

3344 1/201 14930 31370 555 2032 11580 2462/ 26856 2925 18851 19101 19976 1/700 15237 30983 1057 1 16229 21725 14569 17174 21286 26407 26948 18974 4083 27513 24393 2146 30827 15237 30983 324393 2146 30827 1752 31970 323030 1 162.9 21725 14509 17174 2.1286 26407 26948 18974 4083 27513 24393 2140 30621 13237 19849 28126 28127 3 32105 19229 29352 14404 119039 13471 11295 10665 31579 1029 6835 1753 24393 2459 2535 409 2535 4 99.50 32140 19229 29352 14404 110039 13471 11295 10605 31579 1029 0835 1755 11079 22265 2535 10779 8321 17787 20254 14651 12938 32286 6542 71880 14247 5410 808 28837 24755 2408 32094 3372 14590 32093 46730 32094 320 0254 14051 12958 32280 0542 71880 14247 5410 808 28837 24755 2408 52094 5572 14536 2533 10775 6044 19692 2056 0759 26728 1070 15116 24337 9:22140 6750 17911 2466 9897 9519 16321 10443 4570 16087 16727 0000 16173 10175 1017 9/59 26/28 1070 15110 24337 9.22140 6/50 1/911 2460 9897 9519 10321 10443 4370 10067 1072 12983 22114 1884 28270 1877 20508 22927 14799 30962 214620 20224 24815 8955 23869 13689 16171 13457 3273 20224 24815 8955 23869 13689 16171 13457 3273 20224 24815 8955 23869 13689 16171 13457 3273 20224 24815 8955 23869 13689 16171 13457 3273 20224 24815 8955 23869 13689 16171 13457 3273 20224 24815 8955 23869 13689 16171 13457 3273 20224 24815 8955 23869 13689 16171 13457 3273 20224 24815 8955 23869 16171 1617 287 20308 72927 14799 30902 2.14620 20224 24815 8955 23609 13069 10171 13437 3673 20811 7531 8351 4987 27634 71 28422 19573 17294 16449 1321 3413 17063 24673 4947 18057 14195 14594 7073 20811 7531 8351 4987 27634 7073 20811 1 30409 31774 30238 8405 14568 3289 15204 29788 28827 26521 23516 24082 14945 15567 28394 28186 109 47270 27605 17t19131 6271 17024 6338 11930 8940 25939 15670 8375 7015 18353 2551 22823



Data Science Capstone Project

Roberto Aguirre

SPACEX

OUTLINE

Executive Summary

Introduction

Methodology

Results

Conclusion



Executive Sumary

Executive Summary

Summary of methodologies

- Data collection
- Data wrangling
- Exploratory Data Analysis with Data Visualization
- Exploratory Data Analysis with SQL
- Building an interactive map with Folium
- Building a Dashboard with Plotly Dash
- Predictive analysis (classification)

Summary of all results

- Exploratory Data Analysis results
- Interactive analytics demo in screenshots
- Predictive analysis results



Introduction

Project background and context



The most prosperous business of the commercial space era, SpaceX has reduced the cost of space travel. On its website, the firm promotes Falcon 9 rocket flights, which start at 62 million dollars; in comparison, other suppliers charge up to 165 million dollars per launch; a large portion of the cost savings are attributable to SpaceX's ability to reuse the first stage. Thus, we can calculate the cost of a launch if we can ascertain if the first stage will land. We are going to make a prediction about whether SpaceX will reuse the first stage based on publicly available data and learning models. Currently, the need has arisen to provide affordable space traveling. In this scenario, the world known company SpaceX has the business problem of estimating the cost of a Falcon 9 launch, which is determined if the first stage of Falcon 9 will land.

Questions to be answered

- What impact do factors like orbits, payload mass, launch site, and number of flights have on the first stage landing's success?
- Does the percentage of successful landings rise with time?
- What is the best algorithm that can be used for binary classification

Methodology

Methodology



Data collection methodology:

- Primary data source: SpaceX API.
- Secondary data source: data available on Wikipedia.

Perform data wrangling

- SpaceX API data has a relatively easy data wrangling, only filtering Falcon 9 data and missing data quick handling. Data available on Wikipedia required HTML syntaxis cleaning, and a HTML parse to Python data frame.
- Perform exploratory data analysis (EDA) using visualization and SQL.
- Perform interactive visual analytics using Folium and Plotly Dash.
- Perform predictive analysis using classification models.
- A standard mythology for fitting and testing the classification models were carried out.

Data collection

A combination of web scraping data from a table in SpaceX's Wikipedia entry and API queries from the company's REST API were used in the data collection procedure.

In order to obtain comprehensive information about the launches for a more in-depth analysis, we had to employ both of these data collection techniques.

Data Columns are obtained by using SpaceX REST API:
FlightNumber, Date, BoosterVersion, PayloadMass,
Orbit, LaunchSite, Outcome, Flights, GridFins,
Reused, Legs, LandingPad, Block, ReusedCount, Serial,
Longitude, Latitude

Data Columns are obtained by using Wikipedia Web Scraping: Flight No., Launch site, Payload, PayloadMass, Orbit, Customer, Launch outcome, Version Booster, Booster landing, Date, Time

