

Winning Space Race with Data Science

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Outline

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



Executive Summary

Summary of Methodologies

- Data collection: I made GET requests to retrieve historical Falcon 9 launch data through the SpaceX API.
- Web scraping: Used BeautifulSoup to extract an HTML table of historical Falcon 9 launch records from Wikipedia for further analysis.
- Data Wranling: EDA to find some patterns in the data and determine what would be the label for training supervised models.
- EDA: using Pandas and Matplotlib for visualizations and data preparation.
- EDA: Connected to a Db2 database to load the SpaceX dataset and ran SQL queries to answer key questions about the launches.
- Created interactive visual analyses using Folium to explore geospatial data and launch locations.
- Built an interactive Dash application using Plotly, enabling real-time visualization of SpaceX launch data.
- Machine learning: Trained various models including logistic regression, SVM, decision trees, and KNN, evaluating performance to select the best model using test data.

Summary of all results:

- The visualizations with Folium show that the best launch sites are those close to the coast and the equator, in order to avoid urban areas and make the most of the Earth's rotational speed.
- The site with the most successful launches is CCAFS SLC-40, with a success rate of 42.9%, but the site with the highest percentage of successful launches is KSC LC-39A.
- The payload range with the highest number of successful launches is between 2000-5000kg.
- The Booster FT version appears to have the highest rate of successful launches.
- The models Logistic Regression, SVM, Decision Trees, and KNN seem to have similar accuracy, at 0.83, but when we create the GridSearchCV() object to search for the best parameters, the classifier decision tree gets a accuracy=0.875, while the other models have a accuracy=0.848.
- Orbits with 100% success rates are HEO, SSO, ES-L1 and GEO, but the one with the highest percentage of successful launches is SSO.
- Except GTO, all orbits improve your successful launch rate with flight number.

Introduction

Project background and context

Space Exploration Technologies Corp., known as SpaceX, is an American aerospace manufacturing and space transportation services company headquartered in Hawthorne, California. It was founded in 2002 by Elon Musk with the aim of reducing the costs of traveling to space to facilitate the colonization of Mars. SpaceX has developed several launch vehicles, the Starlink constellation, the Dragon cargo ship and carried astronauts to the International Space Station on Dragon 2. SpaceX's achievements include the first privately financed liquid-fuel rocket to reach orbit (Falcon 1 in 2008),7 the first private company to launch and recover a spacecraft (Dragon in 2010), the first private in sending a ship to the International Space Station (Dragon in 2012),8 the first powered landing of an orbital rocket (Falcon 9 in 2015), the first reuse of a orbital rocket (Falcon 9 in 2017),

One of the reasons SpaceX is able to reach these milestones is because rocket launches are relatively affordable. SpaceX lists Falcon 9 rocket launches on its website as costing \$62 million, while other providers charge more than \$165 million per launch. A significant portion of the savings comes from SpaceX's ability to reuse the first stage. Therefore, if we can determine whether the first stage will land, we can estimate the cost of a launch. The second stage helps place the payload into orbit, but most of the work is done by the first stage. This stage is quite large and expensive. Unlike other rocket suppliers, SpaceX's Falcon 9 is capable of recovering the first stage. But sometimes the first stage doesn't land, other times it fails, and other times SpaceX sacrifices the first stage due to mission parameters like payload, orbit, and customer. In this final phase, I will take on the role of a data scientist working for a new rocket company, Space Y, that aims to compete with SpaceX. My job will be to determine the price of each release. To do this, I will collect information about SpaceX and create dashboards for my team. I will also determine if falcon 9 will reuse the first stage. To achieve this, I will train a machine learning model and use publicly available information to predict whether falcon 9 will reuse the first stage.

Problems I want to find answers

- To know if the payload affects the launch success rate.
- To know if the chosen orbit affects the launch success rate.
- To know if the version of the Booster affects the launch success rate.
 - To know if the launch site affects the launch success rate.
 - To know if the number of flights affects the launch success rate.
- And all of this in order to predict whether Falcon 9 will reuse the first stage and to predict the launch cost.





Methodology

Executive Summary

- Data collection methodology:
 - Describe how data was collected
- Perform data wrangling:
 - Describe how data was processed
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - How to build, tune, evaluate classification models

Data Collection – SpaceX API

Import Libraries and Define Auxiliary Functions



Request and parse the SpaceX launch data using the GET request



We create a dictionary with the variables and then convert it into a Dataframe



Use the json_normalize method to convert the json result to a data frame.



Export data to csv.

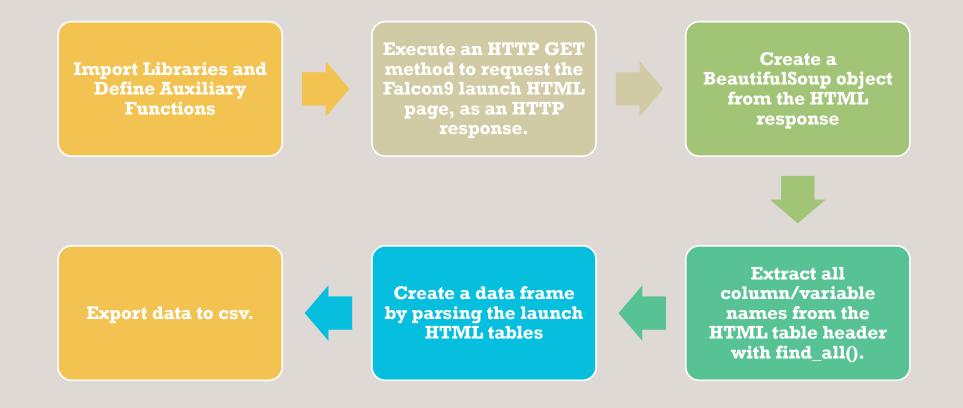


Find missing
values and replace
them with the column
mean using mean()
and replace().



