Executive Summary slide

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.

We first collect the data using python web scraping method then use sql to manipulate the data and using python to do the EDA and visualization, including folium map and dash interactive visualization and at last, we will train machine learning model to predict the successful landing rate.

What we find is that the Flight number has positive correlation with the success rate, payload usually negative correlation. And all four machine learning classification methods we use give similar accuracy on the test data.

Introduction:

I am Ziqin He and in this project, we will predict if the Falcon 9 rocket first stage will land successfully. 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.

Use get request to do data collection and beautiful soup to do web scraping

Task 1: Request and parse the SpaceX launch data using the GET request ¶ To make the requested JSON results more consistent, we will use the following static response object for this project: [37]: ses_data.s3.us.cloud_object_storage.appdomain.cloud/IBM-D50321EN-SkillsNetwork/datasets/API_call_spacex_api.ison We should see that the request was successfull with the 200 status response code [38]: response.status_code [38]: 200 Now we decode the response content as a Json using . i son() and turn it into a Pandas dataframe using .json normalize() [39]: # Use json_normalize meethod to convert the json result into a dataframe data = pd.json normalize(response.json()) Using the dataframe data print the first 5 rows [40]: # Get the head of the dataframe data.head() static fire date utc static fire date unix net window details crev [{'time': 33 Engine failure at 33 1.142554e+09 False 0.0 5e9d0d95eda69955f709d1eb seconds and loss of engine failure'}]

2020 [edit] In late 2019, Gwynne Shotwell stated that SpaceX hoped for as many as 24 launches for Starlink satellites in 2020, [490] in addition to 14 or 15 non-Starlink launches. At 26 launches, 13 of which for Starlink satellites, Falcon 9 had its most prolific year, and Falcon rockets were second most prolific rocket family of 2020, only behind China's Long March rocket family. [491] Version. Launch Launch Booster Flight Payload Payload mass Orbit time (UTC) landing F9 B5 △ CCAES 7 January 2020 Success Starlink 2 v1.0 (60 satellites) 15,600 kg (34,400 lb)[5] 02:19:21[492] B1049.4 SLC-40 Third large batch and second operational flight of Starlink constellation. One of the 60 satellites included a test coating to make the satellite less reflective, and thus less likely to interfere with ground-based astronomical observations. Crew Dragon in-flight abort test[498 12.050 kg (26.570 lb) No attempt 15:30[494] B1046.4 (Dragon C205.1) 79 An atmospheric test of the Dragon 2 abort system after Max Q. The capsule fired its SuperDraco engines, reached an apogee of 40 km (25 mi), deployed parachutes after reentry, and splashed down in the ocean 31 km (19 mi) downrange from the launch site. The test was previously slated to be accomplished with the Crew Dragon Demo-1 capsule, [498] but that test article exploded during a ground test of SuperDraco engines on 20 April 2019. [419] The abort test used the capsule originally intended for the first crewed flight. [499] As expected, the booster was destroyed by aerodynamic forces after the capsule aborted. [500] First flight of a Falcon 9 with only one functional stage — the second stage had a mass simulator in place of its engine. F9 B5 △ Starlink 3 v1.0 (60 satellites) 15,600 kg (34,400 lb) B1051.3 SLC-40 (drone ship) Third operational and fourth large batch of Starlink satellites, deployed in a circular 290 km (180 mi) orbit. One of the fairing halves was caught, while the other was fished out of the ocean. [502] F9 B5 △ Failure 15,600 kg (34,400 lb)[5] Starlink 4 v1.0 (60 satellites) Space) B1056.4 SLC-40 Fourth operational and fifth large batch of Starlink satellites. Used a new flight profile which deployed into a 212 km x 386 km (132 mi x 240 mi) elliptical orbit instead of launching into a circular orbit and firing the second stage engine twice. The first stage booster failed to land on the drone ship[504] due to incorrect wind data.[505] This was the first time a flight proven booster failed to land. 7 March 2020 F9 B5 △ Success 1.977 kg (4.359 lb)[507] NASA (CRS) 04:50[506] B1059.2 (Dragon C112.3 △) Last launch of phase 1 of the CRS contract. Carries Barlolomeo, an ESA platform for hosting external payloads onto ISS [509] Originally scheduled to launch on 2 March 2020, the launch date was pushed back due to a second stage engine failure. SpaceX decided to swap out the second stage instead of replacing the faulty part. [509] It was SpaceX's 50th successful landing of a first stage booster, the third flight of the Dragon C112 and the last launch of the cargo Dragon spacecraft. F9 B5 △ 18 March 2020 Failure Starlink 5 v1.0 (60 satellites) 15.600 kg (34.400 lb) 12:16^[510] B1048.5 LC-39A 83 Fifth operational launch of Starlink satellites. It was the first time a first stage booster flew for a fifth time and the second time the fairings were reused (Starlink flight in May 2019). [511] Towards the end of the first stage burn, the booster suffered premature shut down of an engine, the first of a Merlin 1D variant and first since the CRS-1 mission in October 2012. However, the payload still reached the targeted orbit [512] This was the second Starlink launch booster landing failure in a row, later revealed to be caused by residual cleaning fluid trapped inside a sensor. [513 F9 B5 △ Success Starlink 6 v1.0 (60 satellites) 19:30[514]

