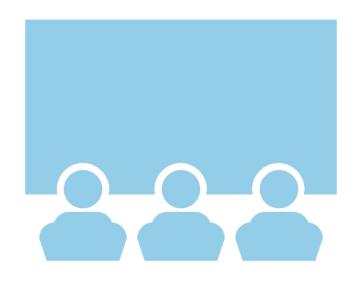
Data Science Capstone project

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<27/08/2021>

Outline



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

Executive Summary

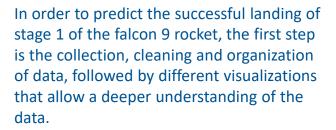
This project consists of the prediction of the successful landing of stage 1 on the Falcon 9 rocket.



One of the main reasons why the launch costs of the Space X company are lower than the rest is due to the correct landing of stage 1 of its rockets. This strategy allows a cost reduction of up to 100 million dollars.



That is why it is of the utmost importance for the company to be able to determine whether the landing of phase one of the rockets will end in success or failure.







Finally, a machine learning model is developed capable of performing the task of predicting the success or otherwise of the landing of stage 1 of the rockets.





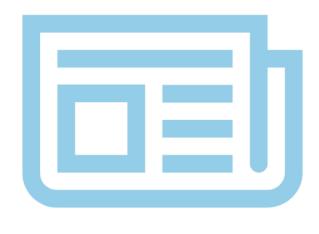


Introduction



- The costs of rocket launches are extremely high in part due to the loss of the stage 1 of the rockets.
- The costs of the spaceX Falcon 9 rocket launches are considerably lower than the rest because they can recover stage 1 of the rockets.
- In order to keep the costs of rocket launches low, it is important to be able to predict the success of the Falcon 9 stage 1 rocket landing.

Methodology



- Perform data wrangling
 - Describe how data were 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

Methodology

Data collection

The information used for this project corresponds to the data obtained from the SpaceX API together with the Wikipedia data on the falcon 9 launches collected using the WebScrapping method.



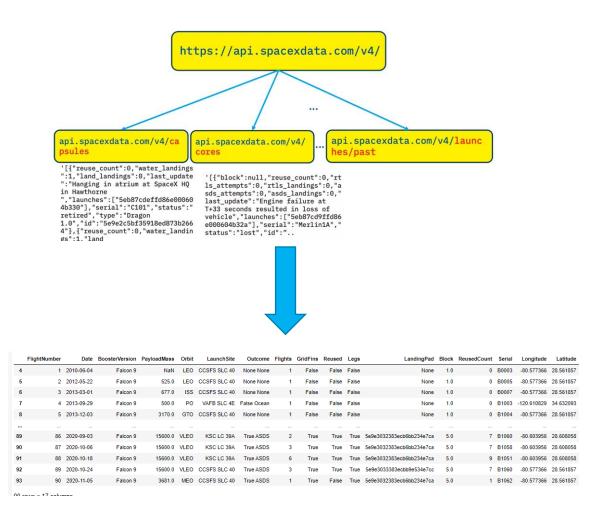
Data collection – SpaceX API

Information and details on Falcon 9 rocket launches are obtained from the SpaceX API.

As seen in the figure, this api contains multiple endpoints that provide different information about Falcon 9 launches, using the beautifulsoup and python requests libraries it is possible to extract this data and structure it for later analysis.

Finally, the data obtained from the API corresponds to:

booster name, the mass of the payload, the orbit that it is going to, the name of the launch site and its position, the outcome of the landing, the type of the landing, number of flights and the core information.



https://github.com/joseMCV/space-y/blob/e289e099c77fd5a5aba82732517da419df227843/space_y_api.ipynb

Data collection – Web scraping

To extract the data from wikipedia, first the raw page must be read, for this the python requests library is used, once the page has been read, BeautifulSoup is used to find the specific information that is required from this page.

requested corresponds to the following:
Flight number, date and time, place of
launch, payload, orbit to which the launch is
directed, client and final status of the
operation (success or failure)





