



SpaceX Falcon 9 Rocket Landing Prediction

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OUTLINE



[Project Github link](#)

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 - Dashboard
 - Predictive Analysis
- Discussion
- Conclusion

EXECUTIVE SUMMARY



- Logistic regression performs best with a prediction accuracy of 94.44%
- Rocket with lighter payload have higher success rate than rocket with heavier payload
- The landing success probability increases with time
- KSC LC 39A site has the highest launch success rate of 76%
- Orbit ES L1, SSO, HEO, and GEO has the best success rate of 100%

INTRODUCTION

- **Background**



- SpaceX launches their rockets at a cost of \$62 million which is almost one-third of other company's cost.
- SpaceX lowers the overall rocket launch cost by recovering and reusing the first stage boosters.
- Successful landing of the Falcon 9 first stage is crucial

- **Problem Statement**

- The objective of this capstone to predict if the Falcon 9 first stage will land successfully.

METHODOLOGY - SUMMARY

- Data Collection
 - Request to the SpaceX API
 - Web scrap Falcon 9 launch records HTML table from Wikipedia Data Wrangling
- Perform exploratory Data Analysis using SQL and visualization
- Interactive visualization using Folium and Plotly Dashboard
- Predictive Analysis using classification algorithms
 - Logistic regression, SVM, KNN, Decision Tree
 - Find the best classifier.

METHODOLOGY

- **Data Collection**

- We collect data from the SpaceX REST API by making a get request and receive the JSON data
- We convert the JSON data to a dataframe by using the `json_normalize` function.
- We also web scrap Wiki pages to obtain Falcon 9 launch data using Python BeautifulSoup package.
- We parse data from the HTML tables and convert it to pandas dataframe.
- We need to filter data which includes both Falcon 1 and Falcon 9 data to separate Falcon 9 booster info.

Data Collection from SpaceX API

Request and parse the SpaceX launch data using the GET request

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

200

Converting to pandas Dataframe

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

Applying custom function to clean the data

```
1 # Call getLaunchSite
2 getLaunchSite(data)
3 # Call getPayloadData
4 getPayloadData(data)
5 # Call getCoreData
6 getCoreData(data)
```

Creating dataframe

```
1 launch_dict = {'FlightNumber': list(data['flight_number']),
2 'Date': list(data['date']),
3 'BoosterVersion':BoosterVersion,
4 'PayloadMass':PayloadMass,
5 'Orbit':Orbit,
6 'LaunchSite':LaunchSite,
7 'Outcome':Outcome,
8 'Flights':Flights,
9 'GridFins':GridFins,
10 'Reused':Reused,
11 'Legs':Legs,
12 'LandingPad':LandingPad,
13 'Block':Block,
14 'ReusedCount':ReusedCount,
15 'Serial':Serial,
16 'Longitude': Longitude,
17 'Latitude': Latitude}
18 data = pd.DataFrame(launch_dict)
```

Data Collection from Wiki

Request the Falcon9 Launch Wiki page from its URL

```
1 html_data = requests.get(static_url).text
2 soup = BeautifulSoup(html_data, 'html.parser')
3 html_tables = soup.find_all('table')
```

Extracting Column name

```
1 column_names = []
2
3 for header in first_launch_table.find_all('th'):
4     col_name = extract_column_from_header(header)
5     if col_name is not None and len(col_name) > 0:
6         column_names.append(col_name)
7
8 launch_dict = dict.fromkeys(column_names)
```

Appending data to key of the dictionary

```
1 extracted_row = 0
2 for table_number, table in enumerate(soup.find_all('table', "wikital
3     for rows in table.find_all("tr"):
4         if rows.th:
5             if rows.th.string:
6                 flight_number=rows.th.string.strip()
7                 flag=flight_number.isdigit()
8             else:
9                 flag=False
10 row=rows.find_all('td')
11 if flag:
12     extracted_row += 1
13     launch_dict['Flight No.'].append(int(flight_number))
14     datatimelist=date_time(row[0])
15     date = datatimelist[0].strip(',')
16     launch_dict['Date'].append(date)
17     time = datatimelist[1]
18     launch_dict['Time'].append(time)
19     bv=booster_version(row[1])
20     if not(bv):
21         bv=row[1].a.string
```


METHODOLOGY

- **Exploratory Data Analysis**
 - Calculate the number of launches on each site
 - Calculate the number and occurrence of each orbit
 - Calculate the number and occurrence of mission outcome per orbit type
 - Create a landing outcome label from Outcome column

METHODOLOGY

- **Exploratory Data Analysis with SQL**

- Display the names of the unique launch sites in the space mission
- Display 5 records where launch sites begin with the string 'CCA'
- Display the total payload mass carried by boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1
- List the date when the first successful landing outcome in ground pad was achieved.

METHODOLOGY

- **Exploratory Data Analysis with SQL**

- List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- List the total number of successful and failure mission outcomes
- List the names of the booster_versions which have carried the maximum payload mass.
- List the records which will display the month names, failure landing_outcomes in drone ship ,booster versions, launch_site for the months in year 2015
- Rank the count of successful landing_outcomes between the date 04-06-2010 and 20-03-2017 in descending order

METHODOLOGY

- **Interactive Map with Folium**
 - Visualize the launch sites in an interactive map using Folium
 - Put a circle marker and label around the launch sites using latitude and longitude coordinates data of the sites.

METHODOLOGY

- **Interactive Dashboard using Plotly Dash**

- Pie chart showing the total launches of all sites and success rate of individual site separately.
- Scatter plot showing the payload masses for different booster versions.

METHODOLOGY - PREDICTIVE ANALYSIS

- **Building model**

- Load the data using Pandas
- Split the data into training and test set
- Choose the machine learning algorithms
- Machine learning model applied: Logistic regression, KNN, SVM, and Decision tree.

METHODOLOGY - PREDICTIVE ANALYSIS

- We tuned the hyperparameters using GridSearchCV to obtain the most accurate model.
- With best parameters, we evaluate their score , plot confusion matrix.
- Logistic regression model has the best score accuracy of 94.44%.

```
parameters ={'C':[0.01,0.1,1],  
             'penalty':['l2'],  
             'solver':['lbfgs']}
```

```
lr=LogisticRegression()  
logreg_cv = GridSearchCV(lr,  
                        param_grid=parameters, cv= 10)  
logreg_cv.fit(X_train,Y_train)
```

```
GridSearchCV(cv=10, estimator=LogisticRegression(),  
            param_grid={'C': [0.01, 0.1, 1], 'penalty': ['l2'],  
                        'solver': ['lbfgs']})
```

```
print("tuned hpyerparameters :(best parameters) ",  
      logreg_cv.best_params_)  
print("accuracy :",logreg_cv.best_score_)
```

```
1 | logreg_cv.score(X_test, Y_test)
```

```
0.9444444444444444
```