Using sql to do EDA, we find there are 4 launch sites. 101 trials and completed the mission 100 times. And we did other analysis such as which boosters carry more than 4000kg payloads and the success rate of landing during a period of time.

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

count	Outc		
1	Failure		
100	Success		

Sql: First time landing successfully. Average payload mass carried by booster version F9 v1.1 And the data analysis in a particular year for example the failure landing (drone ship) records in year 2015.

MIN(Date)

2015-12-22

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

**sql SELECT AVG(PAYLOAD_MASS__KG_) FROM SPACEXTABLE
WHERE Booster_Version LIKE 'F9 v1.1'

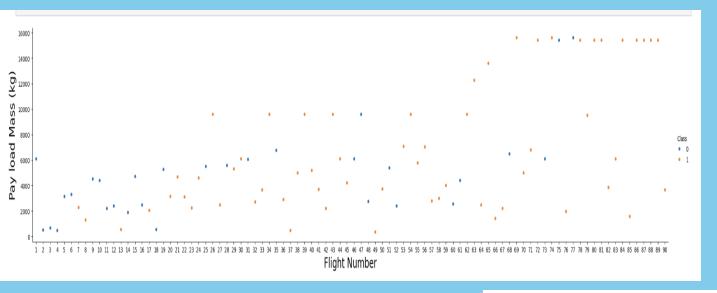
* sqlite://my_data1.db
Done.

AVG(PAYLOAD_MASS__KG_)