	Flight No.	Launch site	Payload	Payload mass	Orbit	Customer	Launch outcome	Version Booster	Booster landing	Date	Time
0	1	CCAFS	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success\n	F9 v1.0B0003.1	Failure	4 June 2010	18:45
1	2	CCAFS	Dragon	0	LEO	NASA	Success	F9 v1.0B0004.1	Failure	8 December 2010	15:43
2	3	CCAFS	Dragon	525 kg	LEO	NASA	Success	F9 v1.0B0005.1	No attempt\n	22 May 2012	07:44
3	4	CCAFS	SpaceX CRS-1	4,700 kg	LEO	NASA	Success\n	F9 v1.0B0006.1	No attempt	8 October 2012	00:35
4	5	CCAFS	SpaceX CRS-2	4,877 kg	LEO	NASA	Success\n	F9 v1.0B0007.1	No attempt\n	1 March 2013	15:10

116	117	CCSFS	Starlink	15,600 kg	LEO	SpaceX	Success\n	F9 B5B1051.10	Success	9 May 2021	06:42
117	118	KSC	Starlink	~14,000 kg	LEO	SpaceX	Success\n	F9 B5B1058.8	Success	15 May 2021	22:56
118	119	CCSFS	Starlink	15,600 kg	LEO	SpaceX	Success\n	F9 B5B1063.2	Success	26 May 2021	18:59
119	120	KSC	SpaceX CRS-22	3,328 kg	LEO	NASA	Success\n	F9 B5B1067.1	Success	3 June 2021	17:29
120	121	CCSFS	SXM-8	7,000 kg	GTO	Sirius XM	Success\n	F9 B5	Success	6 June 2021	04:26

https://github.com/joseMCV/space-y/blob/e289e099c77fd5a5aba82732517da419df227843/space y webscrapping.ipynb

Data wrangling

- To make the prediction about the success or failure of the Falcon 9 rocket stage 1 landing, it is important to label the data, that is, to be able to determine if the rocket stage 1 landed successfully or not.
- In this stage, different columns of the data were analyzed to understand the relevance of each one with respect to the final objective of the mission, that is, how each column affects the success in the landing of stage 1.

EDA with data visualization

- The first visualizations generated correspond to scatter graphics, this is to visualize the relationships between the different variables, the relationships plotted were:
 - FlightNumber vs. PayloadMass.
 - FlightNumber vs LaunchSite.
 - LaunchSites vd PayloadMass.
 - FlightNumber and Orbit type.
 - Payload vs. Orbit type.
- It should be noted that all the visualizations are contrasted with the success or not of the landing, so thanks to these it can be determined whether the graphed variables influence the success of the launch.

EDA with data visualization

• After making the scatter graphs, a line graph was made to determine the behavior of the success of the launches over time, from this graph it appears that since 2013, the success in the landings of stage 1 of the Falcon 9 have been increasing.

EDA with SQL

To obtain relevant information on the data, various SQL queries listed below were performed:

- SELECT DISTINCT LAUNCH SITE FROM SPACEXTBL
- SELECT * FROM SPACEXTBL WHERE LAUNCH SITE LIKE 'CCA%' LIMIT 5
- SELECT customer, SUM(payload_mass__kg_) as total_payload FROM SPACEXTBL WHERE customer = 'NASA (CRS)' GROUP BY customer
- SELECT BOOSTER VERSION, AVG(payload mass kg) as avg_payload FROM SPACEXTBE WHERE BOOSTER_VERSION = 'F9 v1.1' GROUP BY BOOSTER VERSION
- SELECT MIN(DATE) FROM SPACEXTBL WHERE MISSION OUTCOME LIKE '%Succ%'
- SELECT DISTINCT BOOSTER_VERSION FROM SPACEXTBL WHERE MISSION_OUTCOME LIKE '%Succ%' AND payload_mass_kg_>6000
- SELECT MISSION OUTCOME, COUNT(*) FROM SPACEXTBE GROUP BY MISSION_OUTCOME
- SELECT DISTINCT BOOSTER_VERSION
 FROM SPACEXTBL
 WHERE payload mass kg = (SELECT MAX(payload mass kg) FROM SPACEXTBL)
- SELECT DATE, BOOSTER_VERSION, LAUNCH_SITE FROM SPACEXTBL WHERE LANDING OUTCOME = 'Failure (drone ship)' AND YEAR(DATE) = 2015
- SELECT LANDING OUTCOME, COUNT(*) AS outcome_counts FROM SPACEXTBL
 WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20' GROUP BY LANDING OUTCOME ORDER BY count(*) DESC

Build an interactive map with Folium

• To better understand the location of the launches, the visualizations of the python folium library are used, in which markers are used to understand where the launches occurred, these in turn are decorated with an icon that indicates whether or not that launch was successful. It is important to note that the markers were clustered so that the visualization was not overloaded, finally, lines were added to calculate the distances between the launches and the nearest cities.