Filter the dataframe to only include Falcon 9 launches

Data Collection - Scraping



Data Wrangling

Calculate the Calculate the number and **Import Libraries and** number of launches occurrence of each load data. on each site with orbit with value_counts(). value_counts().

Export data to csv.

Calculate the number and occurence of mission outcome of the orbits with value counts().

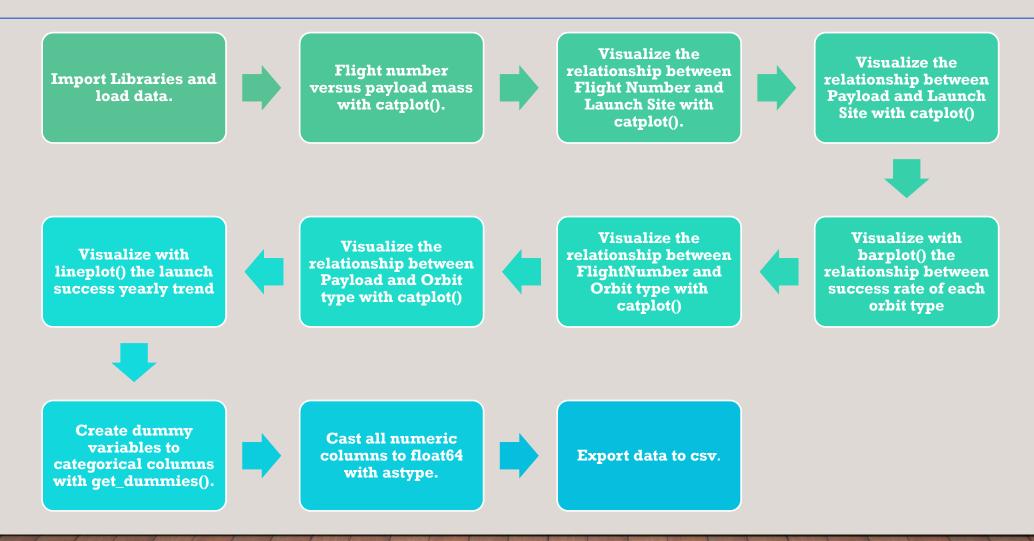


Use mean() on the new column to determine the landing success rate.



Create a landing result label from the Result column, to split landings into successful (1) and unsuccessful (0)

EDA with Data Visualization



EDA with SQL

- **SELECT DISTINCT, FROM:** display the names of the unique launch sites in the space mission.
- **SELECT, FROM, WHERE**: display 5 records where launch sites begin with the string 'CCA'.
- **SELECT, SUM(), FROM, WHERE**: Display the total payload mass carried by boosters launched by NASA (CRS).
- **SELECT, AVG(), FROM, WHERE:** Display average payload mass carried by booster version F9 v1.1.
- **SELECT, MIN(), FROM, WHERE:** List the date when the first successful landing outcome in ground pad was acheived.
- **SELECT, FROM, WHERE, AND, <>:** List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000.
- **SELECT, COUNT(CASE...WHEN...THEN), FROM**: List the total number of successful and failure mission outcomes.
- **SELECT, FROM, WHERE + SUBQUERY:** List the names of the booster versions which have carried the maximum payload mass.
- **SELECT, CASE(WHEN, SUBSTR(), THEN), FROM, WHERE, AND**: 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.
- **SELECT, COUNT(), FROM, WHERE, BETWEEN, AND, IN, GROUPBY():** Rank the count of landing outcomes between the date 2010-06-04 and 2017-03-20, in descending order.

https://github.com/Mai-de-jerez/IBM_Applied_Data_Science_Capstone/blob/main/jupyter-labs-eda-sql-coursera_sqllite.ipynb

Build an Interactive Map with Folium

- Folium.map(): create a map.
- Folium.Circle(): create circular marks at launch sites.
- Folium.map.Marker():name the sites according to their coordinates on the map.
- Folium.Popup(): Create pop-up windows with text by clicking on the marks.
- **DivIcon():** put names next to the circle marks.
- MarkerCluster(): group the markers of the same area according to their class.
- MousePosition(): add an object to display coordinates in a corner of the map, depending on where the mouse pointer is positioned.
- Folium.Marker(): to add the distance from the launch site to another location.
- Folium.Polyline(): add a line from the launch site to a reference point.
- Add_child(): add the marks to the map in question.

https://github.com/Mai-de-jerez/IBM Applied Data Science Capstone/blob/main/lab-jupyter-launch-site-location-v2%20(1).ipynb

Build a Dashboard with Plotly Dash

• Add a Launch Site Drop-down Input Component.

We have four different launch sites, and we would like to first see which one has the largest success count. Then, we would like to select one specific site and check its detailed success rate (class=0 vs. class=1).

- Adds a callback function to render the success pie chart based on the site selected in the dropdown

 The general idea of this callback function is to get the selected launch site from the dropdown site and render a pie chart displaying the launch success counts.
- Add a Range Slider to Select Payload.

The goal is to know if the variable payload is correlated with the mission outcome. From the control panel point of view, we want to be able to easily select different payload ranges and see if we can identify some visual patterns.

• Add a callback function to render the success-payload-scatter-chart scatter plot.