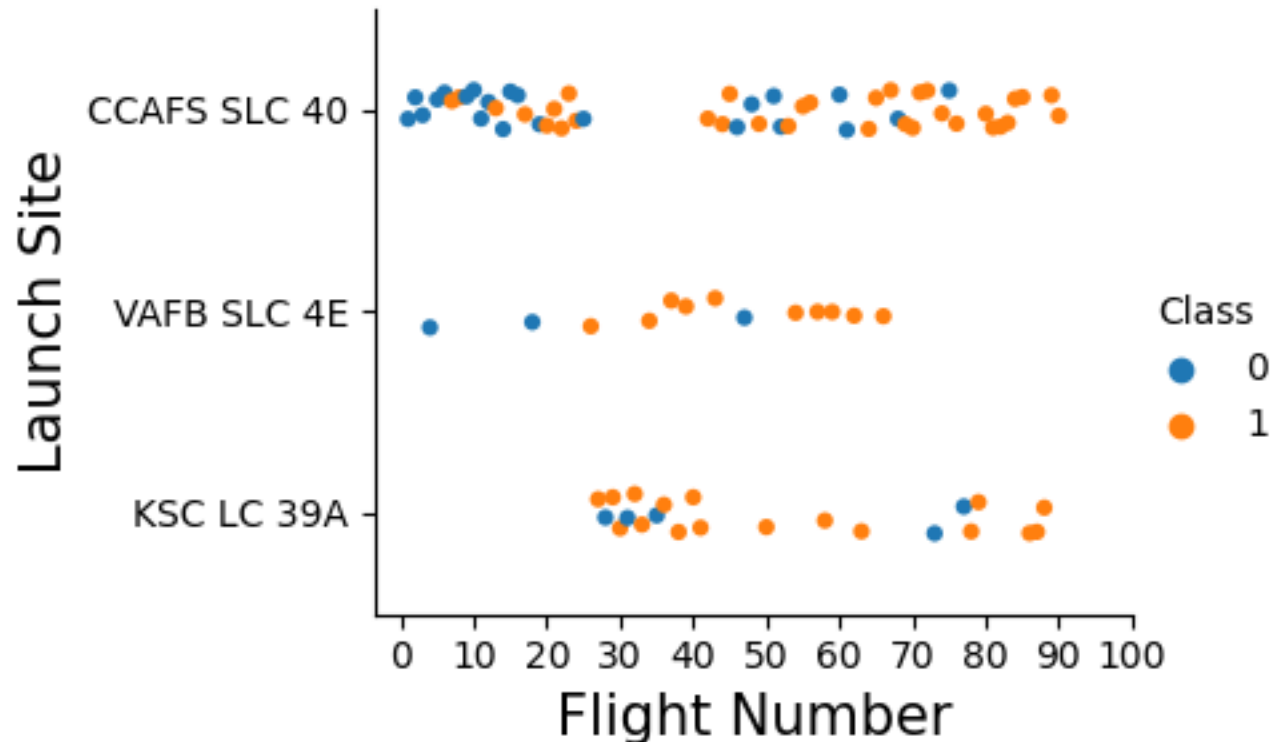
RESULTS

EXPLORATORY DATA ANALYSIS

Data visualization using seaborn

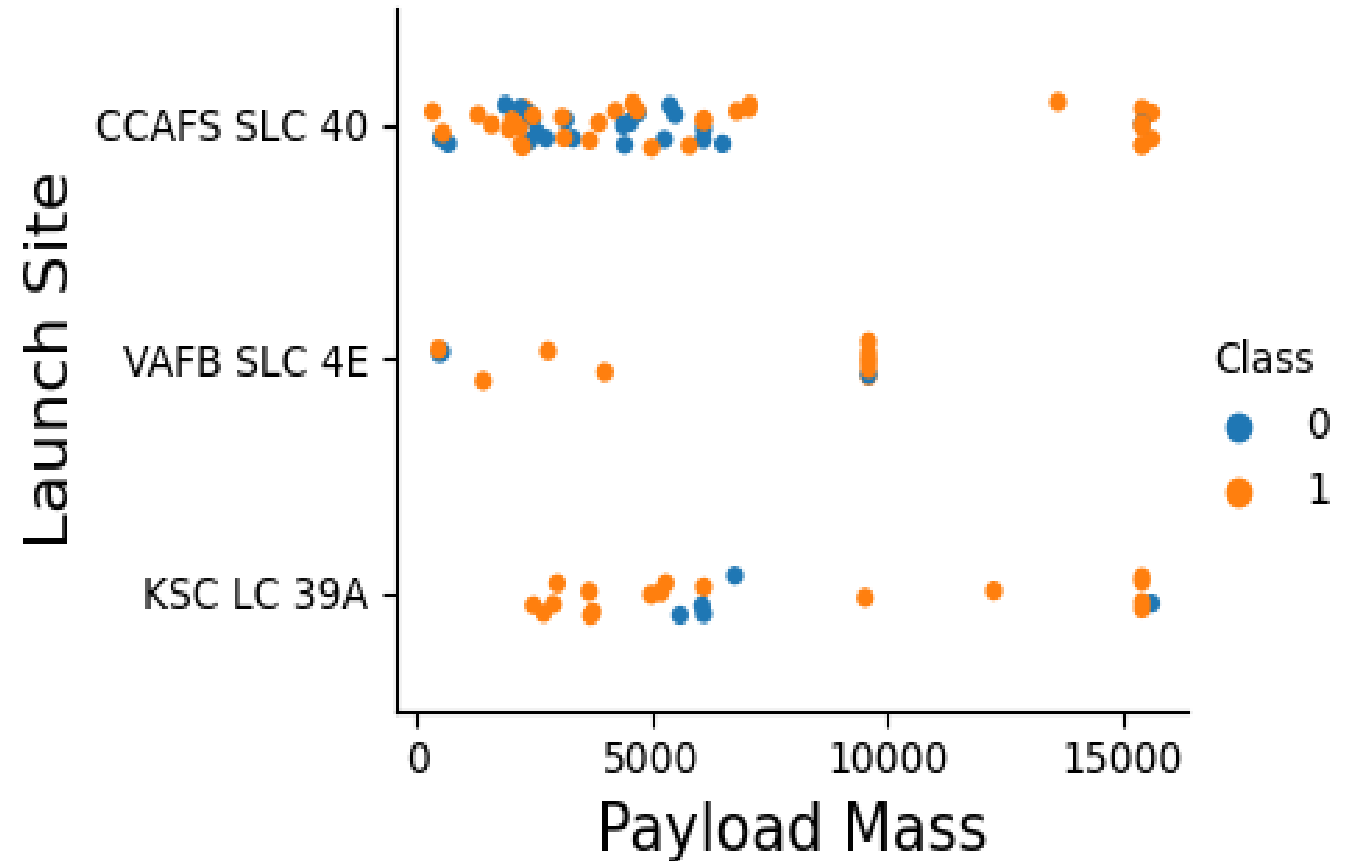
Flight Number vs Launch Site

- Most rockets are launched from CCAFS SLC 40 site.
- Fewer rockets are launched from VAFB SLC 4E SITE
- Success rate is greatest in KSC LC 39A site