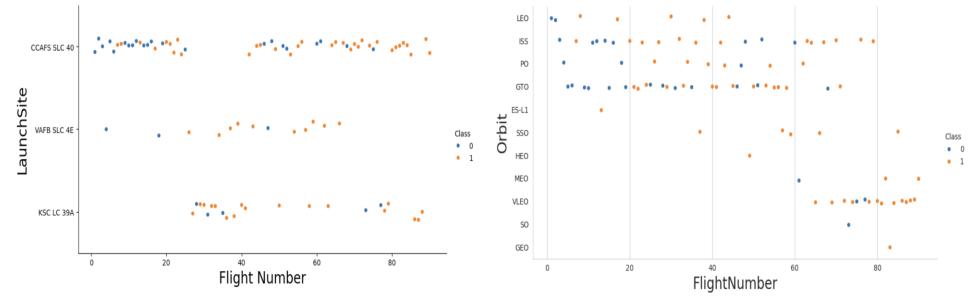
2928.4
```

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

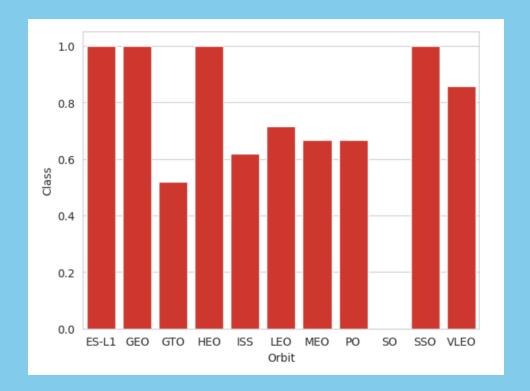
Python visualization EDA:

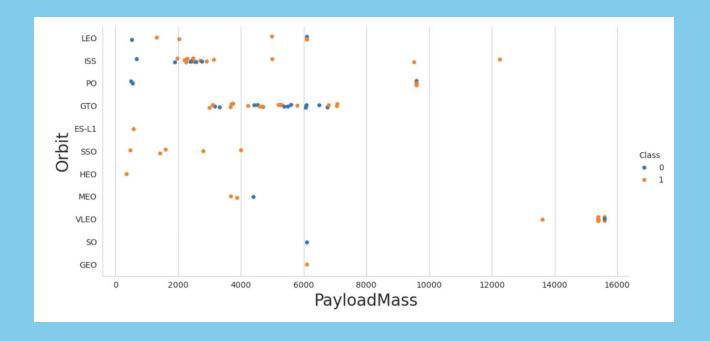


With larger Flight Number, the first stage seems to land with higher successful rate. And even with large pay load, high flight number still shows good landing result trend.



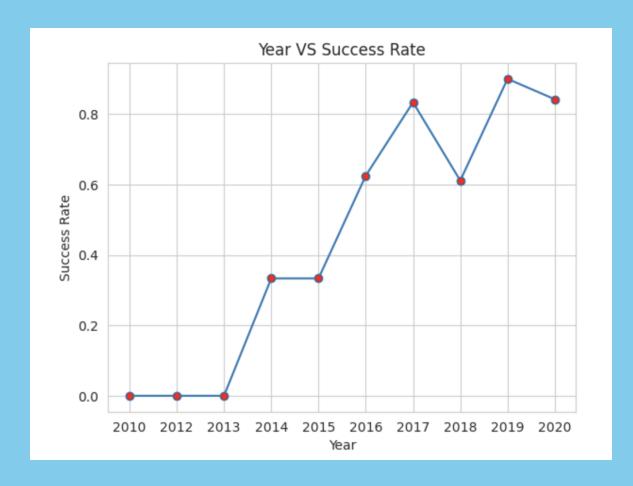
For different launch sites and orbits, larger flight number also shows better landing results.



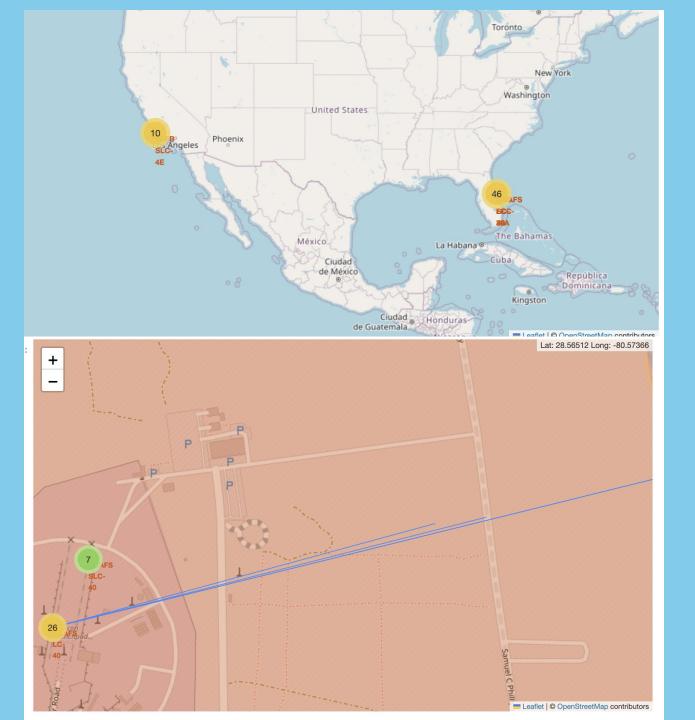


We plot the success rate with respect to different orbit. We see that with orbit ES-L1, GEO, HEO and SSO, the landings are all successful.

With heavy payloads the successful landing or positive landing rate are more for Polar, LEO and ISS. However, for GTO, it's difficult to distinguish between successful and unsuccessful landings as both outcomes are present.