Data extraction

Processing Stage

Data parsing to a phyton frame

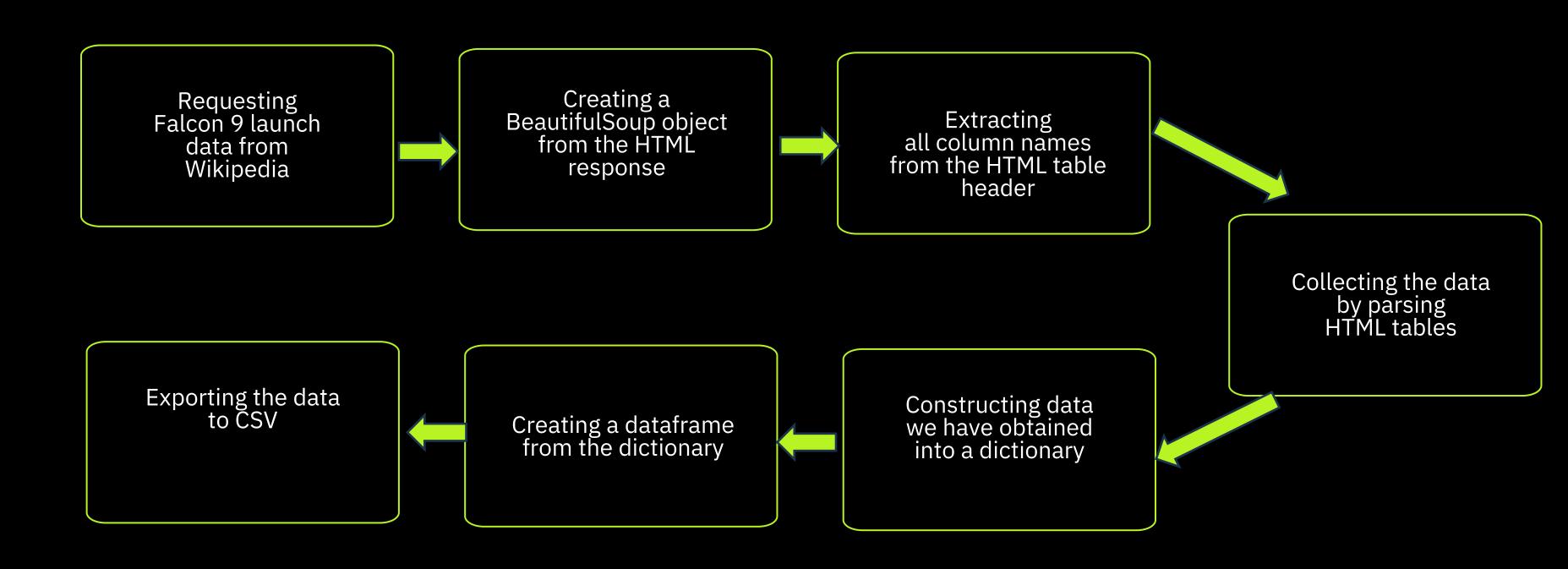
Export Stage

Export results dataframe to csv

Data collection — SpaceX API

Decoding the Requesting needed information about Requesting response content Constructing data using .json() and rocket launch the launches from we have obtained turning it into a data from SpaceX API into a dictionary dataframe using SpaceX API by applying custom functions .json_normalize() Exporting the data Replacing missing values of Payload Creating a dataframe from the dictionary Filtering the to CSV dataframe to only Mass column with include Falcon 9 calculated .mean() launches for this column

Data collection – Web scraping



Data wrangling

There are other instances in the data set where the booster failed to land properly. There are instances where a landing attempt was made but was unsuccessful due to an accident; for instance, True Ocean denotes a successful landing in a particular area of the ocean, whereas False Ocean denotes an unsuccessful landing in a particular area of the ocean.

When a mission is declared successful and lands on a ground pad, it is said to have true RTLS. A mission outcome that was unsuccessfully landed on a ground pad is indicated by a false RTLS. A successful landing of the mission's outcome on a drone ship is referred to as true ASDS. An unsuccessful landing on a drone ship is indicated by a false ASDS.

Mostly, we translate those results into Training Labels, where "0" denotes an unsuccessful landing and "1" indicates a successful booster landing.

Perform exploratory Data Analysis and determine Training Labels



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

Exporting the data to CSV

EDA with data visualization

Charts were plotted:

Flight Number vs. Payload Mass, Flight Number vs. Launch Site, Payload Mass vs. Launch Site, Orbit Type vs. Success Rate, Flight Number vs. Orbit Type, Payload Mass vs Orbit Type and Success Rate Yearly Trend.

Disk plots display the correlation between variables. These could be included in a machine learning model if a relationship is found.

Bar charts display discrete category comparisons. The intention is to illustrate the correlation between a measured value and the particular categories under comparison.

Line charts show trends in data over time (time series).

- Generally speaking, the graph type selected for each feature relationship analysis was directly related to the necessity of understanding the connection among three or two features.
- The bar plot lets us analyze the relationship between success rate and orbit type. Lastly, a line chart shows the relationship between year and average success rate.

Performed SQL queries:

- Listing the names of the space mission's distinct launch locations.
- Showing the average payload mass carried by rocket version F9 v1.1; Showing the total payload mass carried by boosters launched by NASA (CRS); Showing five records where launch sites start with the term "CCA."
- A list of the names of the boosters that have successfully landed on a drone ship with a payload mass of more than 4,000 but less than 6,000, along with the date of the first successful landing outcome in a ground pad
- Counting the overall number of mission outcomes that were successful and unsuccessful.
 Listing the names of the booster versions which have carried the maximum payload mass
- Listing the failed landing outcomes in drone ship, their booster versions and launch site names for the months in year 2015
- Ranking the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20 in descending order

Build an interactive map with Folium

Markers of all Launch Sites:

- Added a NASA Johnson Space Center marker with a circle, pop-up label, and text label, utilizing the center's latitude and longitude as a starting point.

Coloured Markers of the launch outcomes for each Launch Site:

In order to determine which launch sites have comparatively high success rates, colored markers representing successful (Green) and unsuccessful (Red) launches were added using Marker Cluster.

Distances between a Launch Site to its proximities:

- Added coloured Lines to show distances between the Launch Site KSC LC-39A (as an example) and its proximities like Railway, Highway, Coastline and Closest City.

Build a Dashboard with Plotly Dash

Launch Sites Dropdown List:

- Added a dropdown list to enable Launch Site selection.

Pie Chart showing Success Launches (All Sites/Certain Site): Included a pie chart that displays the total number of successful launches across all sites as well as the success versus failure counts for each site that was chosen as a launch location.

Slider of Payload Mass Range:

- Added a slider to select Payload range.

Scatter Chart of Payload Mass vs. Success Rate for the different Booster Versions:

- Added a scatter chart to show the correlation between Payload and Launch Succes

Predictive analysis (Classification)

Creating a NumPy array from the column "Class" in data Standardizing the data with StandardScaler, then fitting and transforming it

Splitting the data into training and testing sets with train_test_split function

Creating a
GridSearchCV object
with cv = 10 to find
the best parameters

Finding the method performs best by examining the Jaccard_score and F1_score metrics