The goal is to plot a scatterplot with the x-axis as the payload and the y-axis as the launch result (i.e., class column). As such, we can visually observe how the payload can correlate with mission results for selected sites. Additionally, we have color-labeled the version of each Booster at each scatter point so we can observe mission results with different drivers.

https://github.com/Mai-dejerez/IBM Applied Data Science Capstone/blob/main/spacex dash app.py

Predictive Analysis (Classification)

We create for the variable Y an array from the 'Class' column using to_numpy()



We assign our set of dummy variables to our variable X and standardize the data with StandardScaler()



We use train_test_split() to split the data into test and training data



We created four objects: one logistic regression, another SVM, a decision tree classifier and another KNN.



We calculate the average precision obtained with the parameters of each model using best_score_



We adjust the object to find the best parameters with best_params_



We train each model with the fit() method



We create the GridSearchCV() objects, with cv=10



We calculate the precision of each model with the score() method.



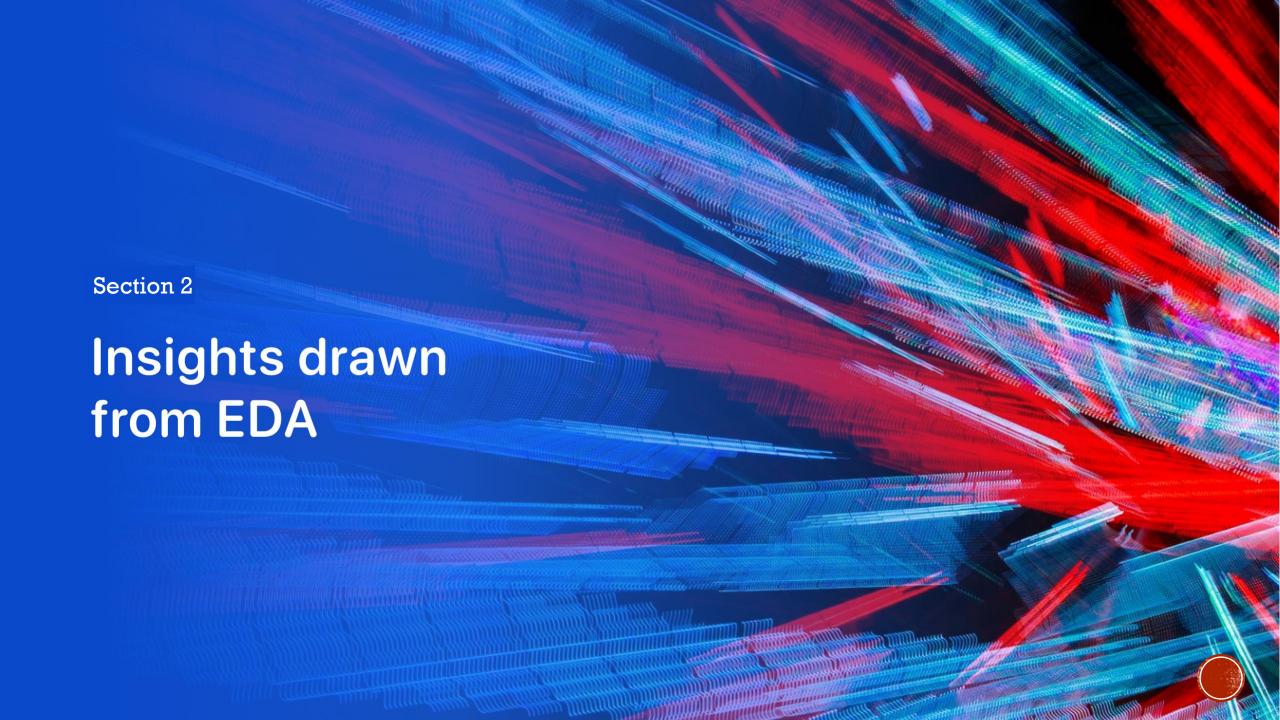
We generated the confusion matrix to evaluate the models.



We use best_score_ to know which model has the best performance

Results

- Exploratory data analysis results
 - When the flight number increases, the success rate of each type of orbit improves.
 - The orbit with the highest percentage of successful launches is SSO.
- Interactive analytics demo in screenshots:
 - The site with the highest percentage of successful launches is KSC LC-39A.
 - The payload mass with the highest rate of successful launches is between 2000 and 5000kg.
- Predictive analysis results
 - The best model is the decision tree with an accuracy of 87.5%.



Flight Number vs. Launch Site

```
# Plot a scatter point chart with x axis to be Flight Number and y axis to be the launch site, and hue to be the class value
    sns.catplot(y="LaunchSite", x="FlightNumber", hue="Class", data=df, aspect = 5)
    plt.xlabel("Flight Number", fontsize=20)
    plt.ylabel("Launch site", fontsize=20)
  plt.show()
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                        and the second of the second o
                                                                                                                          and an extra contract the contract of
Launch site
                                                                                                                                                                                                                                                                                                                                                                                                                                              After the fitter of the control of t
                          KSC LC 39A
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                Flight Number
```

I would say that the success rate improves with the number of flights for the three locations. But clearly CCAFS SLC 40 is the one that improves the most with the number of flights. In the other cases it is not so clear. CCAFS SLC-40 has a lower success rate.

