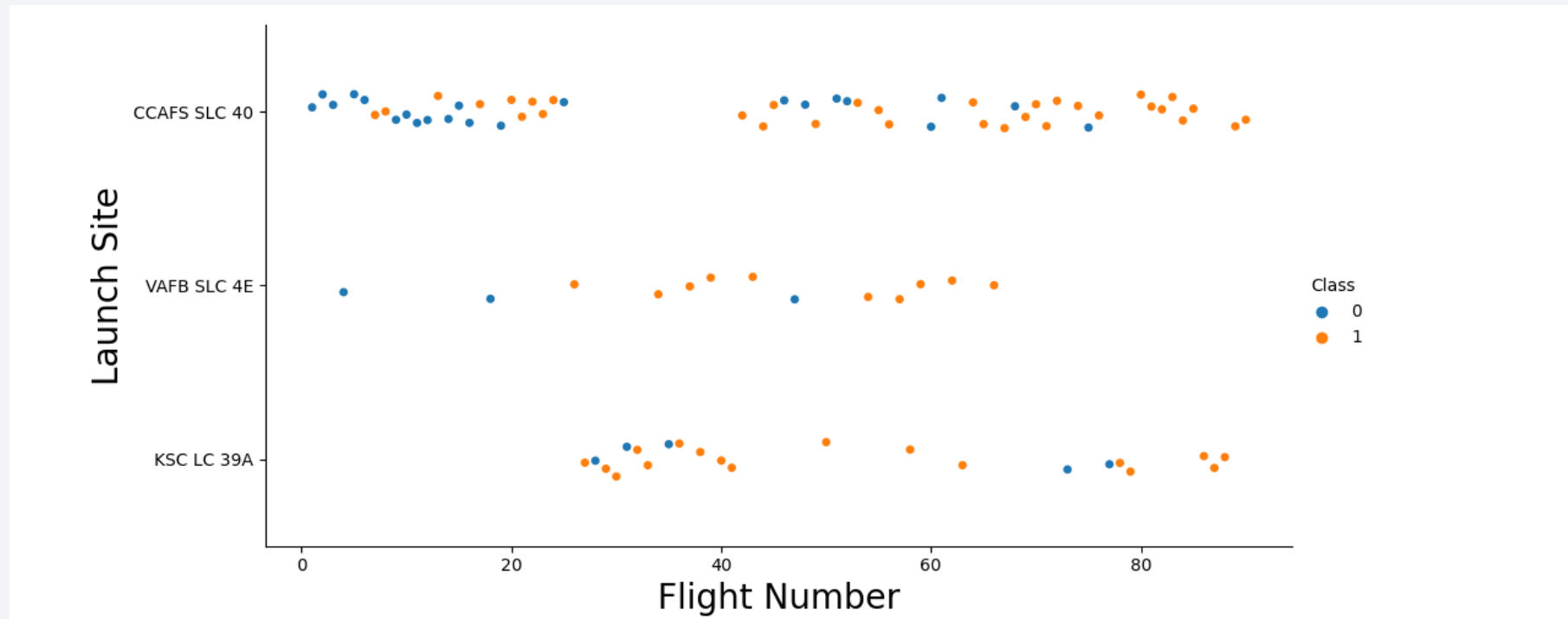
The background of the slide is an abstract composition. It features a dark blue field on the left side, which transitions into a complex pattern of diagonal streaks in shades of blue, red, and teal on the right. These streaks have a textured, almost woven appearance. Overlaid on this pattern is a faint,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

Section 2

Insights drawn from EDA

Flight Number vs. Launch Site

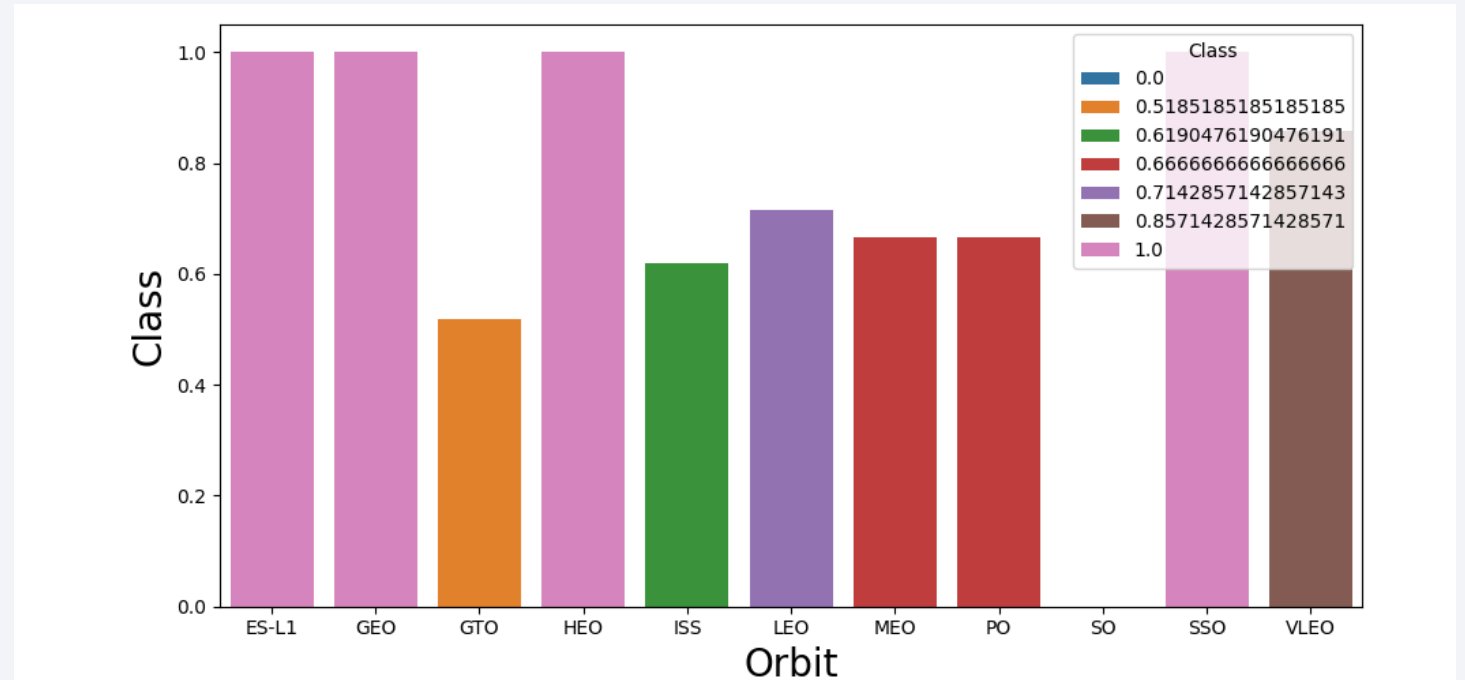
From the scatter plot, we see a relationship where the larger the flight count at a launch site, the greater the success rate.