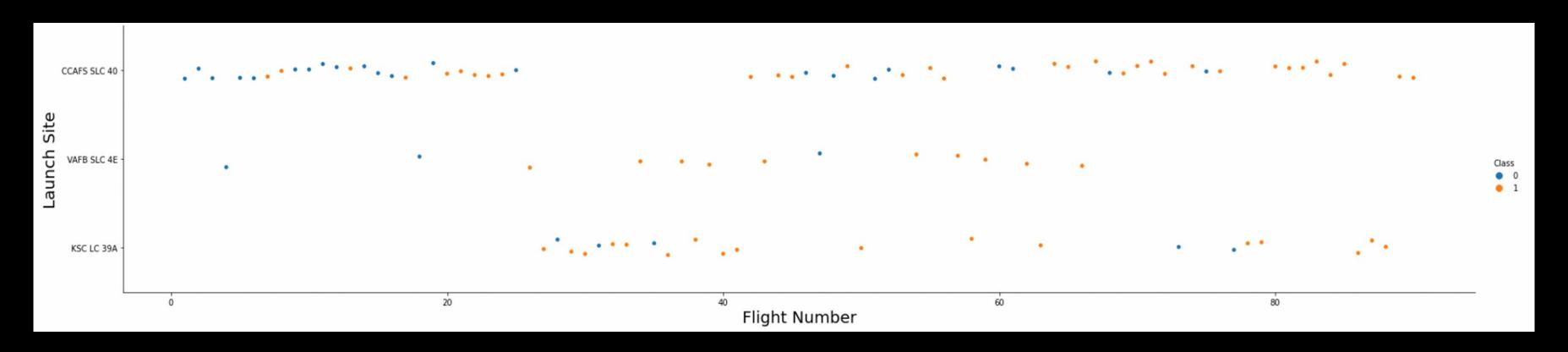
Examining the confusion matrix for all models

Calculating the accuracy on the test data using the method .score() for all models

Applying
GridSearchCV
on LogReg, SVM,
Decision Tree, and
KNN models

RESULTS

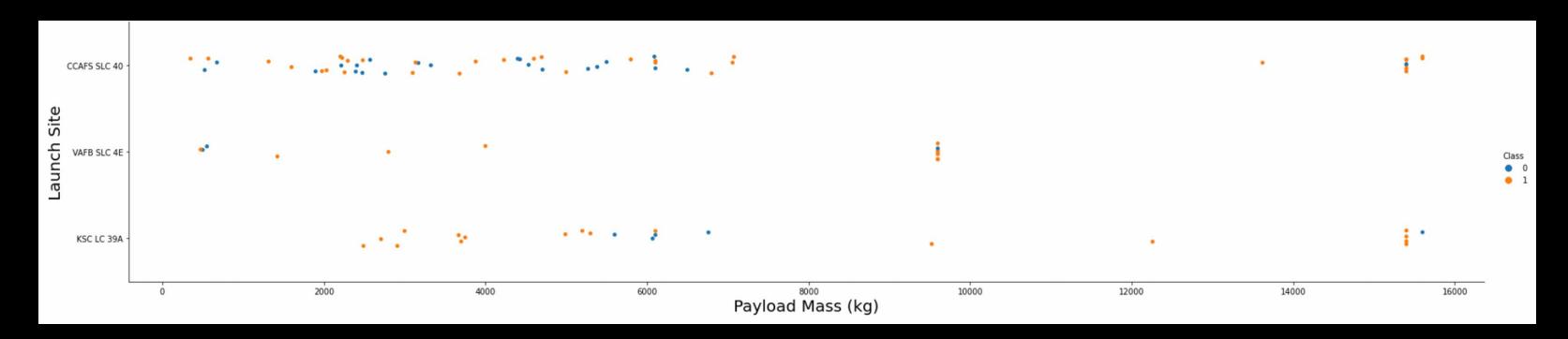
Flight Number vs. Launch Site



All of the most recent flights were successful, whereas none of the earlier ones were.

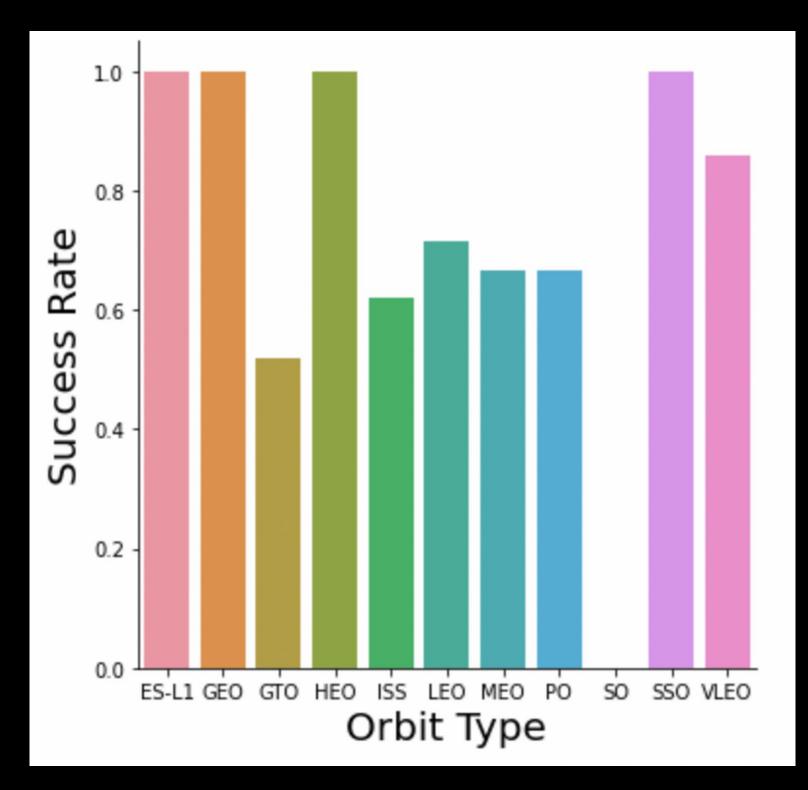
Approximately 50% of all launches occur at the CCAFS SLC 40 launch location. Success rates are greater at KSC LC 39A and VAFB SLC 4E. One can expect that the success rate increases with each fresh launch.

Payload vs. Launch Site



The success rate increases with payload mass at each launch site. With payload masses exceeding 7000 kg, the majority of launches were successful. Under 5500 kg, KSC LC 39A also has a 100% success rate for payload mass.

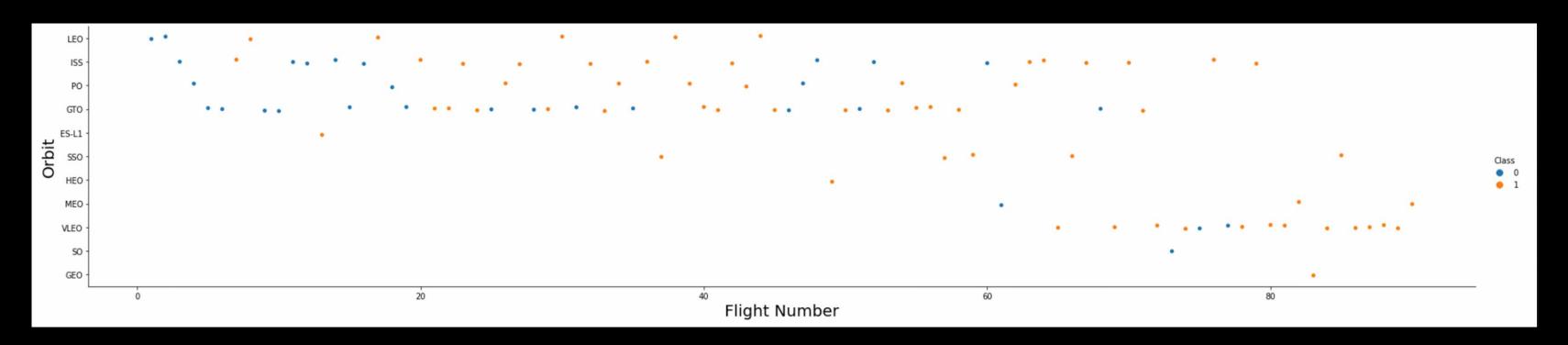
Success rate vs. Orbit type



Orbits with 100% success rate:

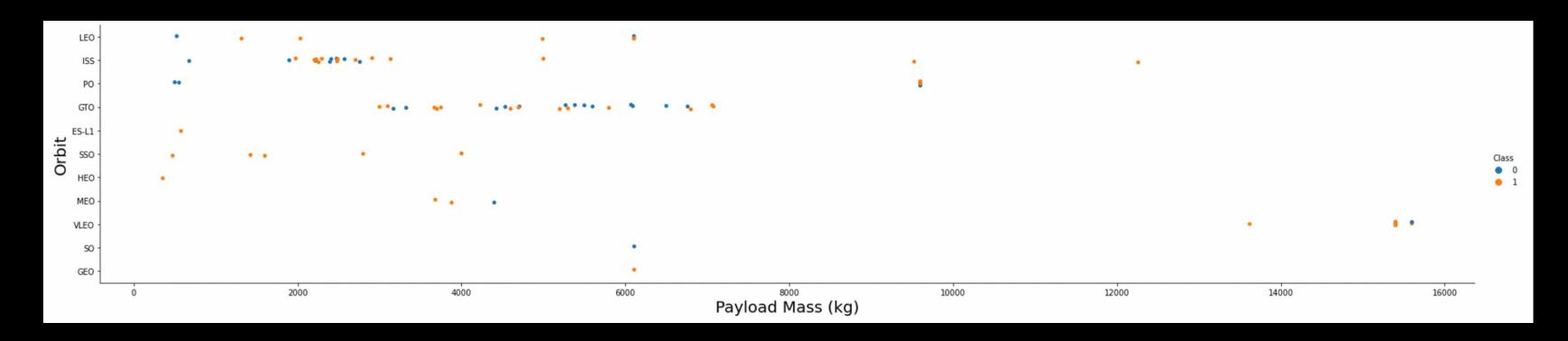
- ES-L1, GEO, HEO, SSO Orbits with 0% success rate:
- SO Orbits with success rate between 50% and 85%:
- GTO, ISS, LEO, MEO, PO

Flight Number vs. Orbit type



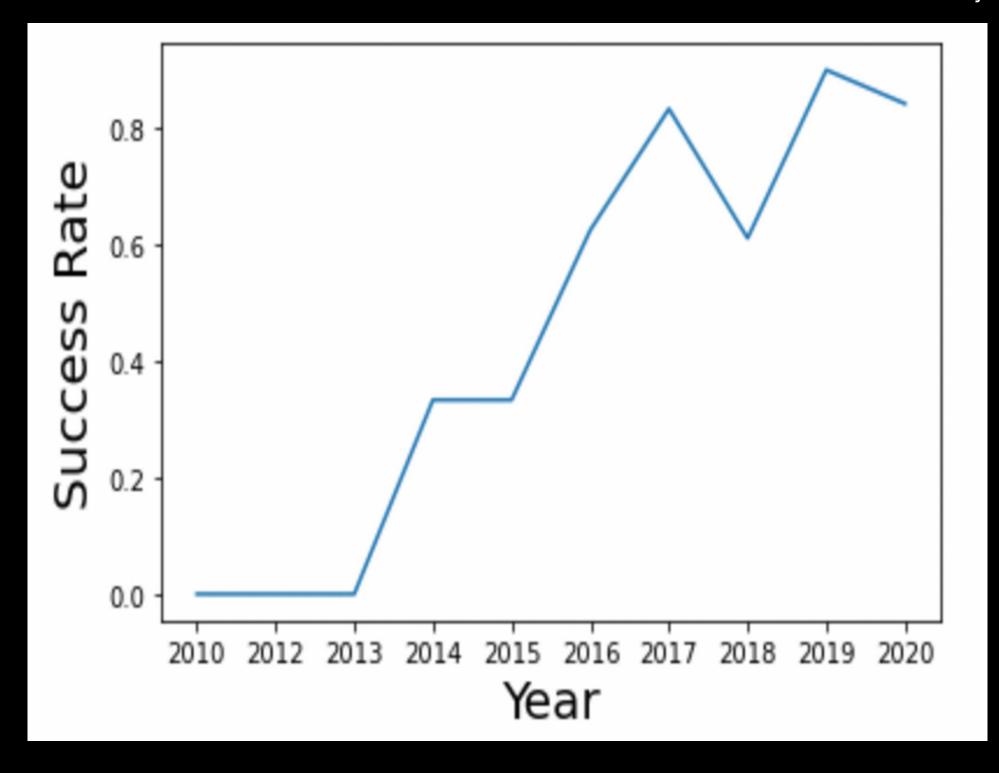
While in GTO orbit, there doesn't seem to be any correlation between the number of flights and success; however, in LEO orbit, it does appear to be related.

Payload Mass vs. Orbit type



Heavy payloads have a negative influence on GTO orbits and positive on GTO and Polar LEO (ISS) orbits.

Launch success yearly trend



The success rate since 2013 kept increasing till 2020.

All launch site names

```
In [4]: %sql select distinct launch_site from SPACEXDATASET;

* ibm_db_sa://wzf08322:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqblod8lcg.databases.appdomain.cloud:31198/bludb
Done.

Out[4]: launch_site

CCAFS LC-40

CCAFS SLC-40

KSC LC-39A

VAFB SLC-4E
```

putting the names of the space mission's distinct launch sites on display.