Payload vs. Launch Site

We also want to observe if there is any relationship between launch sites and their payload mass.

```
# Plot a scatter point chart with x axis to be Pay Load Mass (kg) and y axis to be the launch site, and hue to be the class value
sns.catplot(y="LaunchSite", x="PayloadMass", hue="Class", data=df, aspect = 5)
plt.xlabel("Pay load Mass (kg)", fontsize=20)
plt.ylabel("Launch site", fontsize=20)
plt.show()

CCAMP SICC 40

WAPP SICC 40

RSC IC 300-

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Pay load Mass (kg)

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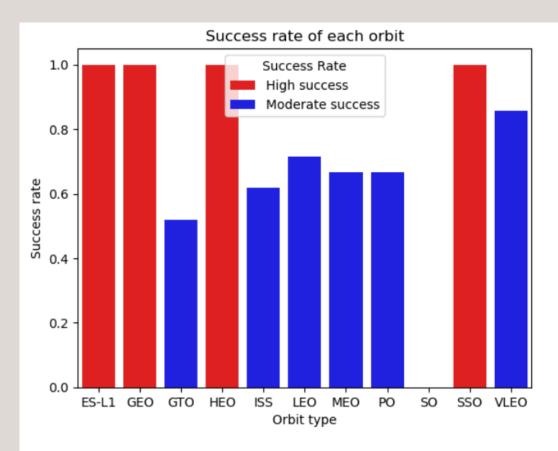
1000

1000

100
```

Now if you observe Payload Vs. Launch Site scatter point chart you will find for the VAFB-SLC launchsite there are no rockets launched for heavypayload mass(greater than 10000). CCAFS SLC-40 shows relatively good performance for lighter loads, with a tendency to have more successes as the program progresses. KSC LC-39A appears to be best suited for missions with large payloads, with a significant proportion of successes even at the upper end of payload. VAFB SLC-4E: This site has a smaller number of missions in the data set and an apparently higher proportion of failure

Success Rate vs. Orbit Type



Highly successful orbits: ES-L1, GEO, HEO, and SSO have impeccable performance. Most challenging orbits: GTO has the lowest success rate. LEO and VLEO: their success rate is moderate.

Flight Number vs. Orbit Type

```
# Plot a scatter point chart with x axis to be FlightNumber and y axis to be the Orbit, and hue to be the class value
sns.catplot(y="Orbit", x="FlightNumber", hue="Class", data=df, aspect = 5)
plt.xlabel("Flight Number", fontsize=20)
plt.ylabel("Orbit type", fontsize=20)
plt.show()
   LEO
Orbit type
  VLEO
                                                                         Flight Number
```

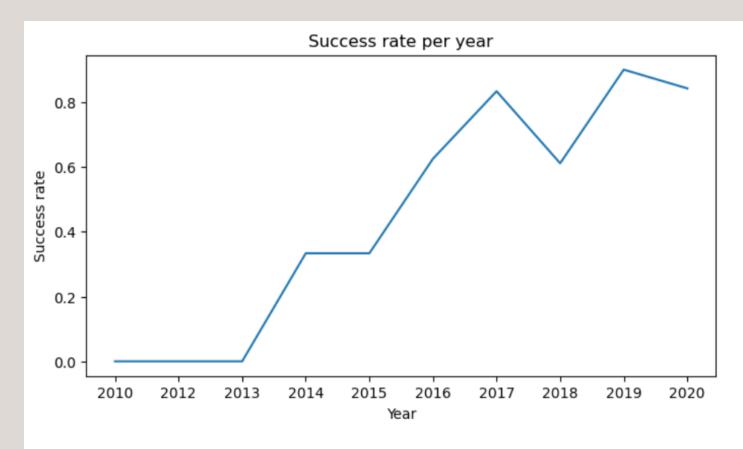
In the LEO orbit, Success appears related to the number of flights. In VLEO, ISS and PO success also seems to improve with the number of flights. However, there seems to be no relationship between the flight number when in GTO orbit, its success rate is also low. The orbits with the highest success rates are LEO, VLEO and especially SSO

Payload vs. Orbit Type

```
# Plot a scatter point chart with x axis to be Payload and y axis to be the Orbit, and hue to be the class value
sns.catplot(y="Orbit", x="PayloadMass", hue="Class", data=df, aspect = 5)
plt.xlabel("Pay load Mass (kg)",fontsize=20)
plt.ylabel("orbit type",fontsize=20)
plt.show()
   PO
  GTO -
  MEO
  VLEO -
   50
  GEO
                                                                        Pay load Mass (kg)
```

SSO and LEO work better with light loads, GTO is a complicated orbit, VLEO and ISS work better with heavier loads. The highest success rate is for SSO with loads between 0 and 4000kg.

Launch Success Yearly Trend



You can see that the success rate since 2013 kept increasing until 2017 (stable in 2014) and after 2015 it started increasing. There was a small decrease from 2017 to 2018.

All Launch Site Names

```
%sql SELECT DISTINCT Launch_Site from SPACEXTABLE
 * sqlite:///my_data1.db
Done.
 Launch_Site
 CCAFS LC-40
 VAFB SLC-4E
  KSC LC-39A
CCAFS SLC-40
```

Here we can see the name of the four launch locations.

Launch Site Names Begin with 'CCA'

%sql SELECT * FROM SPACEXTABLE where Launch Site like 'CCA%' LIMIT 5 * sqlite:///my data1.db Done. **Booster Version** Launch Site Payload PAYLOAD_MASS_KG_ Orbit Customer Mission Outcome Landing Outcome Date Dragon Spacecraft Qualification CCAFS LC-2010-18:45:00 F9 v1.0 B0003 0 LEO SpaceX Failure (parachute) Success 06-04 40 CCAFS LC-Dragon demo flight C1, two NASA (COTS) 2010-LEO 15:43:00 0 Failure (parachute) F9 v1.0 B0004 12-08 CubeSats, barrel of Brouere cheese (ISS) NRO CCAFS LC-2012-7:44:00 F9 v1.0 B0005 Dragon demo flight C2 525 NASA (COTS) No attempt Success (ISS) 05-22 2012-LEO CCAFS LC-0:35:00 500 NASA (CRS) F9 v1.0 B0006 SpaceX CRS-1 No attempt Success 10-08 (ISS) CCAFS LC-2013-LEO 15:10:00 F9 v1.0 B0007 677 NASA (CRS) Success SpaceX CRS-2 No attempt (ISS) 03-01

All five missions were successful, 4 of 5 customers were NASA, all payloads were light, and the orbit is always LEO.