Success Rate vs. Orbit Type

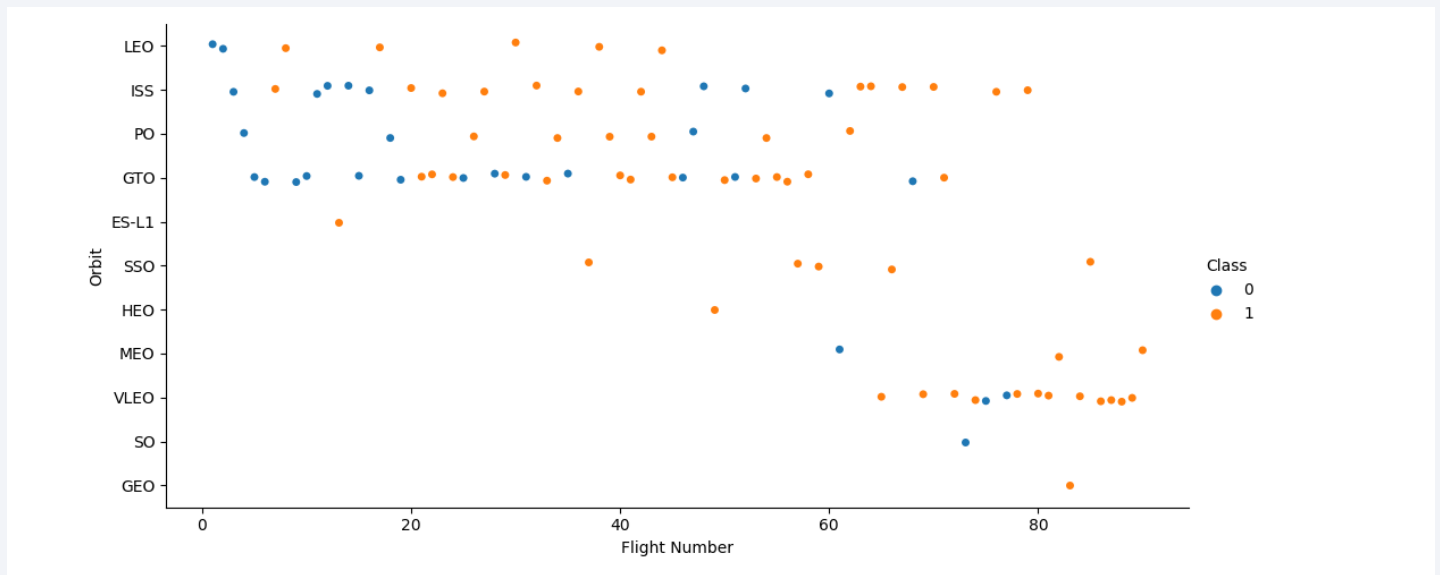
From the bar chart we see that the following orbits have a high success rate:

- ES-L1
- GEO
- HEO
- SSO



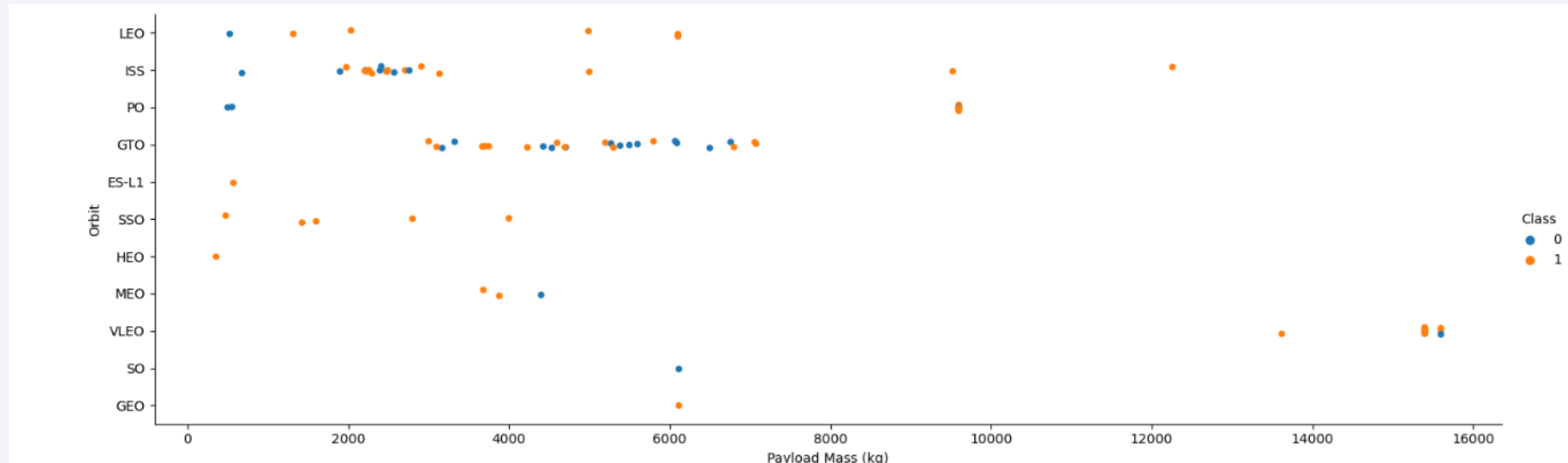
Flight Number vs. Orbit Type

- From the scatter plot, we see that in 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.