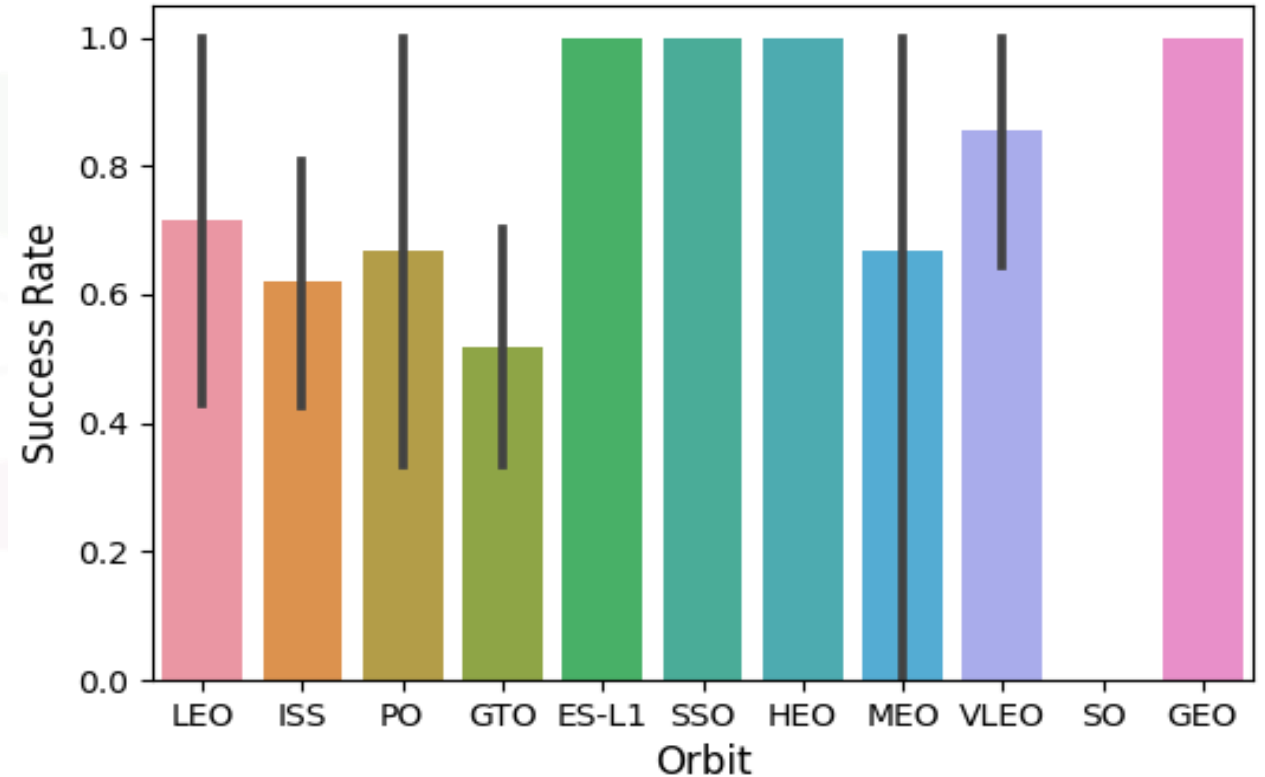
Payload Mass vs Launch Site

- No rockets greater than 10000 kg (heavy payload) launched in VAFB-SLC launch site.
- In CCAFS SLC 40 site, majority of the rocket launched are less than about 8000 kg
- In CCAFS SLC 40 site, in case of payload greater than 10000 kg, the success rate is 100%.