Launch site names begin with `CCA`

%sql select * from SPACEXDATASET where launch site like 'CCA%' limit 5; In [5]: * ibm db sa://wzf08322:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqb1od8lcg.databases.appdomain.cloud:31198/bludb Done. Out[5]: DATE booster_version | launch_site time_utc_ pavload payload mass kg orbit | customer mission outcome landing outcome **Dragon Spacecraft** 2010-CCAFS LC-LEO | SpaceX 18:45:00 F9 v1.0 B0003 Success Failure (parachute) 06-04 40 Qualification Unit Dragon demo flight C1, two NASA **LEO** CCAFS LC-2010-CubeSats, barrel of Brouere (COTS) 15:43:00 F9 v1.0 B0004 0 Success Failure (parachute) 12-08 40 (ISS) **NRO** cheese CCAFS LC-**NASA** 2012-LEO Dragon demo flight C2 07:44:00 F9 v1.0 B0005 525 Success No attempt (COTS) 05-22 40 (ISS) CCAFS LC-2012-**LEO NASA** SpaceX CRS-1 00:35:00 F9 v1.0 B0006 500 Success No attempt 10-08 40 (ISS) (CRS) CCAFS LC-**NASA** 2013-LEO SpaceX CRS-2 15:10:00 F9 v1.0 B0007 677 Success No attempt (CRS) (ISS) 03-01 40

Displaying 5 records where launch sites begin with the string 'CCA'.

Total payload mass

Displaying the total tonnage of payload carried by NASA-launched rockets (CRS)

Average payload mass by F9 v1.1

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

First successful ground landing date

Stating the date on which the ground pad had its first successful landing.

Successful drone ship landing with payload between 4000 and 6000

A list of the boosters' names that have been successful in drone ships with payload masses over 4,000 but under 6,000

Total number of successful and failure mission outcomes

```
In [10]: %sql select mission_outcome, count(*) as total_number from SPACEXDATASET group by mission_outcome;

* ibm_db_sa://wzf08322:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqblod8lcg.databases.appdomain.cloud:31198/bludb Done.

Out[10]: mission_outcome total_number
Failure (in flight) 1
Success 99
Success (payload status unclear) 1
```

Listing the total number of successful and failure mission outcomes.

Boosters carried maximum payload

```
In [11]: \%sql select booster version from SPACEXDATASET where payload mass kg = (select max(payload mass kg) from SPACEXDATASET);
           * ibm db sa://wzf08322:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqb1od8lcg.databases.appdomain.cloud:31198/bludb
          Done.
Out[11]:
          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
```

Listing the names of the booster versions which have carried the maximum payload mass.

2015 launch records

```
In [12]: %%sql select monthname(date) as month, date, booster_version, launch_site, landing_outcome from SPACEXDATASET
where landing_outcome = 'Failure (drone ship)' and year(date)=2015;

* ibm_db_sa://wzf08322:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.

Out[12]: MONTH DATE | booster_version | launch_site | landing_outcome |
January | 2015-01-10 | F9 v1.1 B1012 | CCAFS LC-40 | Failure (drone ship) |
April | 2015-04-14 | F9 v1.1 B1015 | CCAFS LC-40 | Failure (drone ship)
```

A list of the drone ship's unsuccessful landings along with the names of the launch sites and the versions of their boosters for each month in 2015.

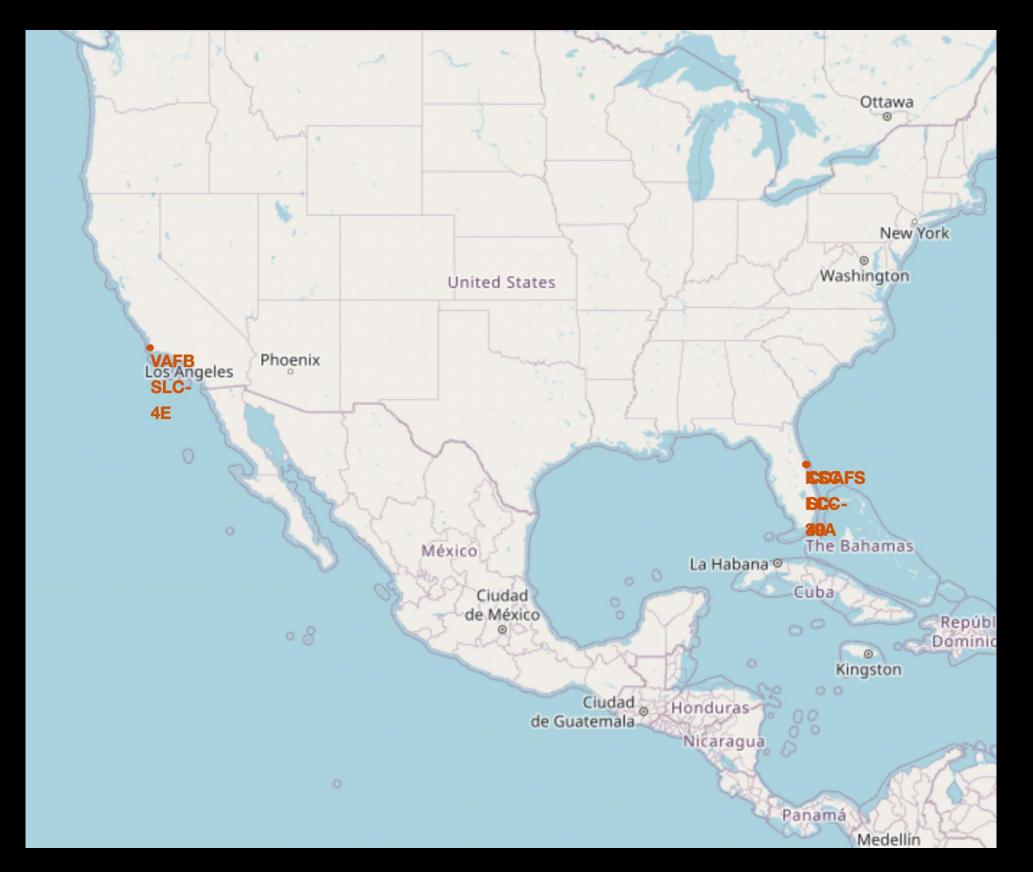
Rank success count between 2010-06-04 and 2017-03-20

```
In [13]: %%sql select landing outcome, count(*) as count outcomes from SPACEXDATASET
                where date between '2010-06-04' and '2017-03-20'
                group by landing outcome
                order by count outcomes desc;
           * ibm db sa://wzf08322:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqb1od8lcg.databases.appdomain.cloud:31198/bludb
          Done.
Out[13]:
          landing outcome
                              count outcomes
          No attempt
          Failure (drone ship)
          Success (drone ship)
          Controlled (ocean)
          Success (ground pad)
          Failure (parachute)
          Uncontrolled (ocean)
```

Precluded (drone ship)

Sorting the number of landing results (e.g., ground pad success or drone ship failure) between 2010-06-04 and 2017-03-20 in descending order.