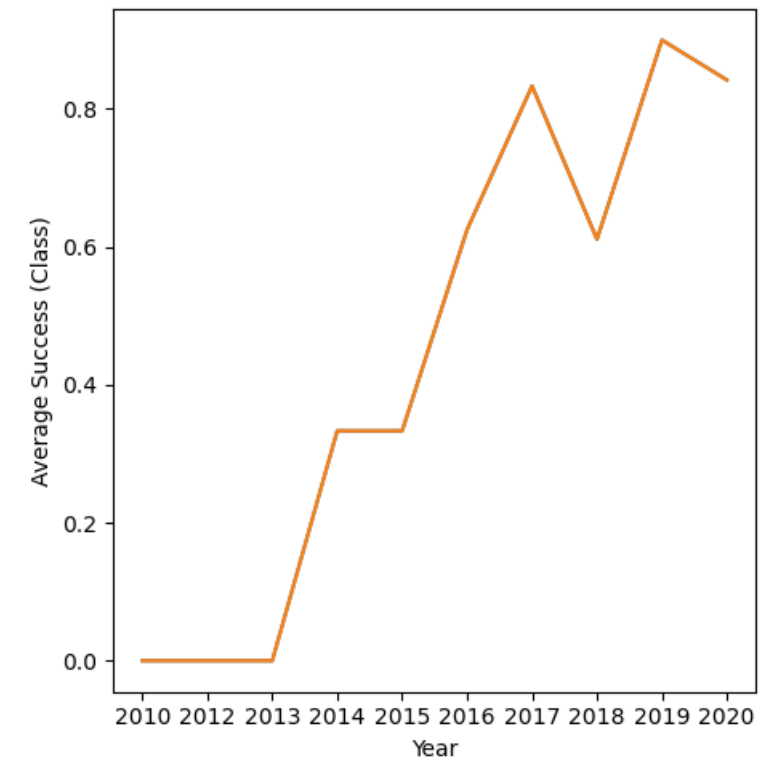
Payload vs. Orbit Type

- From the scatter plot, we see that 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.



Launch Success Yearly Trend

From the line plot we see that the success rate since 2013 kept increasing till 2020



All Launch Site Names

```
%sql select distinct launch_site from spacex
```

- Display the names of the unique launch sites in the space mission

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

Launch Site Names that begin with 'KSC'

```
%%sql select *
```

```
from spacex
```

```
where launch_site like 'KSC%' limit 5
```

- Display 5 records where launch sites begin with the string 'KSC'

DATE	TIME__UTC_	BOOSTER_VERSION	LAUNCH_SITE	PAYLOAD	PAYLOAD_MASS__KG_	ORBIT	CUSTOMER	MISSION_OUTCOME	LANDING__OUTCOME
2017-01-05	11:15:00	F9 FT B1032.1	KSC LC-39A	NROL-76	5300	LEO	NRO	Success	Success (ground pad)
2017-03-06	21:07:00	F9 FT B1035.1	KSC LC-39A	SpaceX CRS-11	2708	LEO (ISS)	NASA (CRS)	Success	Success (ground pad)
2017-05-07	23:38:00	F9 FT B1037	KSC LC-39A	Intelsat 35e	6761	GTO	Intelsat	Success	No attempt
2017-07-09	14:00:00	F9 B4 B1040.1	KSC LC-39A	Boeing X-37B OTV-5	4990	LEO	U.S. Air Force	Success	Success (ground pad)
2017-11-10	22:53:00	F9 FT B1031.2	KSC LC-39A	SES-11 / EchoStar 105	5200	GTO	SES EchoStar	Success	Success (drone ship)

Total Payload Mass

```
%%sql select sum(payload_mass__kg_) from spacex  
where customer='NASA (CRS)'
```

- Display the total payload mass carried by boosters launched by NASA (CRS)

In [67]:

```
%%sql select sum(payload_mass__kg_)  
from spacex  
where customer='NASA (CRS)'
```

```
* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb  
Done.
```

Out[67]:

1

22007

Average Payload Mass by F9 v1.1

```
%%sql select avg(payload_mass__kg_)
      from spacex
      where booster_version='F9 v1.1'
```

- Display average payload mass carried by booster version F9 v1.1

In [66]:

```
%%sql select avg(payload_mass__kg_)
      from spacex
      where booster_version='F9 v1.1'
```

```
* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

Out[66]:

```
1
3676
```

First Successful Ground Landing Date

```
%%sql select min(DATE)
      from spacex
      where landing__outcome='Success (drone ship)'
```

- List the date where the first successful landing outcome in drone ship was achieved.

In [65]:

```
%%sql select min(DATE)
      from spacex
      where landing__outcome='Success (drone ship)'
```

```
* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

Out[65]:

```
1
2016-06-05
```

Successful Drone Ship Landing with Payload between 4000 and 6000

```
%%sql select booster_version, payload_mass__kg_, landing__outcome
      from spacex
      where landing__outcome='Success (ground pad)' AND payload_mass__kg_
      between 4000 AND 6000
```

- List the names of the boosters which have success in ground pad and have payload mass greater than 4000 but less than 6000

```
In [64]: %%sql select booster_version, payload_mass__kg_, landing__outcome
        from spacex
        where landing__outcome='Success (ground pad)' AND payload_mass__kg_ between 4000 AND 6000

* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

```
Out[64]:
```

booster_version	payload_mass__kg_	landing__outcome
F9 FT B1032.1	5300	Success (ground pad)
F9 B4 B1040.1	4990	Success (ground pad)
F9 B4 B1043.1	5000	Success (ground pad)

Total Number of Successful and Failure Mission Outcomes

```
%%sql select
count(mission_outcome) as
Successes

from spacex

where mission_outcome
like 'Success%'
```

- List the total number of successful mission outcomes

```
%%sql select
count(mission_outcome) as
Failures

from spacex

where mission_outcome
like 'Failure%'
```

- List the total number of failure mission outcomes

```
In [11]: %%sql select count(mission_outcome) as Successes
        from spacex
        where mission_outcome like 'Success%'
```

```
* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

```
Out[11]: successes
        _____
        45
```

```
In [12]: %%sql select count(mission_outcome) as Failures
        from spacex
        where mission_outcome like 'Failure%'
```

```
* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

```
Out[12]: failures
        _____
        0
```

Boosters Carried Maximum Payload