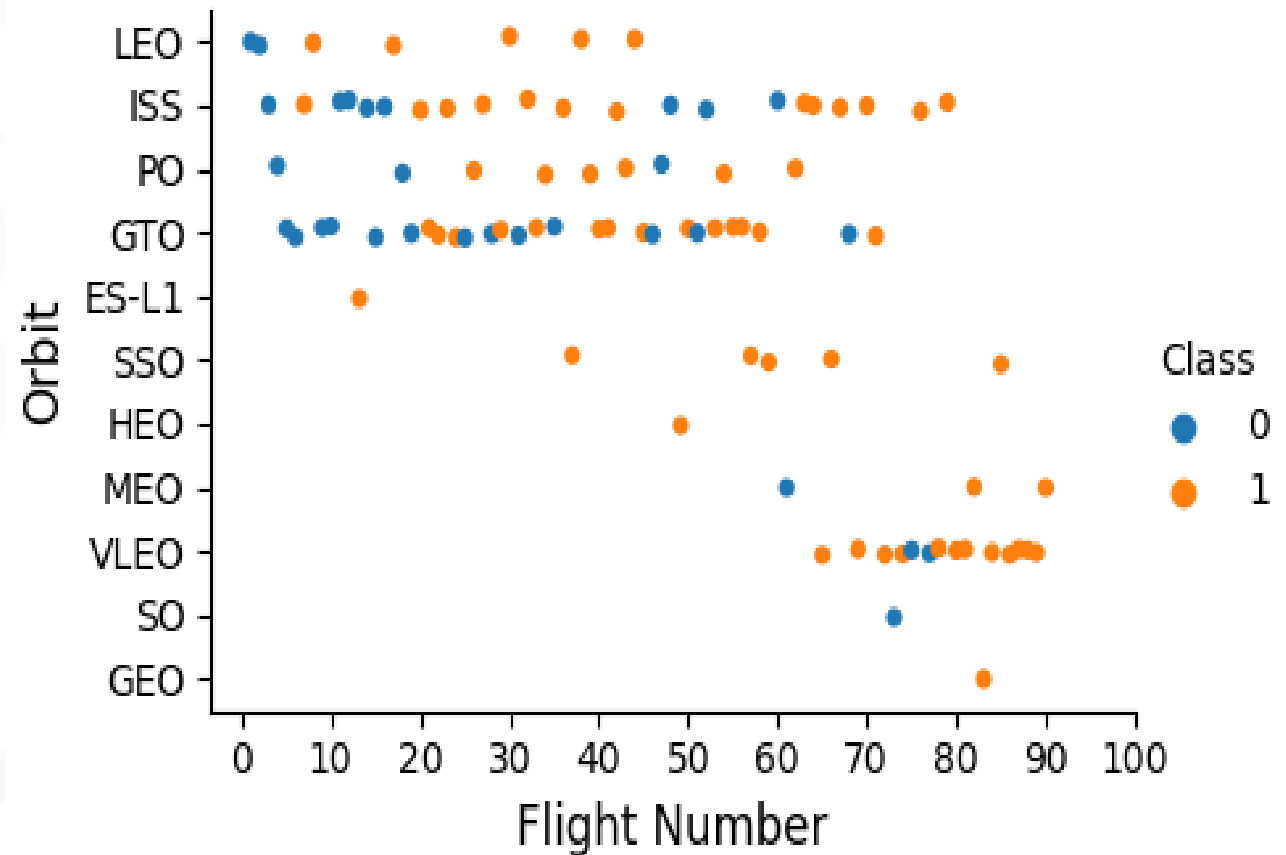
Success Rate vs Orbit

- The ES-L1, SSO, HEO, and GEO orbits have 100% success rate



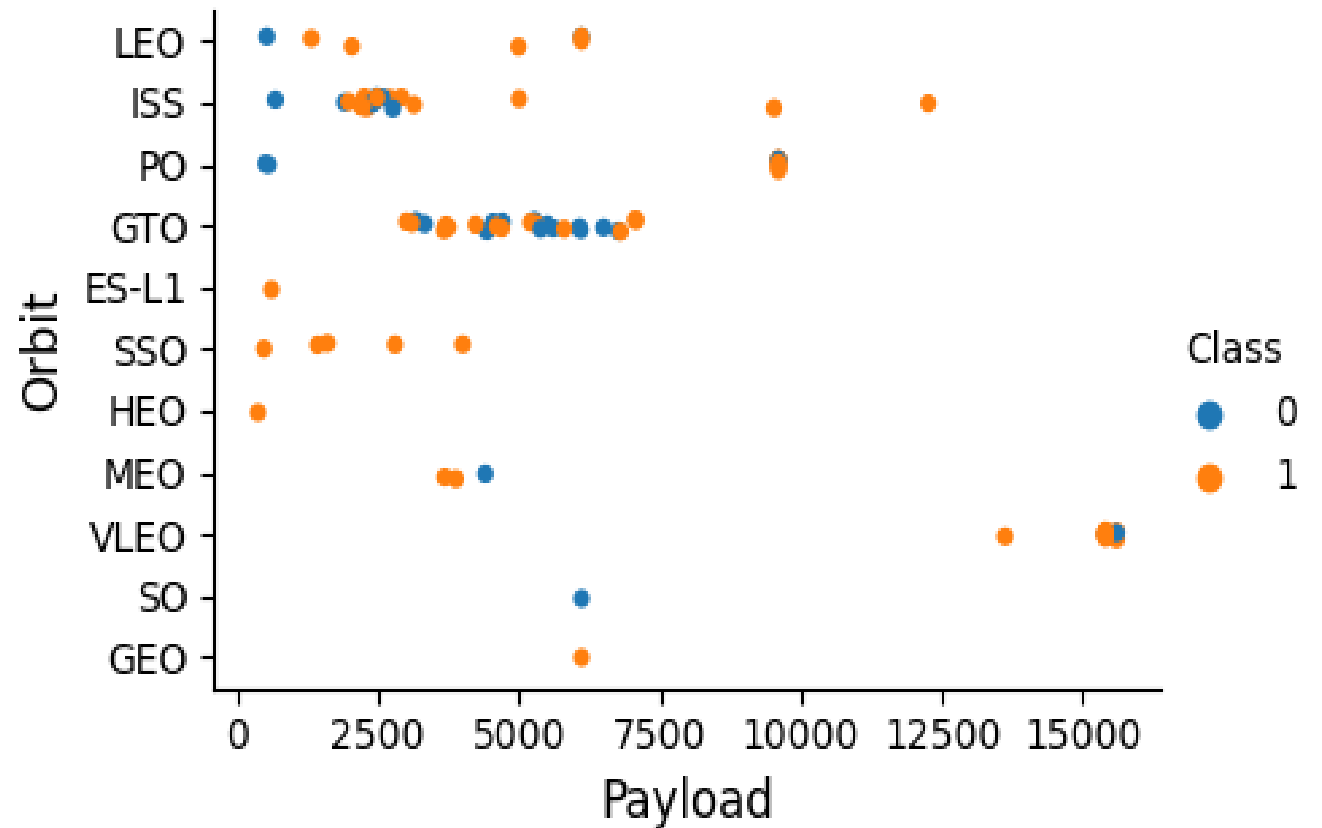
Flight Number vs Orbit

- In the LEO orbit, the initial flights are not successful but later success rate improves significantly in later flights.
- In GTO orbit, there seems to be no relationship between flight number.
- Majority of the launches in recent years are in VLEO orbit



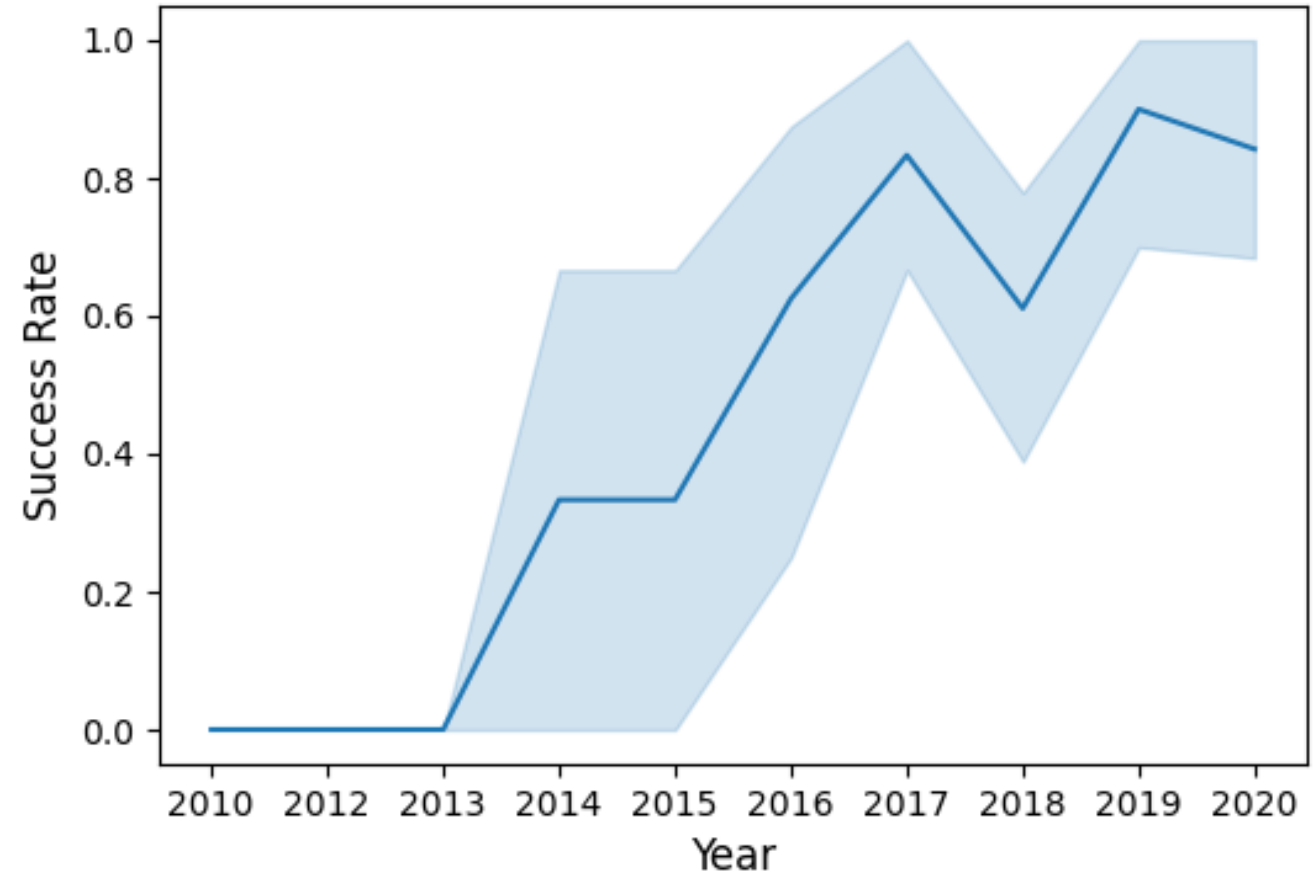
Payload Mass vs Orbit

- For Polar, LEO and ISS orbit, heavy payloads have higher success rate.
- In case of GTO orbit success rate are about 50%



Success Rate with time

- Success probability increases significantly with time since 2013 due to maybe technological advancement.



RESULTS

EXPLORATORY DATA ANALYSIS

Data exploration with SQL

Data Exploration using SQL

- Display the names of the unique launch sites in the space mission
- Display 5 records where launch sites begin with the string 'CCA'

	Launch_Site
0	CCAFS LC-40
1	VAFB SLC-4E
2	KSC LC-39A
3	CCAFS SLC-40

	Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG_	Orbit	Customer	Mission_Outcome	Landing_Outcome
0	04-06-2010	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
1	08-12-2010	15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of...	0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
2	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
3	08-10-2012	00:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
4	01-03-2013	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

Data Exploration using SQL

- Display the total payload mass carried by boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1

```
SUM("PAYLOAD_MASS_KG")
```

0	45596
---	-------

```
AVG("PAYLOAD_MASS_KG")
```

0	2928.4
---	--------

Data Exploration using SQL

- List the date when the first succesful landing outcome in ground pad was acheived.
- List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

MIN(Date)

0	01-05-2017
---	------------

Booster_Version

0	F9 FT B1022
---	-------------

1	F9 FT B1026
---	-------------

2	F9 FT B1029.1
---	---------------

3	F9 FT B1021.2
---	---------------

4	F9 FT B1036.1
---	---------------

Data Exploration using SQL

- List the total number of successful and failure mission outcomes
- List the names of the booster_versions which have carried the maximum payload mass.