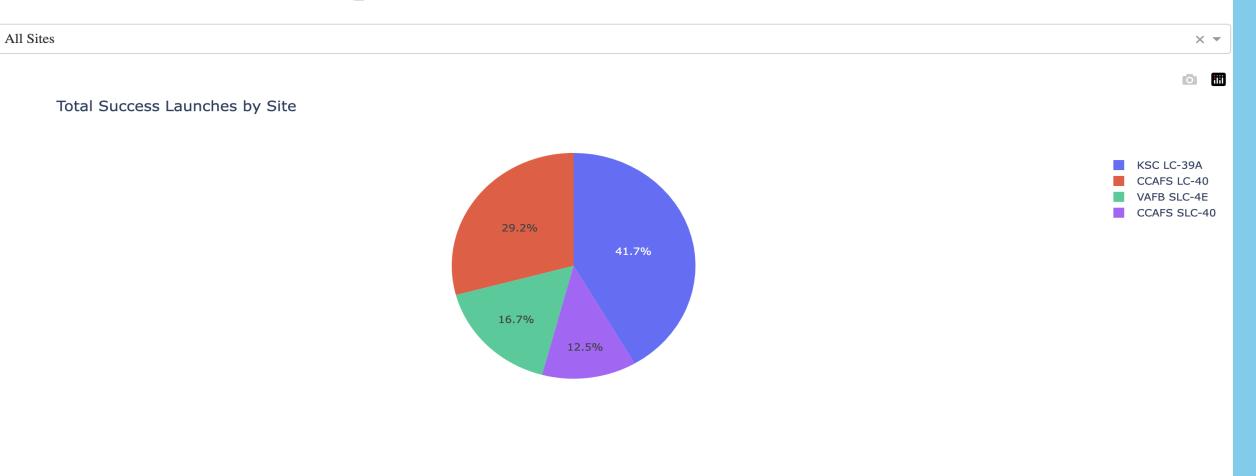


you can observe that the sucess rate since 2013 kept increasing till 2020



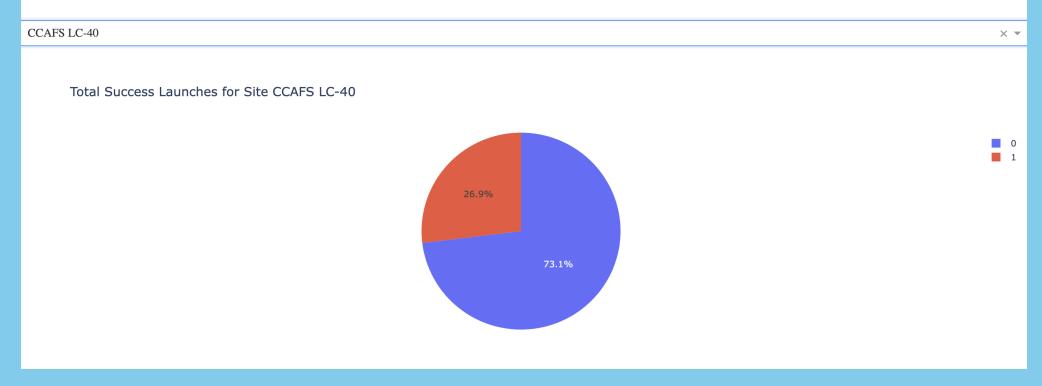
We use python folium to do the map visualization and we can see the launch sites are all close to the coastline. And we can check out how many landings at each launch sites with the line connecting the launch sites with nearest railway, highway, coastline and city.

SpaceX Launch Records Dashboard

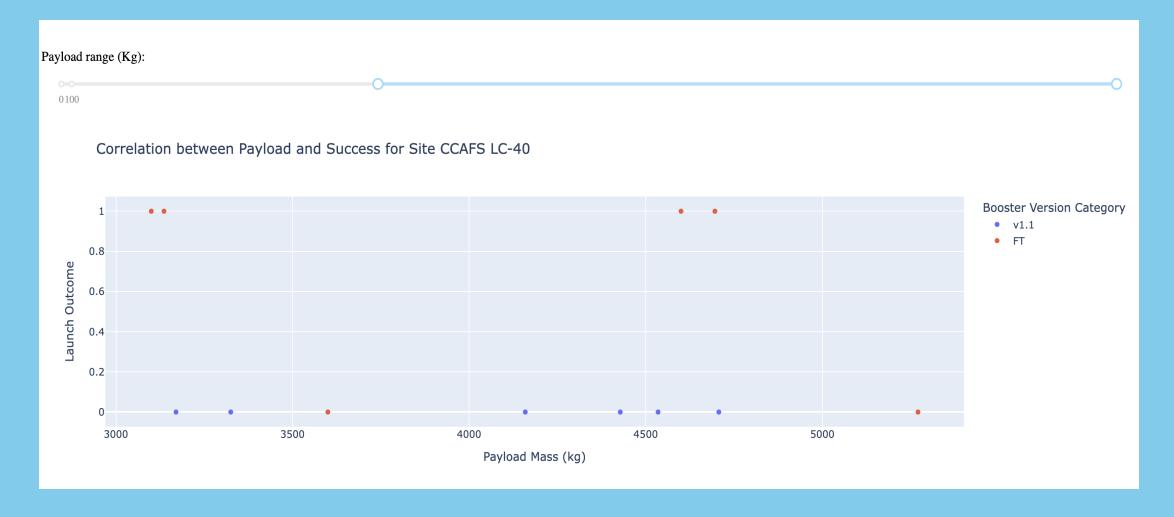


Using python dash, we can see the percentage of the total success lauches by each site

SpaceX Launch Records Dashboard



And we can pick a particular launch site to see the successful rate

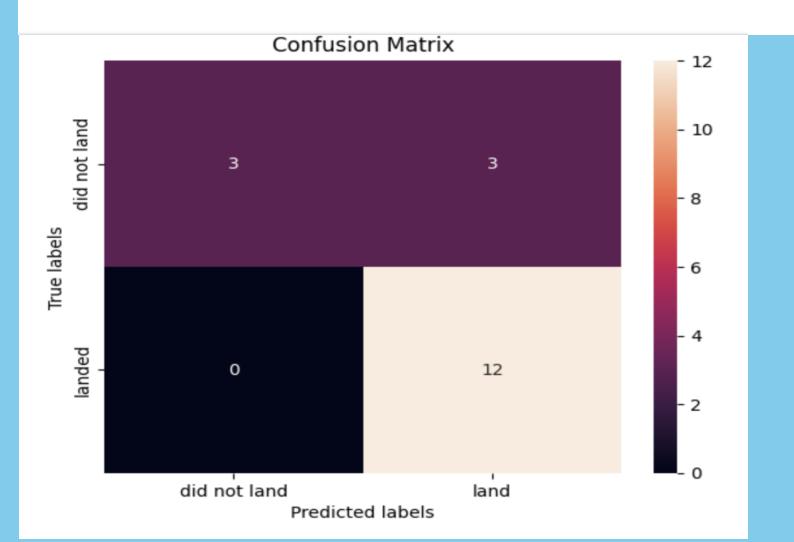


We can pick different Payload Mass an visualize the success rate

Classification ML using Logistic Regression, SVM, Decision Tree and KNN