```
%%sql select booster_version, payload_mass__kg_  
      from spacex where payload_mass__kg_ =  
      (select max(payload_mass__kg_) from spacex)
```

- List the names of the *booster_versions* which have carried the maximum payload mass using a subquery

```
In [62]: %%sql select booster_version, payload_mass__kg_  
        from spacex where payload_mass__kg_ =  
        (select max(payload_mass__kg_) from spacex)
```

```
* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb  
Done.
```

```
Out[62]: booster_version  payload_mass__kg_
```

F9 B5 B1048.4	15600
F9 B5 B1049.4	15600
F9 B5 B1049.5	15600
F9 B5 B1060.2	15600
F9 B5 B1058.3	15600

2017 Launch Records

```
%%sql select to_char(to_date(month(DATE), 'MM'), 'MONTH') as
month, landing__outcome, booster_version, launch_site
from spacex
where landing__outcome = 'Success (ground pad)' and
DATE like '2017%'
```

- List the records which will display the month names, successful *landing_outcomes* in ground pad, *booster versions*, *launch_site* for the months in year 2017

```
[91]: %%sql select to_char(to_date(month(DATE), 'MM'), 'MONTH') as month, landing__outcome, booster_version, launch_site
from spacex
where landing__outcome = 'Success (ground pad)' and DATE like '2017%'
```

```
* ibm_db_sa://mvr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

```
[91]: MONTH    landing__outcome  booster_version  launch_site
-----
JANUARY  Success (ground pad)  F9 FT B1032.1  KSC LC-39A
MARCH    Success (ground pad)  F9 FT B1035.1  KSC LC-39A
JULY     Success (ground pad)  F9 B4 B1040.1  KSC LC-39A
```

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

```
%%sql select DATE,
count(landing__outcome) as count
from spacex

where DATE between '2010-06-04'
and '2017-03-20' and
landing__outcome like '%Success%'

group by DATE

order by
count(landing__outcome) DESC
```

- Rank Landing Outcomes Between 2010-06-04 and 2017-03-20 in Descending Order

```
%%sql select
count(mission_outcome) as
Failures

from spacex

where mission_outcome
like 'Failure%'
```

- Display all the Successful Landing Outcomes

```
In [103... %%sql select DATE, count(landing__outcome) as count from spacex
where DATE between '2010-06-04' and '2017-03-20' and landing__outcome like '%Success%'
group by DATE
order by count(landing__outcome) DESC

* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

```
Out[103...
  DATE  COUNT
2016-06-05    1
2016-08-04    1
2017-01-05    1
2017-03-06    1
```

```
In [101... %%sql select DATE, landing__outcome
from spacex
where DATE between '2010-06-04' and '2017-03-20' and landing__outcome like '%Success%'

* ibm_db_sa://mwr47214:***@6667d8e9-9d4d-4ccb-ba32-21da3bb5aafc.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:30376/bludb
Done.
```