Count("Mission_Outcome")

0

100

Booster_Version

0 F9 B5 B1048.4

1 F9 B5 B1049.4

2 F9 B5 B1051.3

3 F9 B5 B1056.4

4 F9 B5 B1048.5

Data Exploration using SQL

- List the records which will display the month names, failure landing_outcomes in drone ship ,booster versions, launch_site for the months in year 2015
- Rank the count of successful landing_outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

	Month_names	Landing_Outcome	Booster_Version	Launch_Site
0	01	Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
1	04	Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40

COUNT("Landing_Outcome")

0	34
---	----

RESULTS

INTERACTIVE DATA ANALYSIS

- Launch Site map using Folium
- Dashboard with Plotly Dash

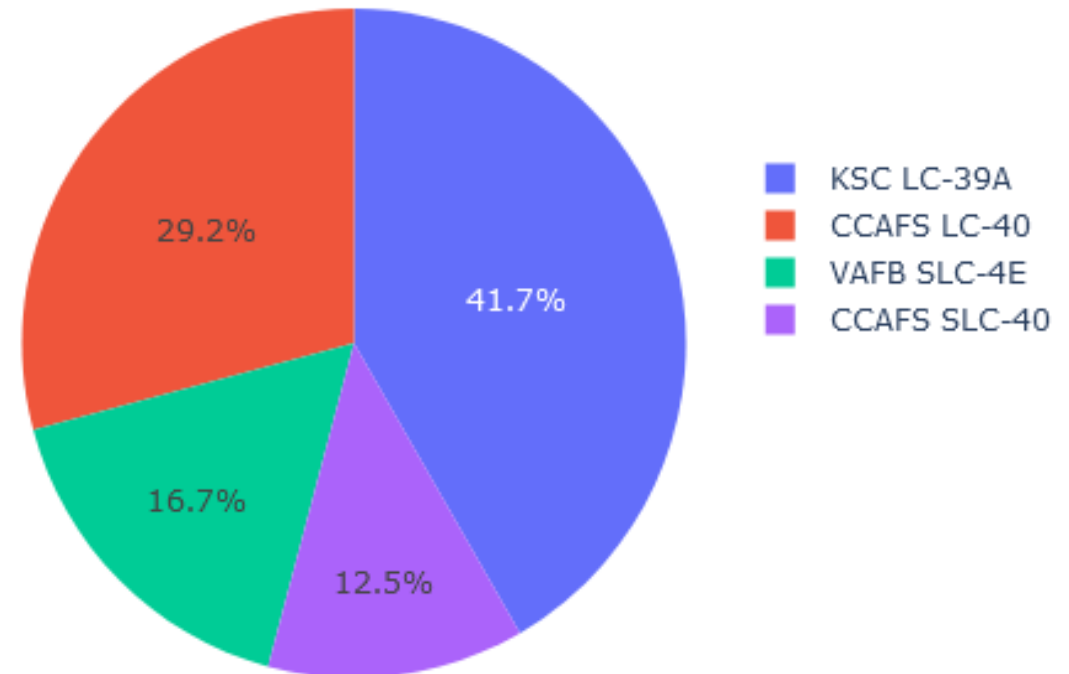
Map with marked launch sites



Dashboard: Success Rate for all launch sites

Success Count for all launch sites

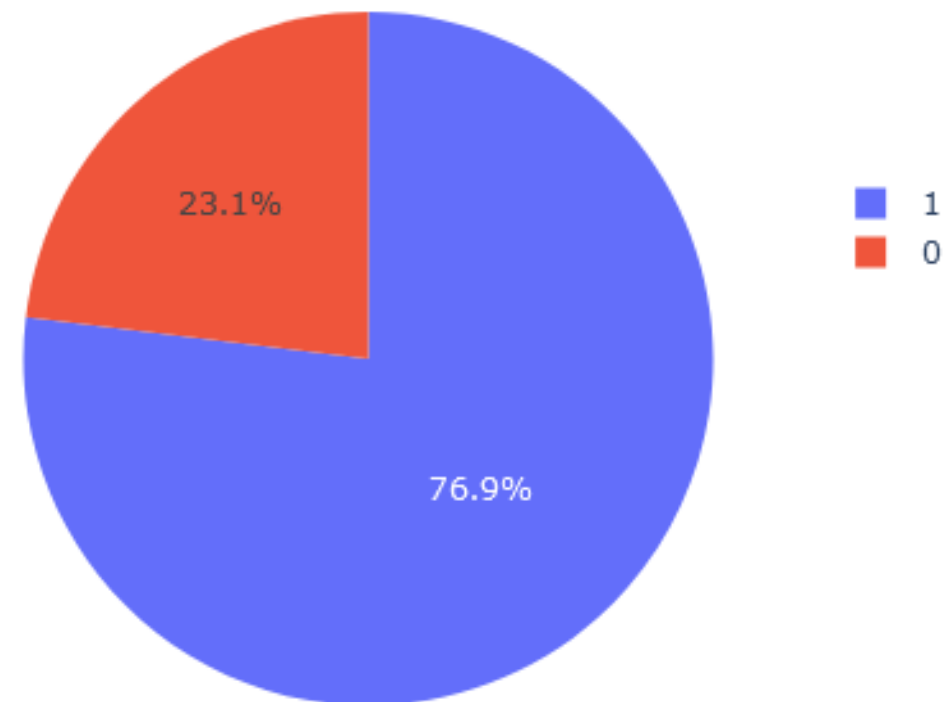
Payloads between 2000 and 5000 kg have the highest success rate.



Dashboard: Which site has the highest launch success rate?

KSC LC 39A has the highest success rate of 76.9%

Total Success Launches for site KSC LC-39A



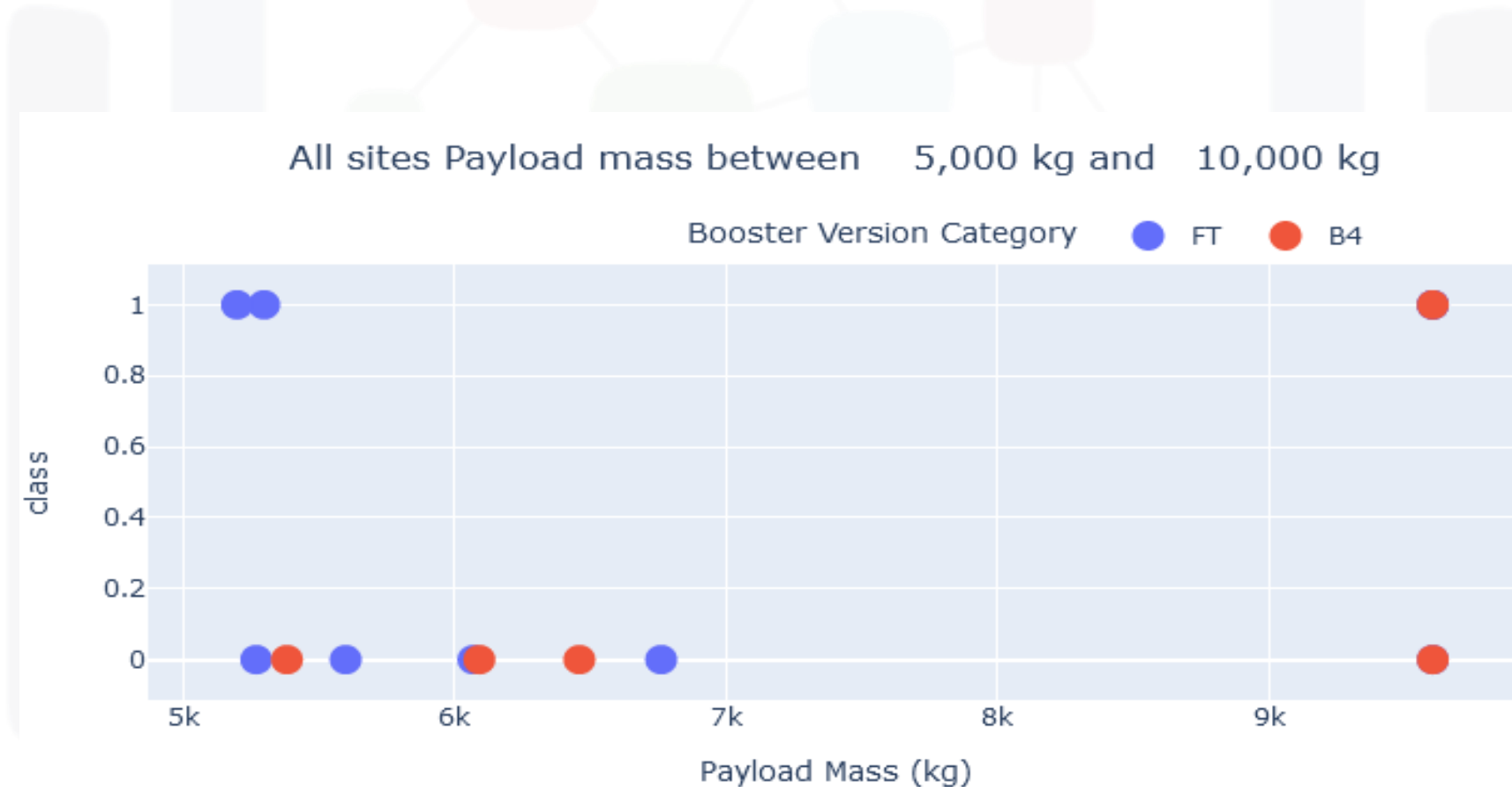
Dashboard: Which payload range(s) has the highest launch success rate?