Total Payload Mass

Total mass of payload carried by NASA-launched boosters (CRS): 45596

Average Payload Mass by F9 v1.1

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

%%sql

SELECT AVG(PAYLOAD_MASS__KG_) AS total_payload_mass
FROM SPACEXTABLE

WHERE Booster_Version='F9 v1.1'

* sqlite:///my_data1.db
Done.

total_payload_mass

2928.4
```

Average payload mass carried by the F9 v1.1 booster version: 2928.4

First Successful Ground Landing Date

Date on which the first successful landing on a terrestrial platform was achieved: 2015-12-22

Successful Drone Ship Landing with Payload between 4000 and 6000

```
%%sql
SELECT Booster Version
FROM SPACEXTABLE
WHERE Landing Outcome = 'Success (drone ship)'
AND PAYLOAD MASS KG > 4000
AND PAYLOAD MASS KG < 6000
 * sqlite:///my data1.db
Done.
Booster_Version
    F9 FT B1022
   F9 FT B1026
  F9 FT B1021.2
  F9 FT B1031.2
```

Names of thrusters that are successful in unmanned spacecraft and have a charge mass greater than 4000 but less than 6000:

F9 FT B1022 F9 FT B1026 F9 FT B1021.2 F9 FT B1031.2

Total Number of Successful and Failure Mission Outcomes

```
%%sql
SELECT
    COUNT(CASE WHEN Mission_Outcome LIKE '%Success%' THEN 1 END) AS Total_Success,
    COUNT(CASE WHEN Mission_Outcome LIKE '%Failure%' THEN 1 END) AS Total_Failure
FROM SPACEXTABLE
    * sqlite:///my_data1.db
Done.

Total_Success Total_Failure

100 1
```

Total number of successful mission results: 100

Total number of failure mission results: 1

Boosters Carried Maximum Payload

```
%%sql
SELECT booster_version
FROM SPACEXTABLE
WHERE PAYLOAD MASS KG = (
    SELECT MAX(PAYLOAD MASS KG )
    FROM SPACEXTABLE
* sqlite:///my_data1.db
Done.
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
```

Names of booster versions that have carried the maximum payload mass: 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

2015 Launch Records

```
%%sql
SELECT
    CASE
        WHEN substr(Date, 6, 2) = '01' THEN 'Enero'
        WHEN substr(Date, 6, 2) = '02' THEN 'Febrero'
        WHEN substr(Date, 6, 2) = '03' THEN 'Marzo'
        WHEN substr(Date, 6, 2) = '04' THEN 'Abril'
        WHEN substr(Date, 6, 2) = '05' THEN 'Mayo'
        WHEN substr(Date, 6, 2) = '06' THEN 'Junio'
        WHEN substr(Date, 6, 2) = '07' THEN 'Julio'
        WHEN substr(Date, 6, 2) = '08' THEN 'Agosto'
        WHEN substr(Date, 6, 2) = '09' THEN 'Septiembre'
        WHEN substr(Date, 6, 2) = '10' THEN 'Octubre'
        WHEN substr(Date, 6, 2) = '11' THEN 'Noviembre'
        WHEN substr(Date, 6, 2) = '12' THEN 'Diciembre'
    END AS Month,
    Booster Version,
    Launch Site,
    Landing Outcome
FROM SPACEXTABLE
WHERE substr(Date, 0, 5) = '2015'
AND Landing Outcome = 'Failure (drone ship)'
* sqlite:///my data1.db
Done.
Month Booster Version Launch Site Landing Outcome
          F9 v1.1 B1012 CCAFS LC-40 Failure (drone ship)
 Enero
          F9 v1.1 B1015 CCAFS LC-40 Failure (drone ship)
  Abril
```

Names of the months, results of unsuccessful landing on the unmanned ship, booster versions and launch site for the months of the year 2015:

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

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