```
Out[101...
  DATE      landing__outcome
2016-08-04  Success (drone ship)
2016-06-05  Success (drone ship)
2017-01-05  Success (ground pad)
2017-03-06  Success (ground pad)
```

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.

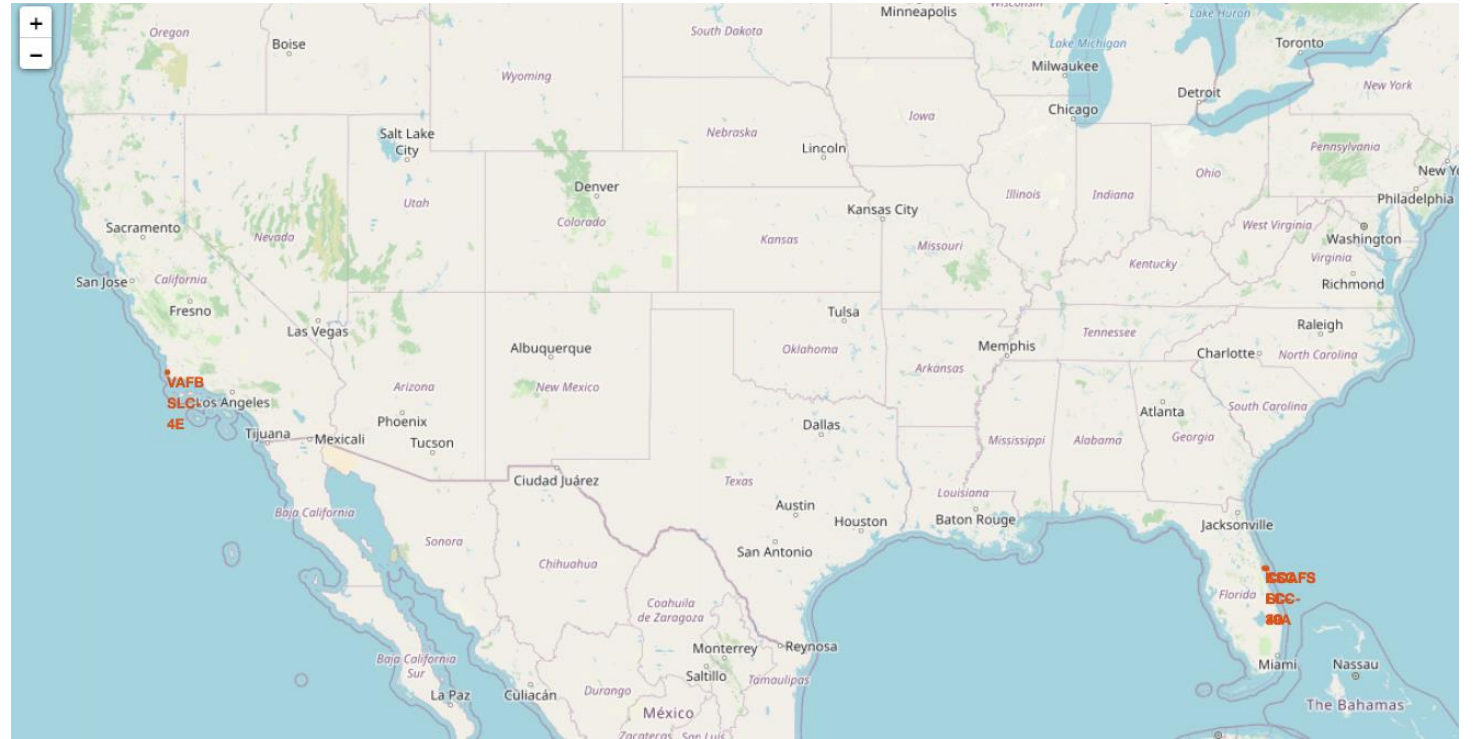
Section 3

Launch Sites Proximities Analysis

Mark all Launch Sites on a Map

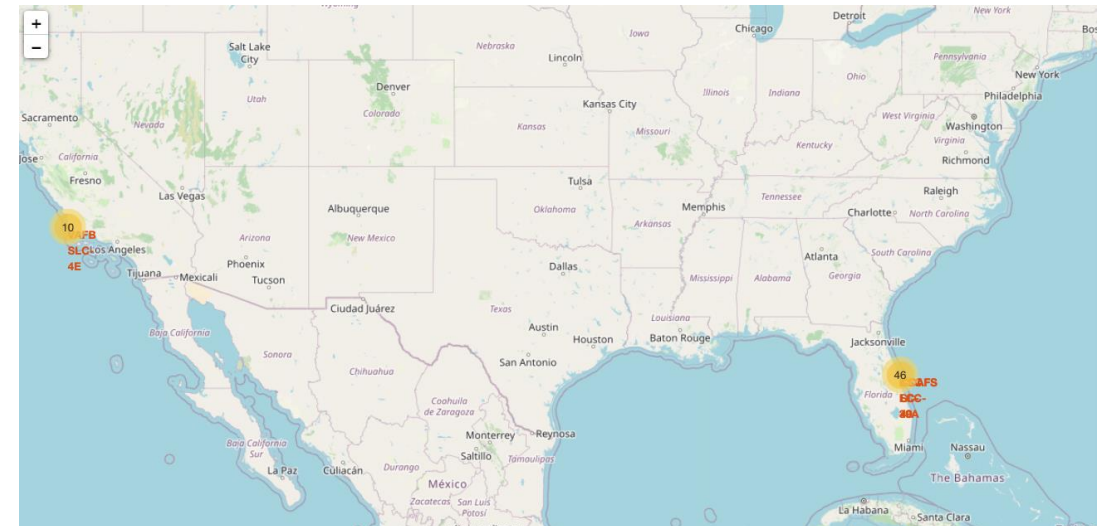
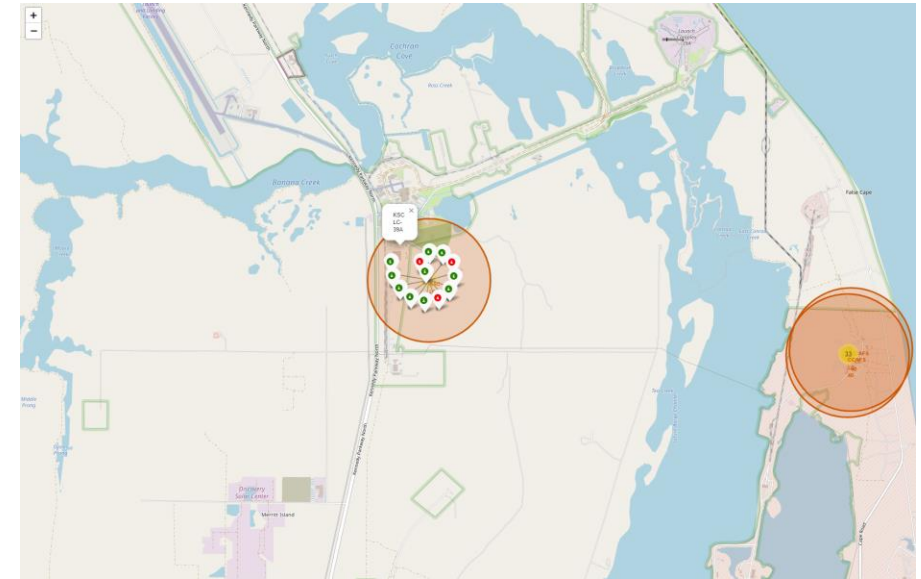
From the map, we're able to see that

- All launch sites are in proximity to the equator line
- All launch sites are in very close proximity to the coast



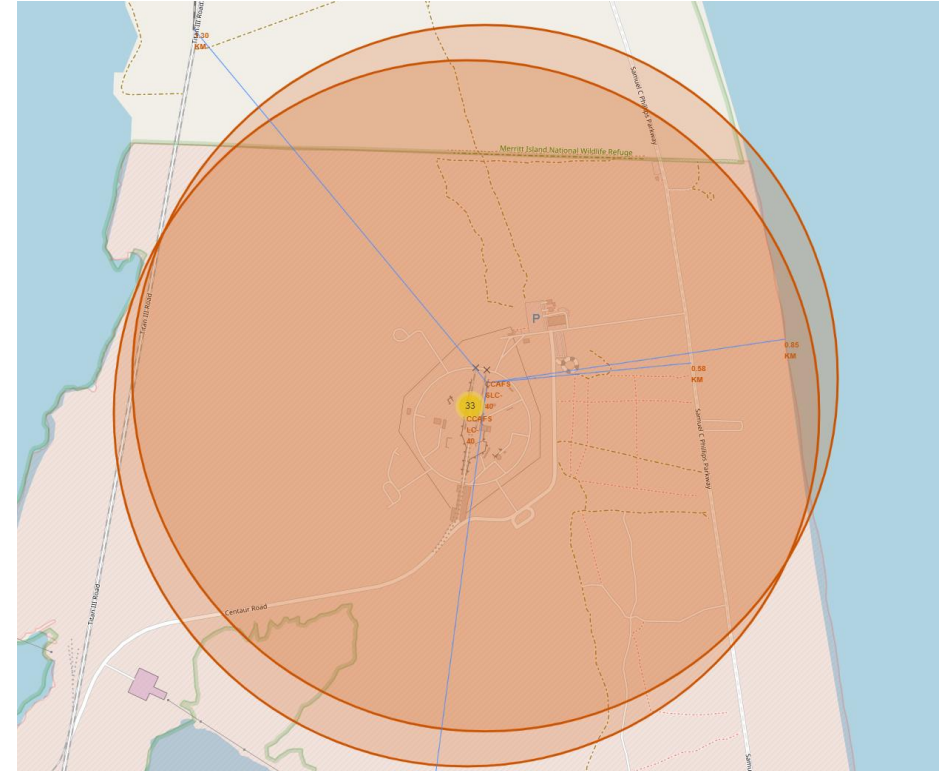
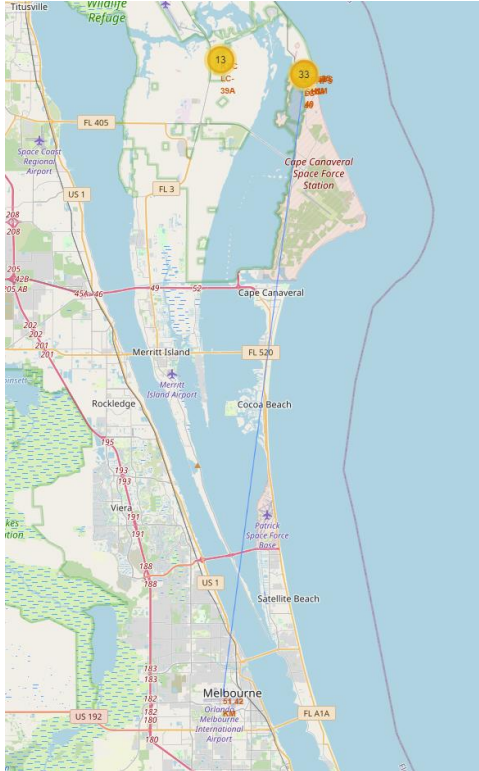
Mark the success/failed launches for each site on the map

- From the color-labeled markers in marker clusters, we can easily identify which launch sites have relatively high success rates.
- **Green Markers** show successful launches
- **Red Markers** show failures



Calculate the distances between a launch site to its proximities