All launch sites' location markers on a global map

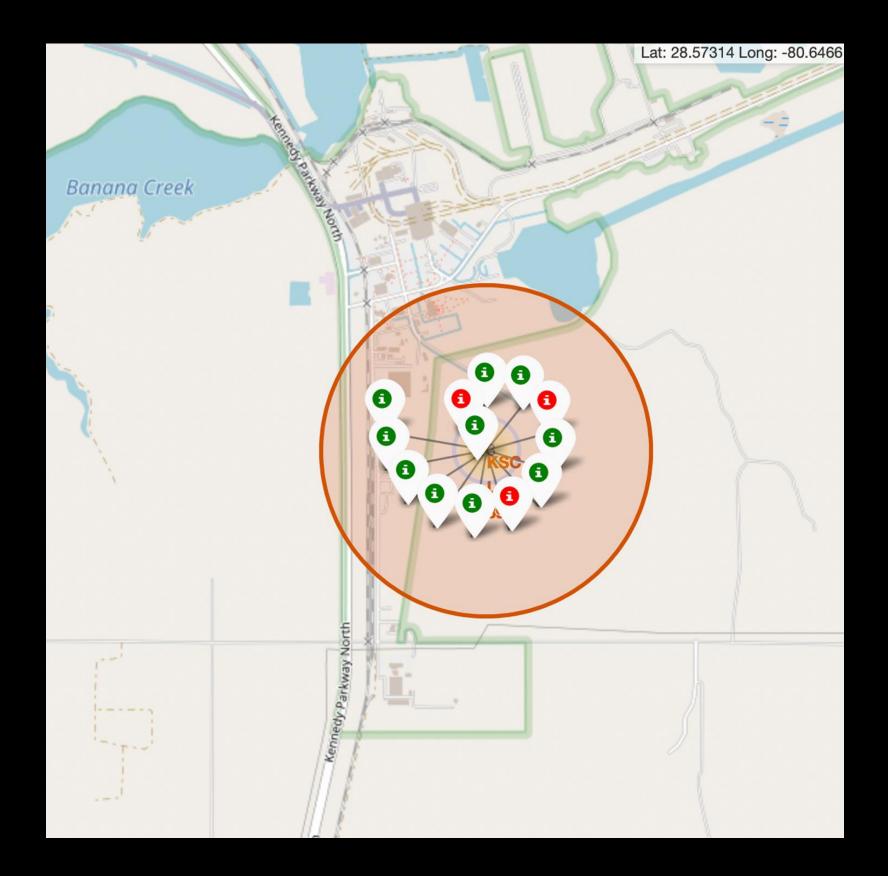


The majority of launch sites are close to the Equator. At the equator, land moves more quickly than it does anywhere else on Earth's surface. At the equator, everything on Earth is already travelling at a speed of 1670 km/h.

A ship launched from the equator travels through space at the same speed as it did prior to launch, as well as around the planet. Inertia is the cause of this. This velocity will assist the spacecraft in maintaining a sufficient speed to remain in orbit.

All launch sites are in very close proximity to the coast, while launching rockets towards the ocean it minimises the risk of having any debris dropping or exploding near people.

Colour-labeled launch records on the map



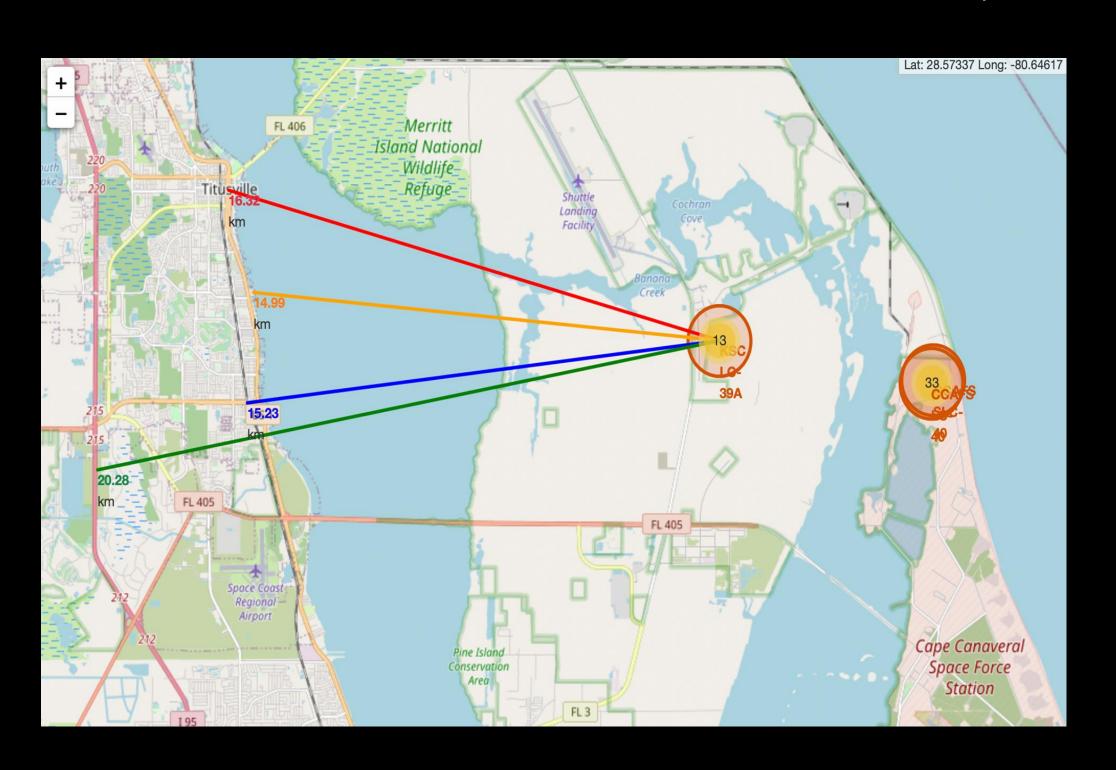
From the colour-labeled markers we should be able to easily identify which launch sites have relatively high success rates.

- Green Marker = Successful

- Launch
- Red Marker = Failed Launch

Launch Site KSC LC-39A has a very high Success Rate.