Payloads between 2000 and 5000 kg have the highest success rate.



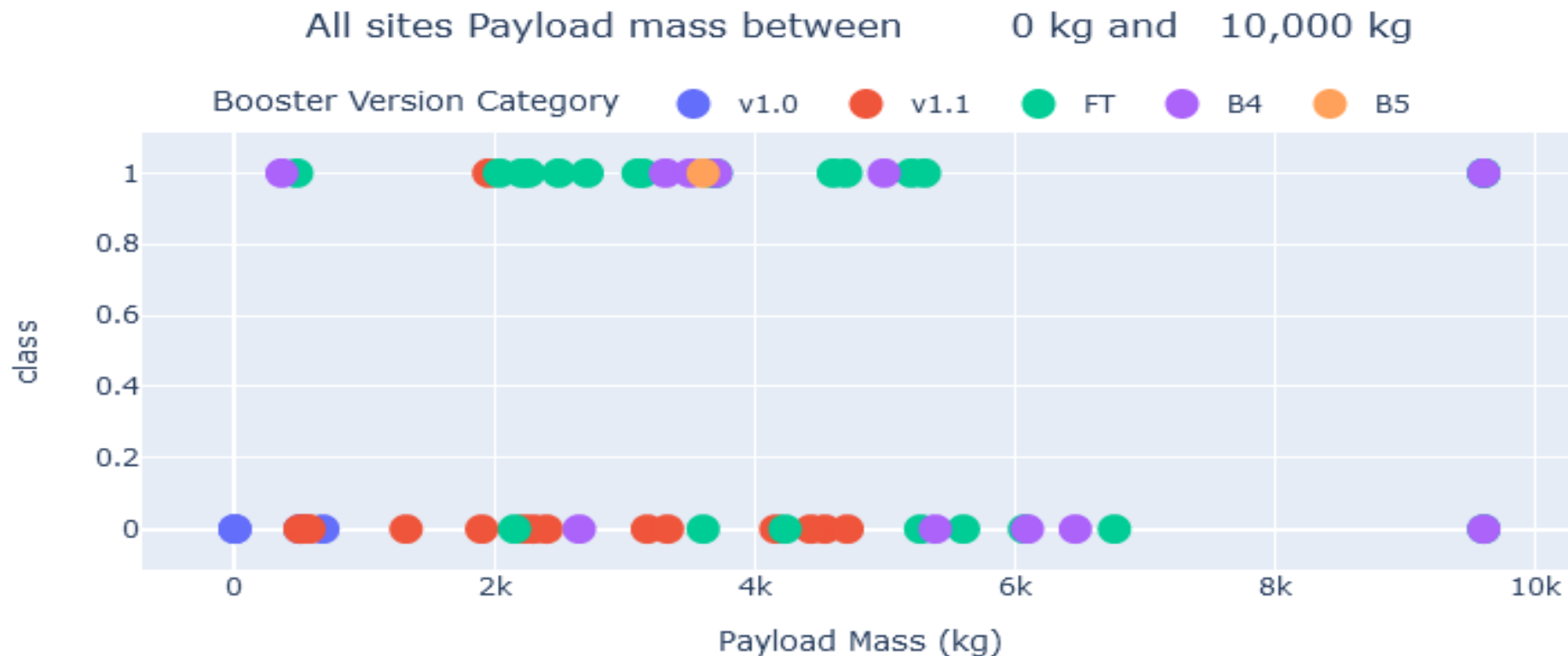
Dashboard: Which payload range(s) has the lowest launch success rate?

Payloads between 5000 and 10000 kg have the lowest success rate.



Dashboard: Which F9 Booster version (v1.0, v1.1, FT, B4, B5, etc.) has the highest launch success rate?

F9 Booster version FT has the highest success rate.



RESULTS

PREDICTIVE ANALYSIS

- Different machine learning algorithm performance to find the best performed method

Results: Find the method performs best

```
1 print("Logistic Regression: ", logreg_cv.score(X_test, Y_test))
2 print("SVM: ", svm_cv.score(X_test, Y_test))
3 print("Decision Tree: ", tree_cv.score(X_test, Y_test))
4 print("KNN: ", knn_cv.score(X_test, Y_test))
```