- Launch Sites are in close proximity to railways
- Launch Sites are in close proximity to highways
- Launch Sites are in close proximity to the coastline
- Launch Sites keep far away from cities





Section 4

Build a Dashboard with Plotly Dash

SpaceX Falcon9 Interactive Launch Dashboard

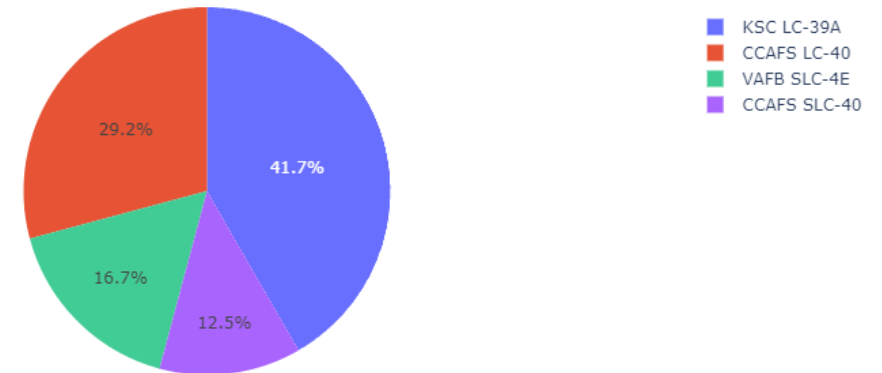
- From the pie chart, we can see all the successful launches from each site.
- KSC LC-39A has the most successful launches.
- CCAFS SLC-40 has the least.

SpaceX Launch Records Dashboard

All Sites

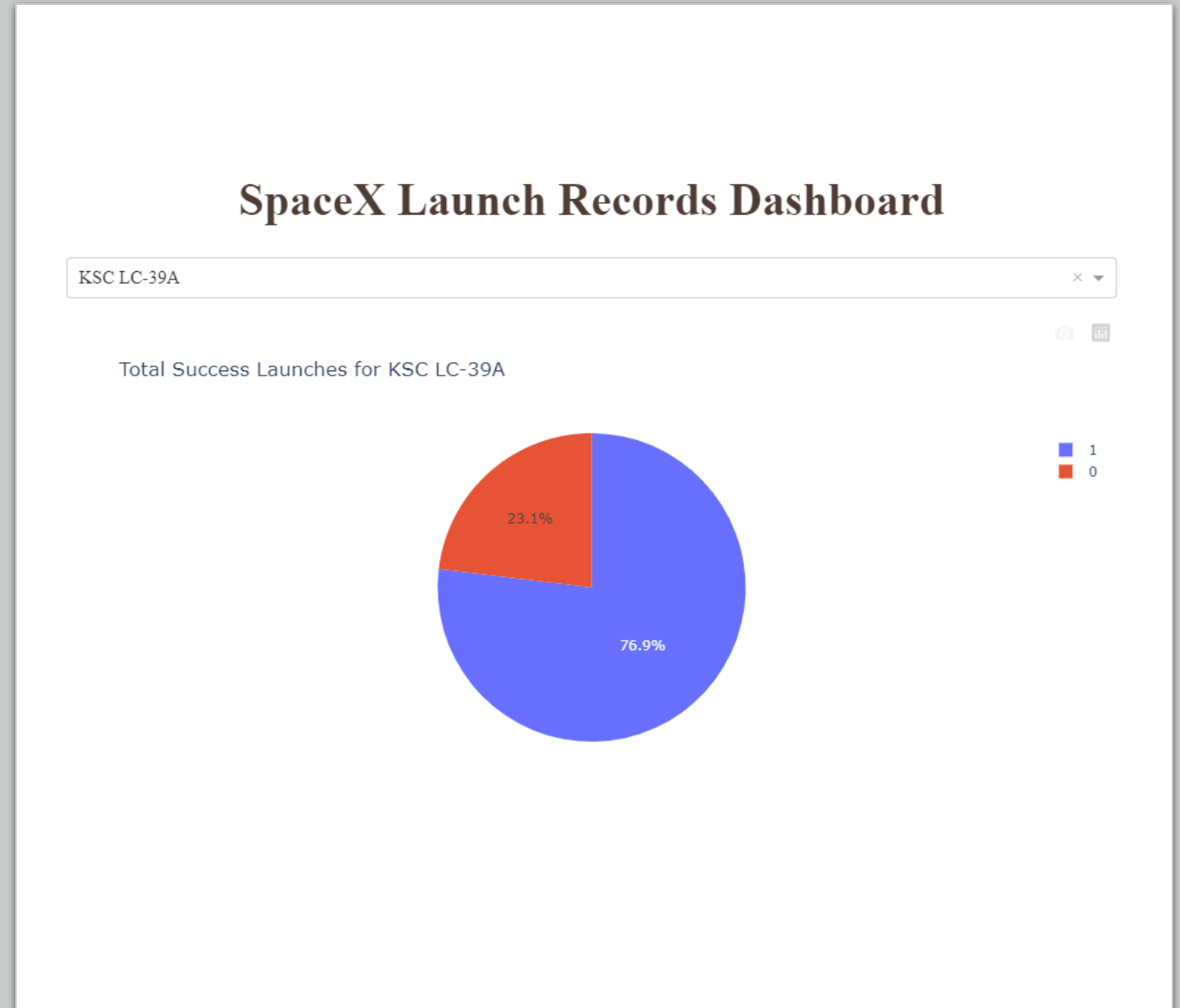
×

Total Successful Launches count for all sites



Launch Site with the Highest Launch Success Ratio

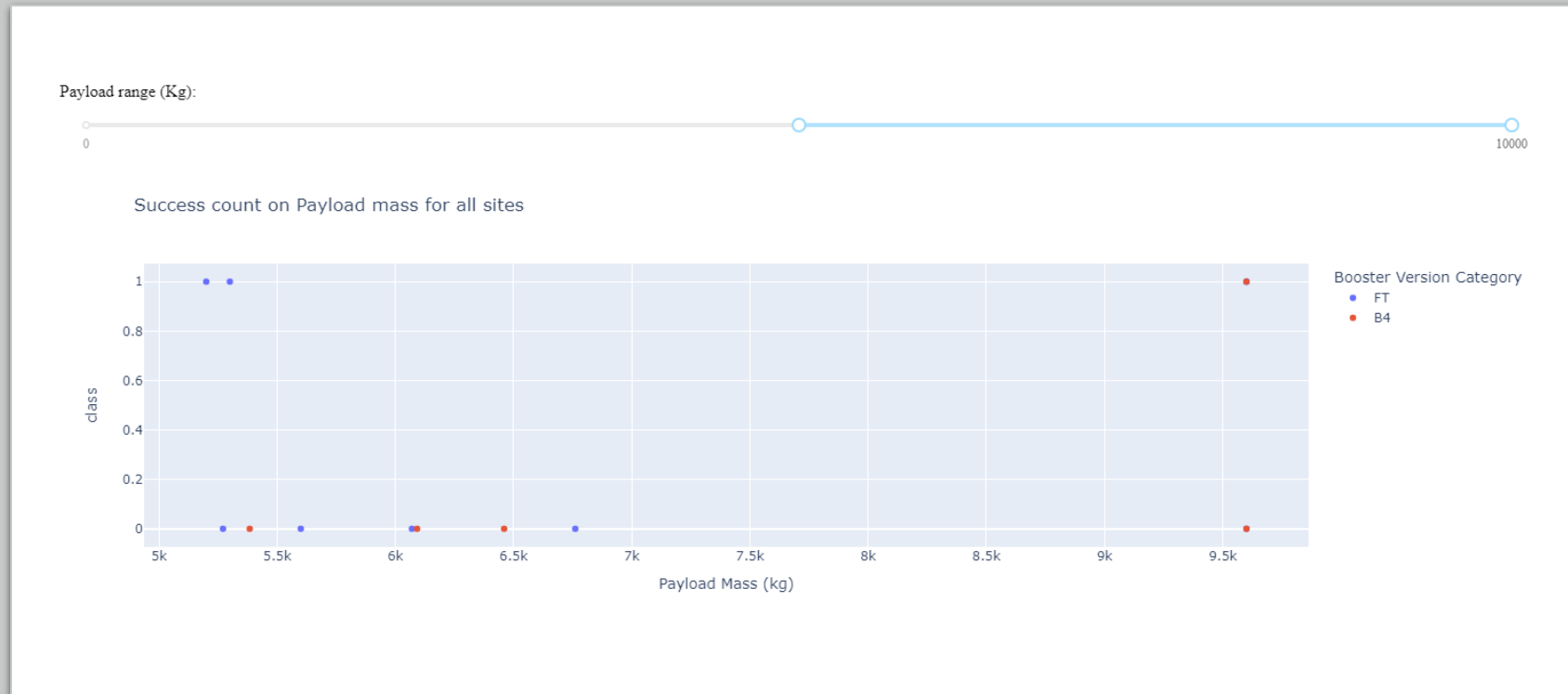
- KSC LC-39A achieved a 76.9% success rate while getting a 23.1% failure rate





Payload vs
Launch Outcome
Interactive Visual Analysis:
Low Payload Mass (kg)

- A scatter plot is created to display the Success Count vs Payload Mass for all sites
- Adjust range slider to display low payload mass (0-4k)