```
%%sql
SELECT COUNT(landing_outcome) AS total, Landing_Outcome
FROM SPACEXTABLE
WHERE Date BETWEEN '2010-06-04' AND '2017-03-20'
and Landing Outcome IN ('Failure (drone ship)', 'Success (ground pad)')
GROUP BY Landing Outcome
 * sqlite:///my data1.db
Done.
       Landing_Outcome
total
       Failure (drone ship)
   3 Success (ground pad)
```

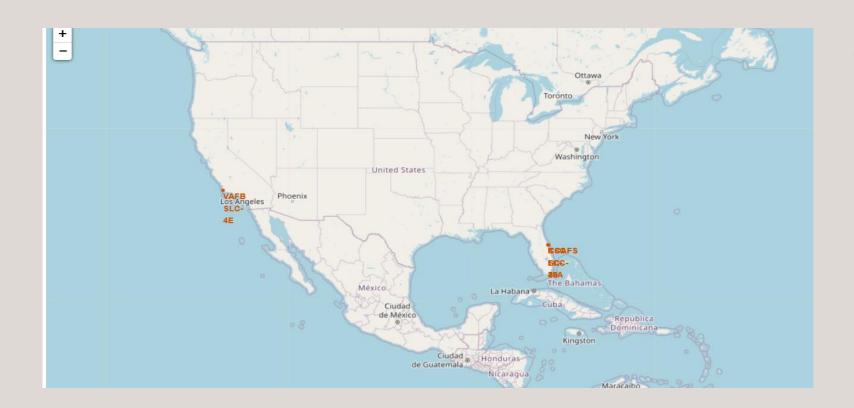
Count of landing results, between the date 2010-06-04 and 2017-03-20, in descending order:

Failure (drone ship):

5
Success (ground pad):
3



Launch sites



All launch sites are close to the equator to take full advantage of the Earth's considerable rotation speed. The launch sites are near the coast because if during the rocket's ascent trajectory, the rocket explodes and falls back to earth, it is better for it to fall in an unpopulated area, and that is easier near the sea.

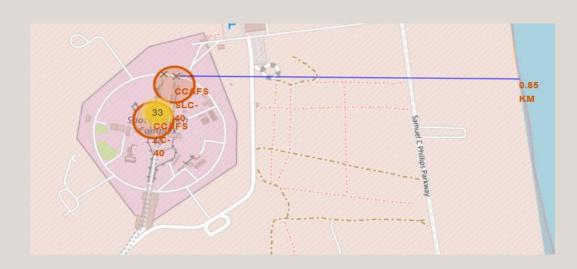
Number of launches per site and success rate



If we look at the first map, we can see the number of pitches grouped by zone.

If we look at the second map, we can see that by clicking on the mark, we are shown the result of each launch, green for the successful ones and red for the unsuccessful ones. And we can also easily see that KSC LC 39-A is the site with the most successful launches.

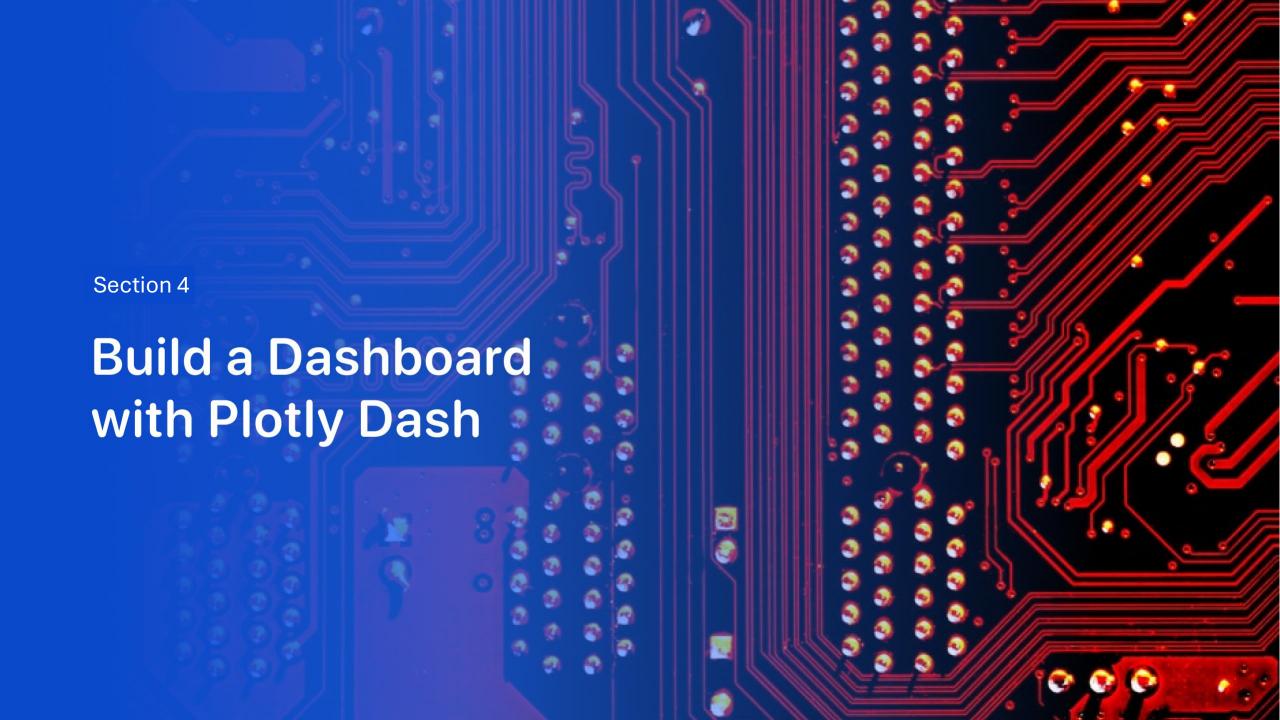
Distance between the launch site and a reference point



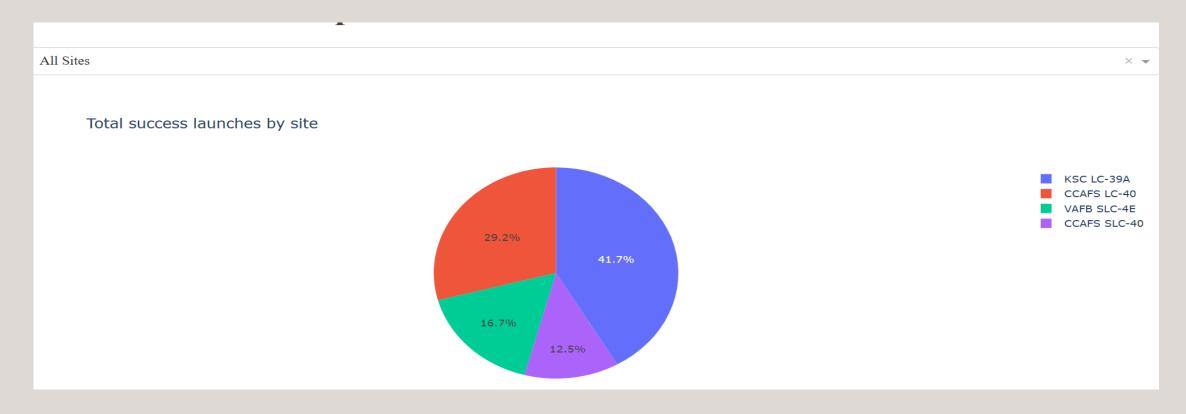
The launch site is close to the coast (0.85 km), but is far from cities, railways and highways, maintaining a safe distance of more than 20 km.





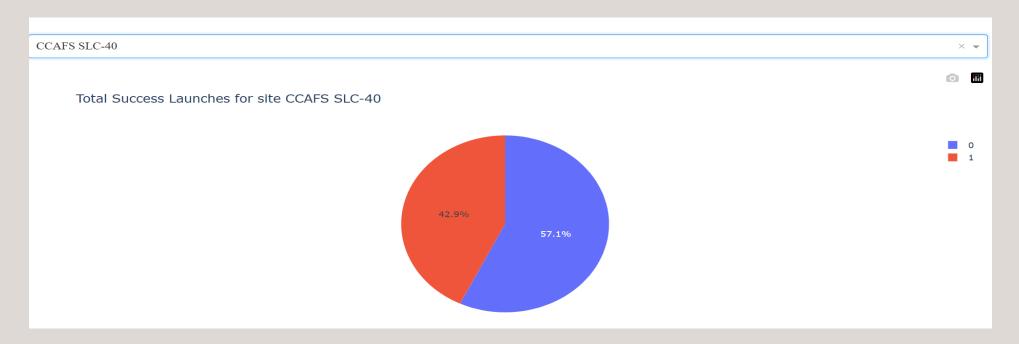


Launch success count for all sites



We can clearly see that the site with the highest number of launches is KSC LC-39A

Launch site with highest launch success ratio



We can see that the percentage 42.9% corresponds to the color red, the red color represents class 1, which means successful launch.

The other launch sites have a lower percentage of 42.9%.

Relationship between the choice of payload and Booster version with mission success

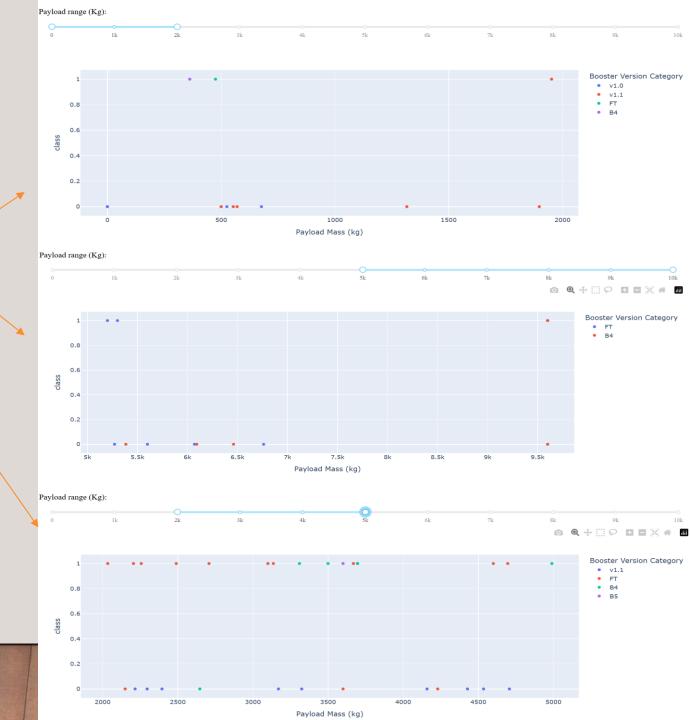
Payload ranges with the lowest success rate:

0-2000kg 5000-10000kg.

Payload range with highest launch success rate:

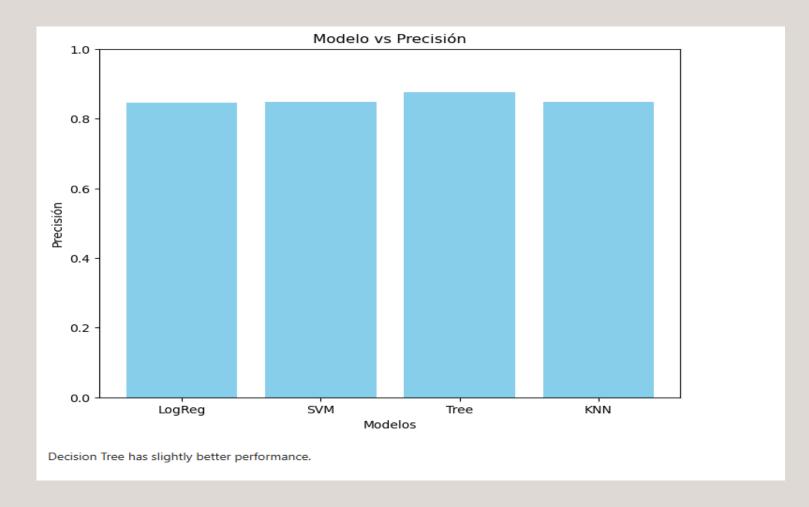
2000-5000kg.

Booster version with the highest successful launch rate:
FT



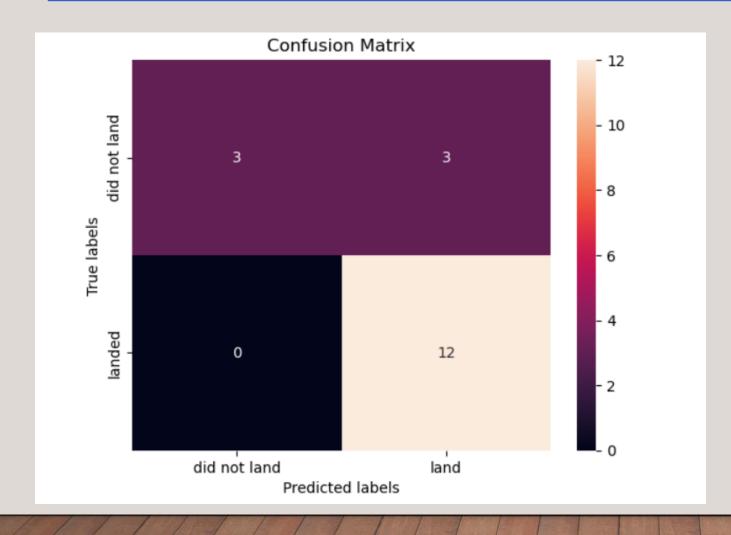


Classification Accuracy



Decision Tree Classifier has slightly better performance.

Confusion Matrix



The confusion matrix is similar for all models, with a precision of 83.3%.

But when we create the GridSearchCV() object to search for the best parameters, the classifier decision tree gets a accuracy=0.875, while the other models have a accuracy=0.848.

Conclusions

- Launch site: the site with the highest rate of successful launches is KSC LC-39A
- Orbit: SSO is the orbit with the highest percentage of successful launches, and has a 100% success rate.
- Payload mass(kg): between 2000 and 5000 kg the rate of successful launches is higher.
- Flight number: as the flight number increases, the number of successful launches usually improves.
- Yearly trend: The successful launch rate for falcon 9 has improved over the years.
- Booster version: FT is the version of the booster with the highest rate of successful launches.
- Launch site location: the best launch sites are those close to the coast and the equator.
- Model: The best model is the classifier decision tree, with an accuracy of 87.5%.

Appendix

- Resources:
 - https://api.spacexdata.com
 - https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-DS0321EN-SkillsNetwork/datasets/API call spacex api.json

• https://en.wikipedia.org/w/index.php?title=List_of-Falcon_9 and Falcon Heavy launche_s&oldid=1027686922

[hide] Flight No.	Date and time (UTC)	Version, Booster ^[b]	Launch site	Payload ^[c]	Payload mass	Orbit	Customer	Launch outcome	Booster landing
78	7 January 2020, 02:19:21 ^[492]	F9 B5 △ B1049.4	CCAFS, SLC-40	Starlink 2 v1.0 (60 satellites)	15,600 kg (34,400 lb) ^[5]	LEO	SpaceX	Success	Success (drone ship
	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. (403)								
79	19 January 2020, 15:30 ^[494]	F9 B5 △ B1046.4	KSC, LC-39A	Crew Dragon in-flight abort test ^[495] (Dragon C205.1)	12,050 kg (26,570 lb)	Sub-orbital ^[496]	NASA (CTS) ^[497]	Success	No attemp