Logistic Regression: 0.9444444444444444

SVM: 0.8333333333333334

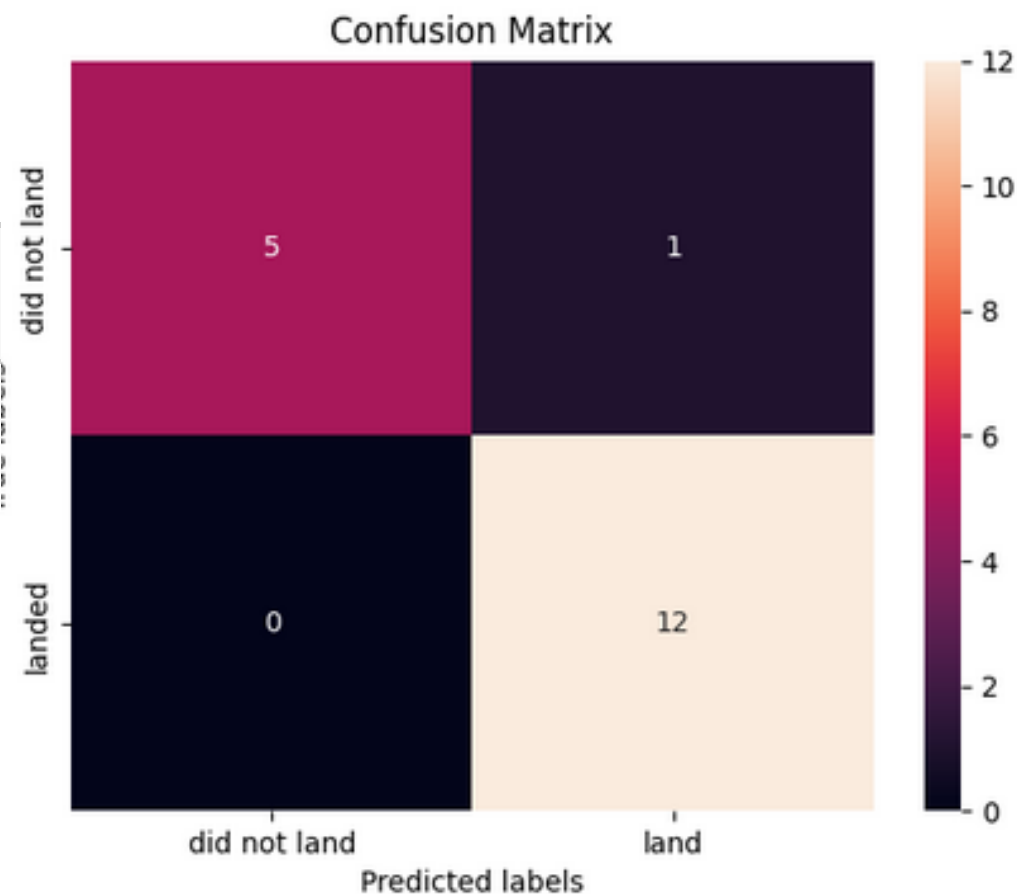
Decision Tree: 0.8888888888888888

KNN: 0.8333333333333334

Logistic Regression

```
1 logreg_cv.score(X_test, Y_test)
```

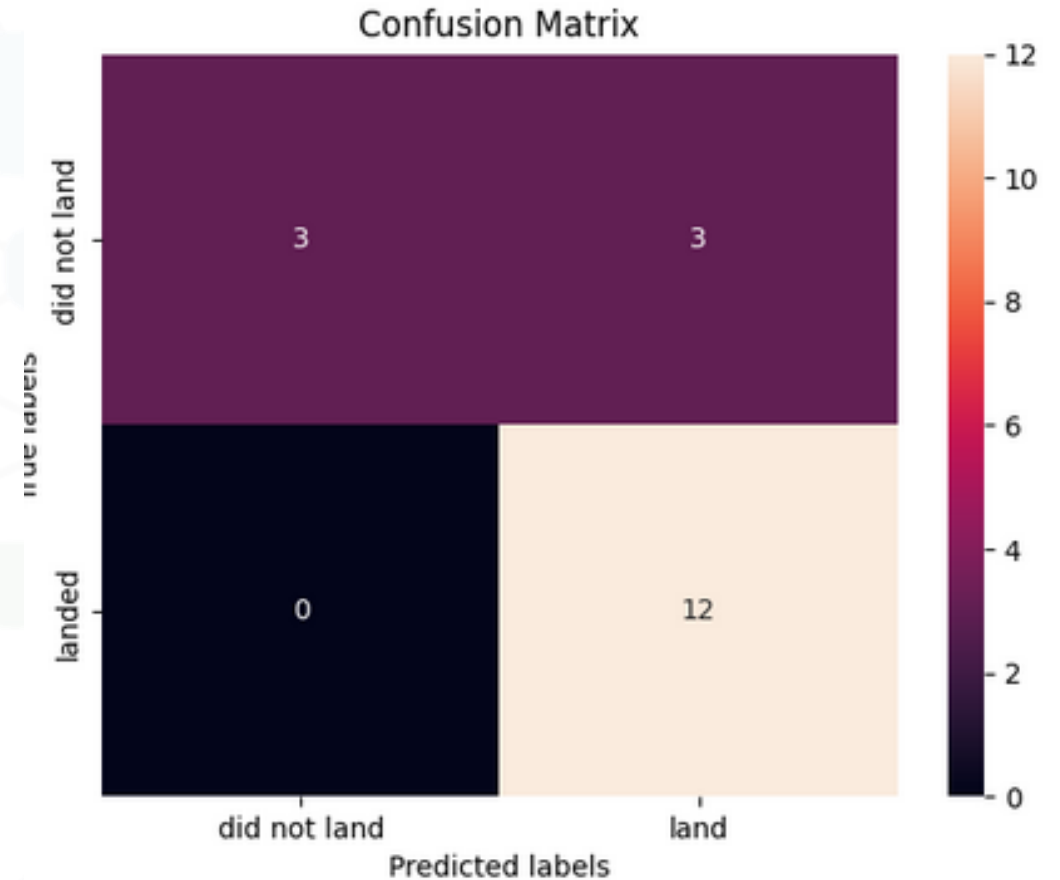
```
0.9444444444444444
```



Support Vector Machine

```
1 svm_cv.score(X_test, Y_test)
```

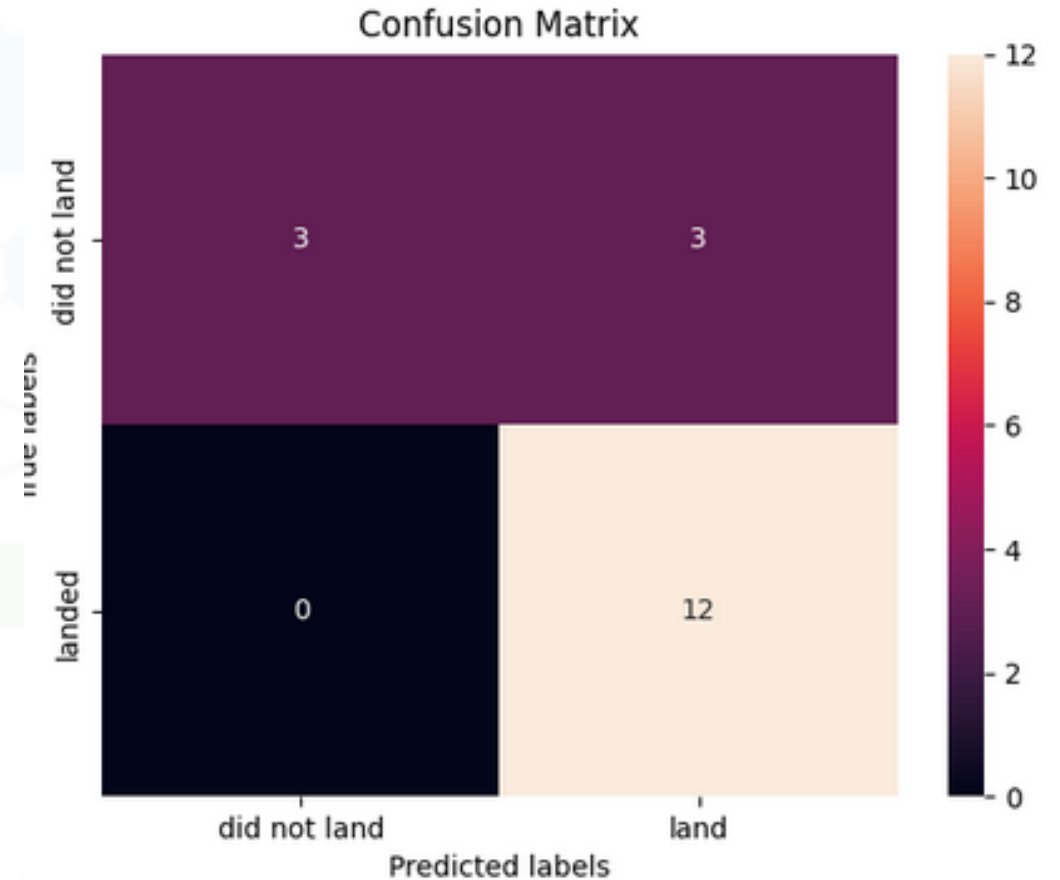
```
0.8333333333333333
```



Decision Tree Classifier

```
1 tree_cv.score(X_test, Y_test)
```

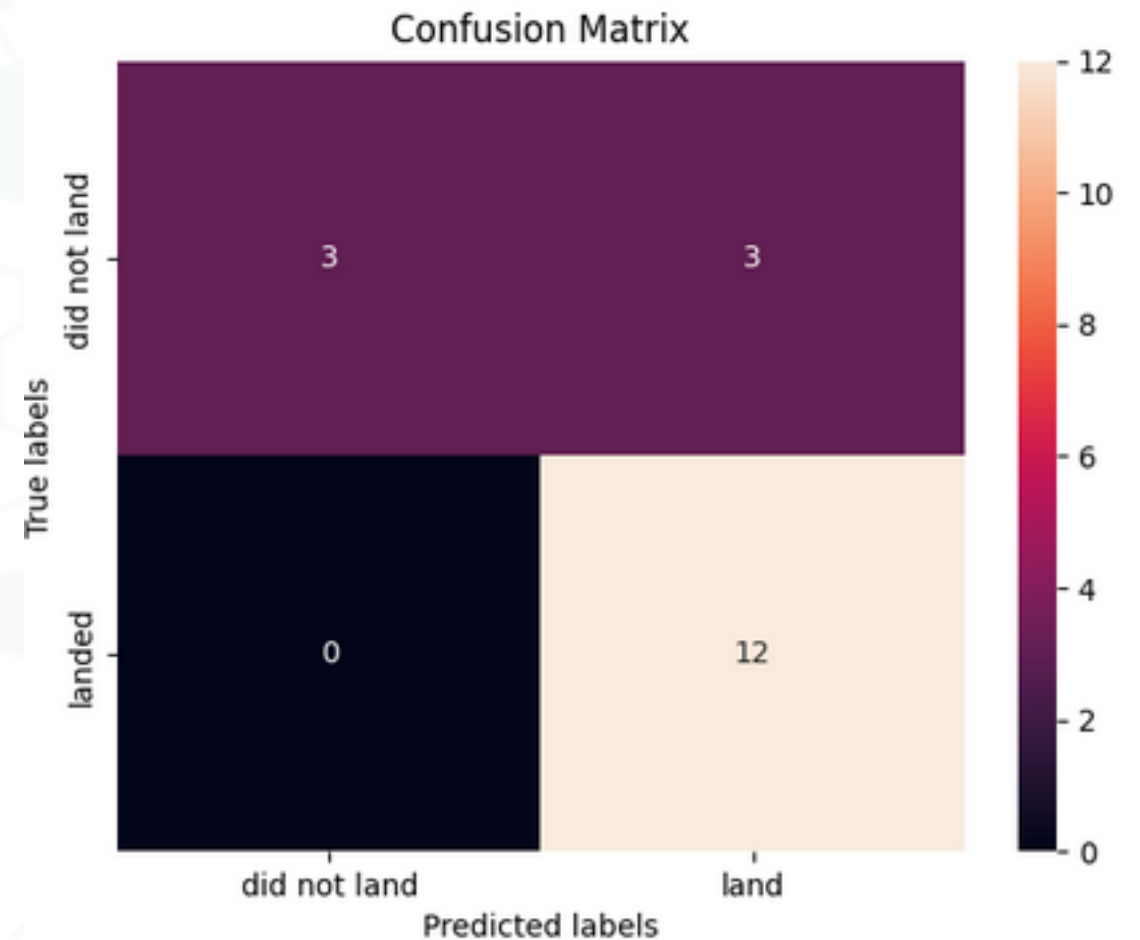
```
0.8888888888888888
```



K Nearest Neighbors

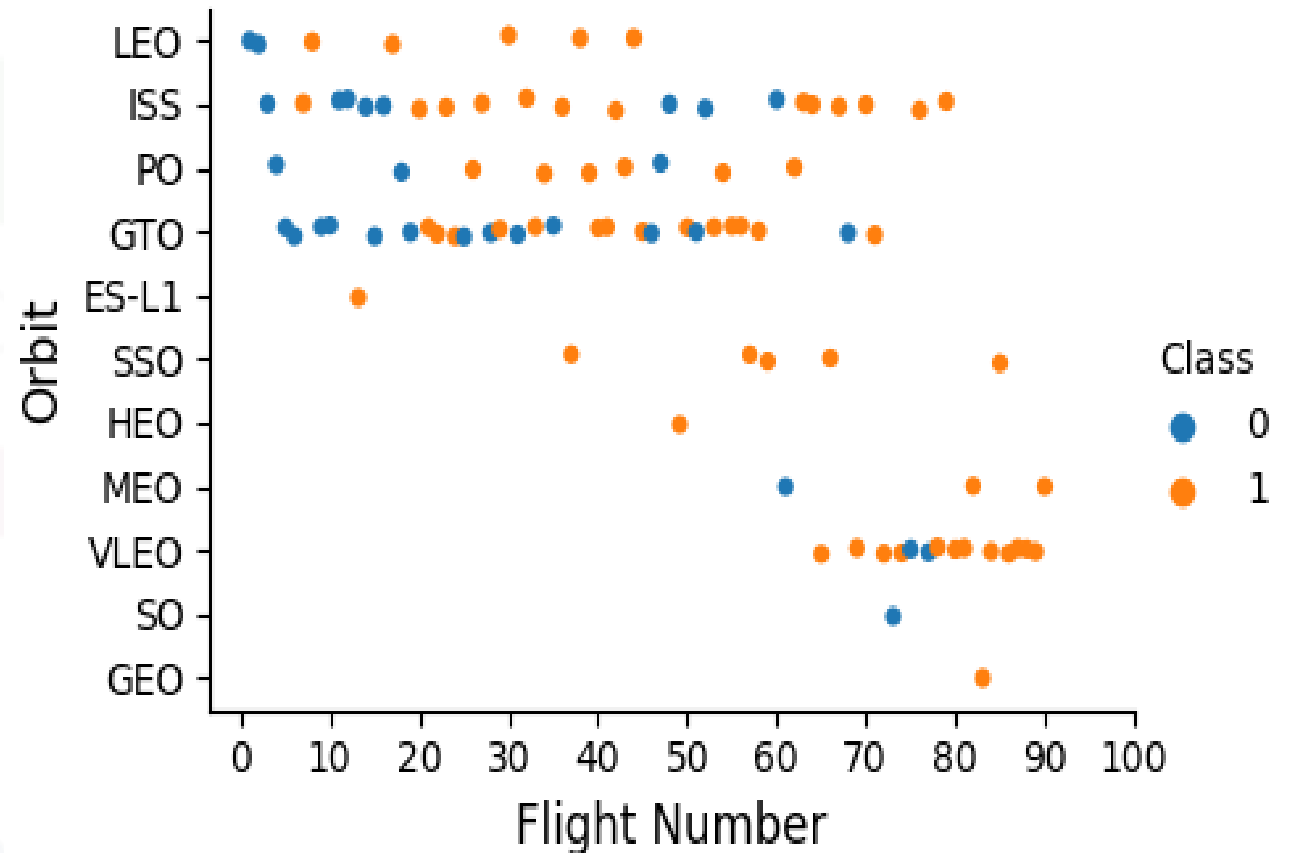
```
1 knn_cv.score(X_test, Y_test)
```

```
0.8333333333333333
```



DISCUSSION

- Earlier flight launches are mostly unsuccessful.
- Success of launching and landing improves with time and experiences.



Success Rate Of Different Launch Site

Launch Site	Success rate (%)	Failure rate (%)
KSC LC 39A	76.9	23.1
CCAFS LC 40	73.1	26.9
VAFB SLC 4E	60	40
CCAFS SLC 40	57.1	42.9

CONCLUSION



- Logistic regression performs best with a prediction accuracy of 94.44%
- Rocket with lighter payload have higher success rate than rocket with heavier payload
- The landing success probability increases with time.
- KSC LC 39A site has the highest launch success rate of 76%
- Orbit ES L1, SSO, HEO, and GEO has the best success rate of 100%