Distance from the launch site KSC LC-39A to its proximities



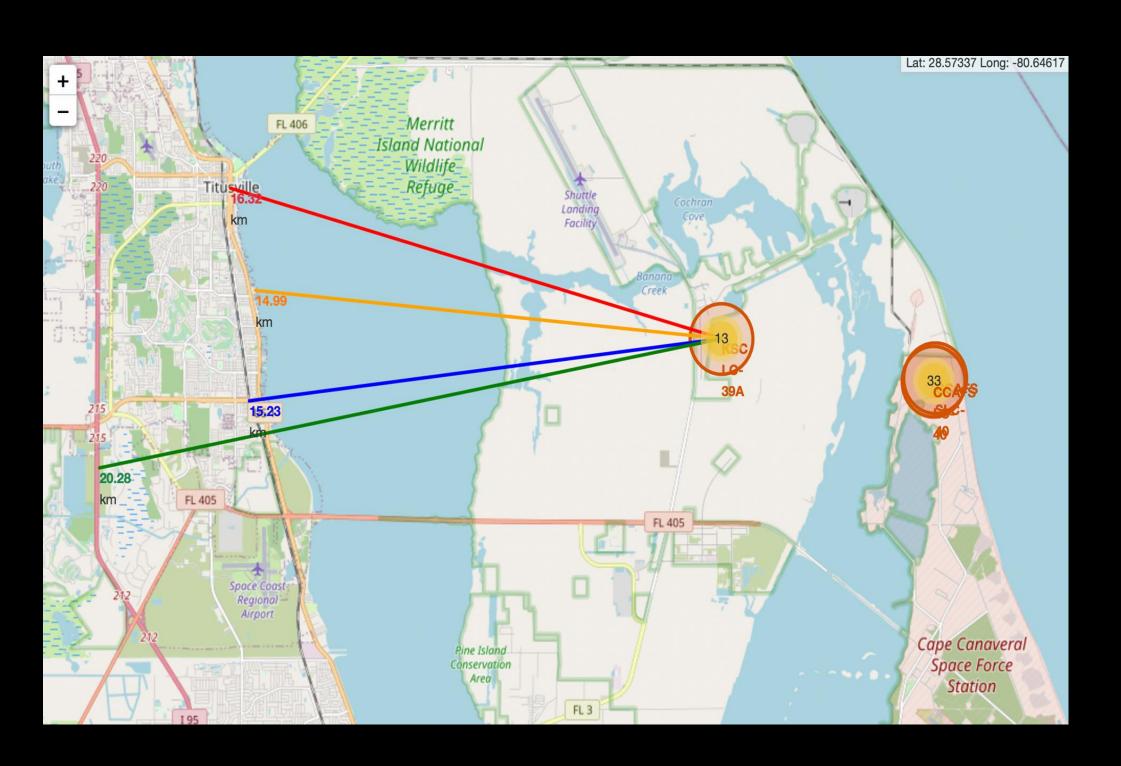
From the visual analysis of the launch site KSC LC-39A we can clearly see that it is:

- relative close to railway (15.23 km)
- rélative close to highway (20.28 km)
- rélative close to coastline (14.99 km)

Additionally, Titusville, the closest city (16.32 km) is relatively close to the launch site KSC LC-39A.

A failed rocket can travel up to 15-20 km in a matter of seconds due to its high speed. It might pose a threat to densely inhabited areas.

Distance from the launch site KSC LC-39A to its proximities



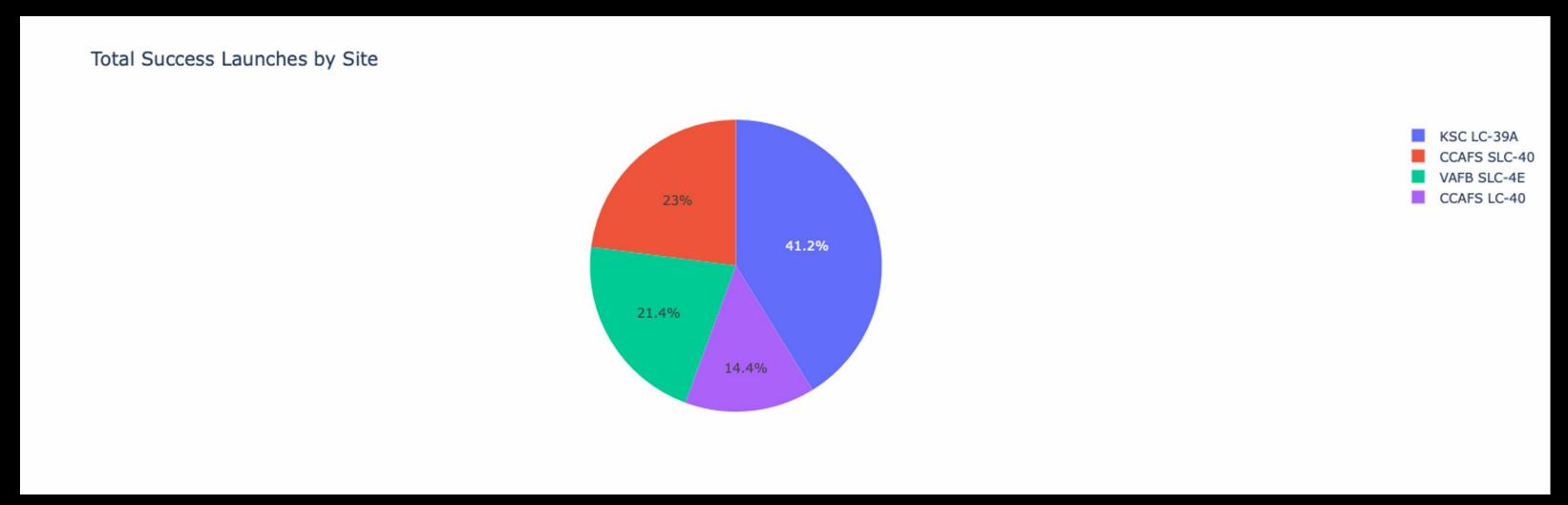
From the visual analysis of the launch site KSC LC-39A we can clearly see that it is:

- relative close to railway (15.23 km)
- rélative close to highway (20.28 km)
- relative close to coastline (14.99 km)

Also the launch site KSC LC-39A is relative close to its closest city Titusville (16.32 km).

Failed rocket with its high speed can cover distances like 15-20 km in few seconds. It could be potentially dangerous to populated areas.