```
tuned hpyerparameters :(best parameters) {'C': 0.01, 'penalty': 'l2', 'solver': 'lbfgs'}
accuracy: 0.8464285714285713
tuned hpyerparameters :(best parameters) {'C': 1.0, 'gamma': 0.03162277660168379, 'kernel': 'sigmoid'}
accuracy: 0.8482142857142856
tuned hpyerparameters :(best parameters) {'criterion': 'gini', 'max_depth': 4, 'max_features': 'sqrt', 'min_sampl
es_leaf': 1, 'min_samples_split': 2, 'splitter': 'random'}
accuracy: 0.875
GridSearchCV(cv=10, estimator=KNeighborsClassifier(),
               param_grid={'algorithm': ['auto', 'ball_tree', 'kd_tree', 'brute'],
                              'n_neighbors': [1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
                              'p': [1, 2]})
```

test accuracy: 0.83333333333333333



The accuracy rate in training set of different classification method is relatively close to each other and the test accuracy and confusion matrix is the same for all 4 methods. So we cannot tell which method is better for this classification problem. We may need to use a larger set of data so we can have a larger test data set.

Summary: With high Flight number, the landing is more likely to success. And the successful rate is increasing since 2013 to 2020. (Our data ends at 2020). The orbit will higher successful rate usually combines with high Flight number and low payload.