- Note the high success count
- We also see that the FT booster has the highest launch success rate



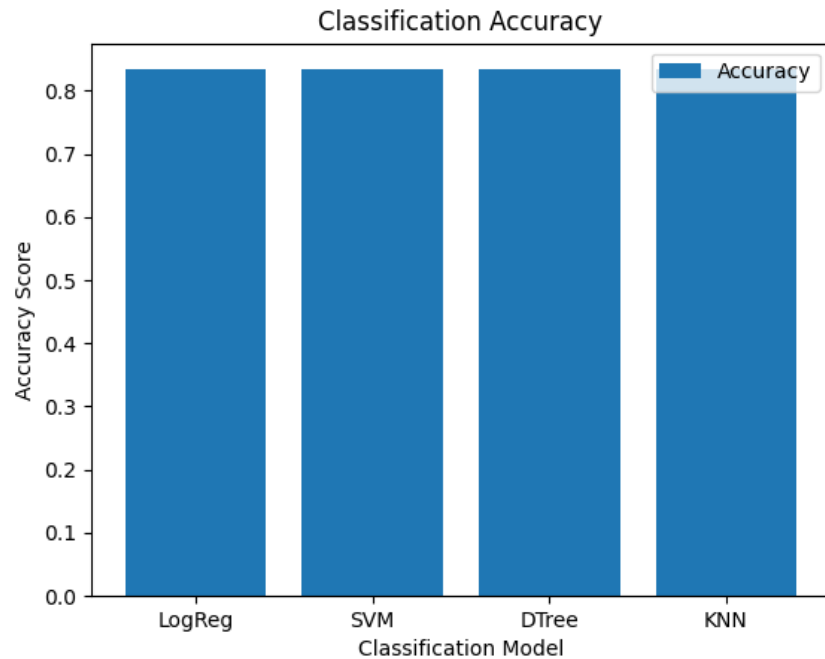
Payload vs Launch Outcome Interactive Visual Analysis: High Payload Mass (kg)

- Adjust range slider to display high payload mass (5k-10k)
- Note the low success count



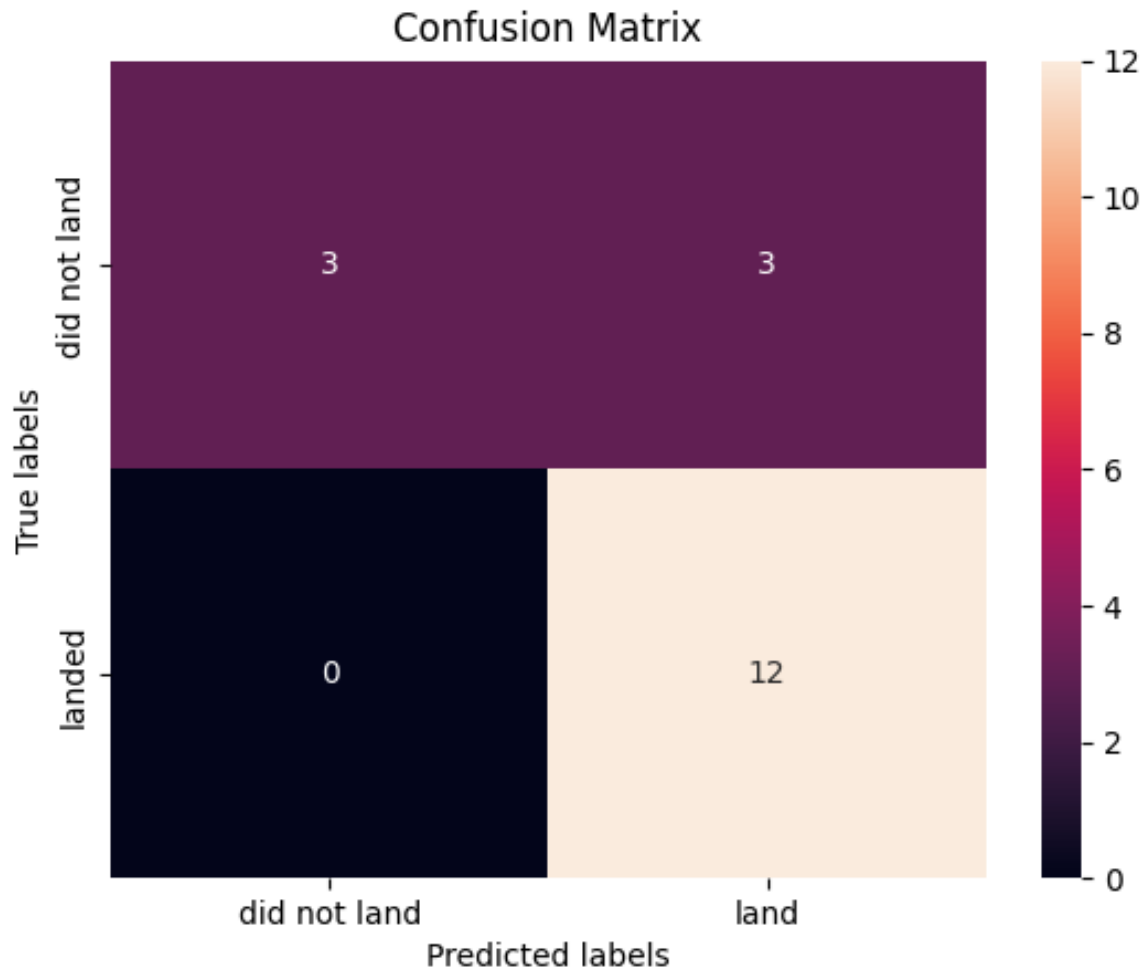
Section 5

Predictive Analysis (Classification)



Classification Accuracy

- Classification Models:
 - Logistic Regression
 - Support Vector Machine
 - Decision Tree Classifier
 - K-Nearest Neighbors
- All models had a test data accuracy of around .83%



Confusion Matrix

- All models shared the same confusion matrix
- The confusion matrix shows that the classifier can distinguish between the different classes.
- The major problem is the false positives (unsuccessful landing marked as successful landing)

Conclusions

- The larger the flight count at a launch site, the greater the success rate at said launch site.
- Launch success rate started to increase in 2013 until 2020.
- Orbits ES-L1, GEO, HEO, & SSO had the highest success rate.
- KSC LC-39A had the most successful launches of any sites with a 76.9% success rate and 23.1% failure rate
- Any of the aforementioned classification algorithms work well for this task.

Appendix

Cross-Validation data using GridSearchCV:

- LogReg
 - `tuned hyperparameters : (best parameters) {'C': 0.01, 'penalty': 'l2', 'solver': 'lbfgs'}`
 - **accuracy : 0.8464285714285713**
- SVM
 - `tuned hyperparameters : (best parameters) {'C': 1.0, 'gamma': 0.03162277660168379, 'kernel': 'sigmoid'}`
 - **accuracy : 0.8482142857142856**
- Decision Tree
 - `tuned hyperparameters : (best parameters) {'criterion': 'entropy', 'max_depth': 6, 'max_features': 'auto', 'min_samples_leaf': 4, 'min_samples_split': 10, 'splitter': 'random'}`
 - **accuracy : 0.8910714285714286**
- KNN
 - `tuned hyperparameters : (best parameters) {'algorithm': 'auto', 'n_neighbors': 10, 'p': 1}`
 - **accuracy : 0.8482142857142858**

Thank you!