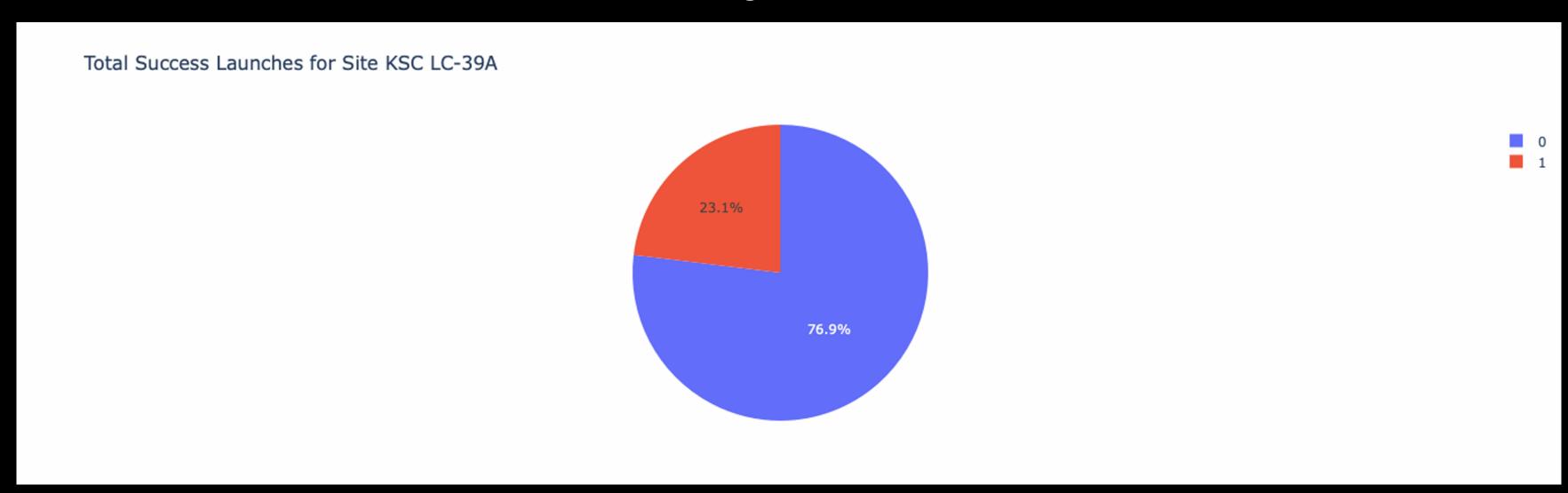
Launch success count for all sites



The chart clearly shows that from all the sites, KSC LC-39A has the most successful launches.

Build a Dashboard with Plotly Dash

Launch site with highest launch success ratio



KSC LC-39A has the highest launch success rate (76.9%) with 10 successful and only 3 failed landings.

Build a Dashboard with Plotly Dash

Payload Mass vs. Launch Outcome for all sites





The charts show that payloads between 2000 and 5500 kg have the highest success rate.

Predictive analysis (Classification)

Classification Accuracy

Scores and Accuracy of the Test Set

	LogReg	SVM	Tree	KNN
Jaccard_Score	0.800000	0.800000	0.800000	0.800000
F1_Score	0.888889	0.888889	0.888889	0.888889
Accuracy	0.833333	0.833333	0.833333	0.833333

Scores and Accuracy of the Entire Data Set

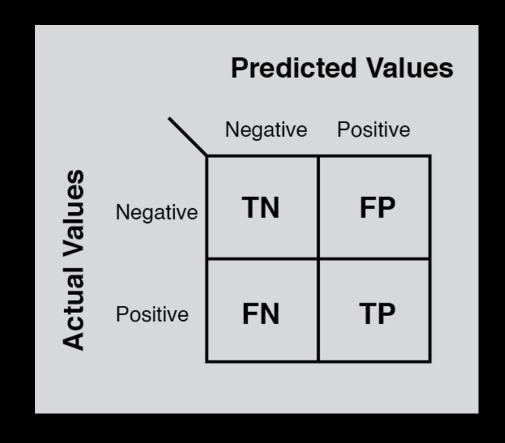
	LogReg	SVM	Tree	KNN
Jaccard_Score	0.833333	0.845070	0.882353	0.819444
F1_Score	0.909091	0.916031	0.937500	0.900763
Accuracy	0.866667	0.877778	0.911111	0.855556

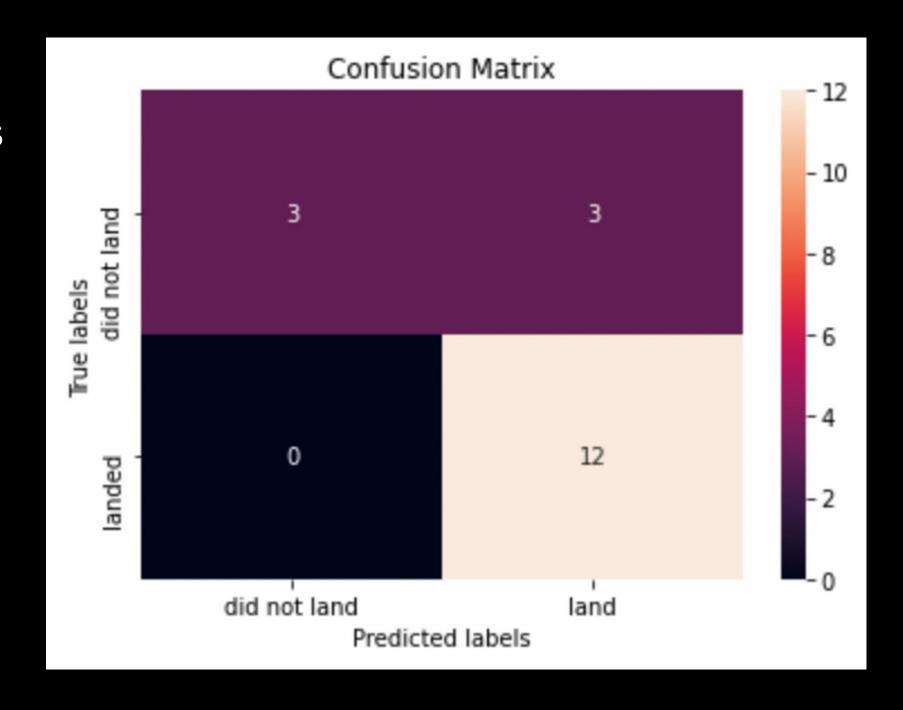
- We are unable to verify the optimal strategy based on the Test Set scores.
- The short test sample size (18 samples) may be the cause of the same test set scores. As a result, we used the entire dataset to test every approach.
- The Decision Tree Model is the best model, as confirmed by the scores of the entire dataset. This model offers the highest accuracy as well as higher scores.

Predictive analysis (Classification)

Confusion Matrix

It is evident by looking at the confusion matrix that logistic regression is capable of differentiating between the various classes. It is evident that false positives are the main issue.





CONCLUSIONS

Decision Tree Model is the best algorithm for this dataset.

The success rate of launches increases over the years.

Launches with a low payload mass show better results than launches with a larger payload mass.

KSC LC-39A has the highest success rate of the launches from all the sites.

Most of launch sites are in proximity to the Equator line and all the sites are in very close proximity to the coast.

Orbits ES-L1, GEO, HEO and SSO have 100% success rate.