	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 salted to be accomplished with the Crew Dragon Engine Super But that test article exploded during a ground test of SuperDraco engines on 20 April 2019. The abort test use the capsule originally intended for the 1 crewed flight. 1997 As expected, the booster was destroyed by aerodynatic roces after the capsule aborted. A Falson 9 with only one functional stage — the second stage had a mass similarity in place of its engine.								
80	29 January 2020, 14:07 ^[501]	F9 B5 △ B1051.3	CCAFS, SLC-40	Starlink 3 v1.0 (60 satellites)	15,600 kg (34,400 lb) ^[5]	LEO	SpaceX	Success	Success (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. [602]								
81	17 February 2020, 15:05 ^[503]	F9 B5 △ B1056.4	CCAFS, SLC-40	Starlink 4 v1.0 (60 satellites)	15,600 kg (34,400 lb) ^[5]	LEO	SpaceX	Success	Failure (drone shi
	Fourth operational and fifth large batch of Starlink satellities. 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. [605] This was the first time a flight proven booster failed to land.								
82	7 March 2020, 04:50 ^[506]	F9 B5 🛆 B1059.2	CCAFS, SLC-40	SpaceX CRS-20 (Dragon C112.3 △)	1,977 kg (4,359 lb) ^[507]	LEO (ISS)	NASA (CRS)	Success	Succes (ground pa
	Last launch of phase 1 of the CRS contract. Carries Bartolomeo, an ESA platform for hosting external payloads onto ISS [1008] Originally scheduled to launch on 2 March 2020, the launch date was pushed back due to a second stage engine failure. Space decided to swap out the second stage instead of replacing the faulty part [1009] 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.								
	18 March 2020, 12:16 ^[510]	F9 B5 △ B1048.5	KSC, LC-39A	Starlink 5 v1.0 (60 satellites)	15,600 kg (34,400 lb) ^[5]	LEO	SpaceX	Success	Failure (drone sh
83	Fifth operational launch of Starfink 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 Meriin 10 variant and first since the CRS-1 mission in October 2012. However, the payload still reached the targeted orbst, [512] This was the second Starlink launch booster landing failure in a row, later revealed to be caused by residuel cleaning fluid trapped inside a sensor. [513]								