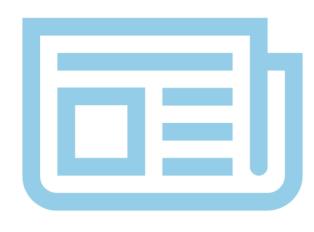
Build a Dashboard with Plotly Dash

- For the designed dashboard, two interactions and two different graphics were implemented, the interactions consisted of selecting a launch site through a dropdown and selecting the payload at launch through a range slider.
- The graphs designed were linked to the previously mentioned interactions, the pie graph when all the launch sites were selected, showed the total of successful launches for each launch site, however, if a specific launch site was selected, the graph it showed the number of times it had successful and failed landings. On the other hand, the Scatter chart was linked to both interactions where both were used as filters for the chart.
- These visualizations were added in order to improve the understanding of the data and the relationships that showed a greater correlation with the output of each landing of each launch.

Predictive analysis (Classification)

- The sklearn library was used to create 4 different models that would allow to perform the designated classification task, that is, to predict whether the Falcon 9 rocket phase one landing status.
- The four selected models were KNN, Logistic Regression, SVM and decision tree, for these four models a gridsearch method was used to find the hyperparameters that could deliver the best results over the training set, 10 iterations of cross validation were used to find the best parameters.
- Finally, the models were tested on the test set to determine their performance on data they had never "seen" before.
- The models used present an identical behavior with respect to the general precision, that is, the ability to predict whether stage 1 of the rocket will successfully land or not.

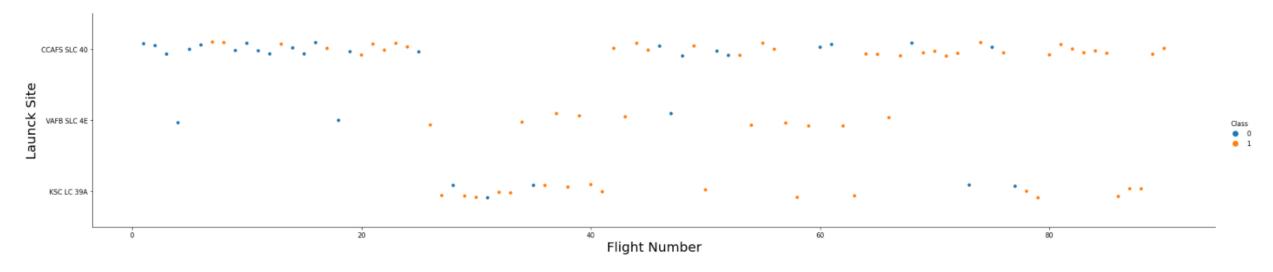
Results



- Exploratory data analysis results
- Interactive analytics demo in screenshots
- Predictive analysis results

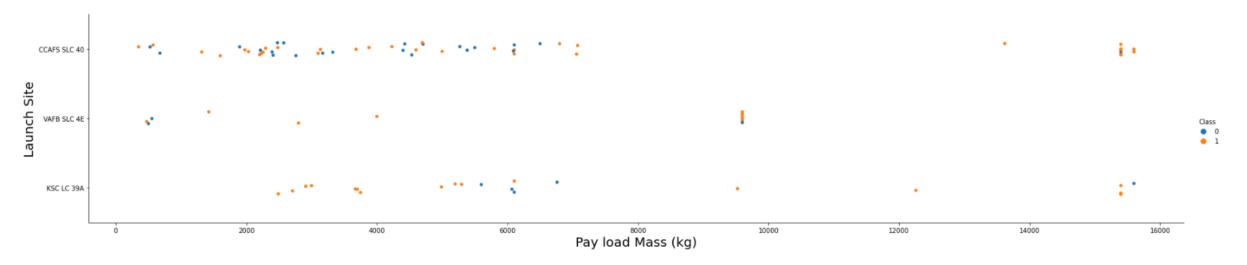
EDA with Visualization

Flight Number vs. Launch Site



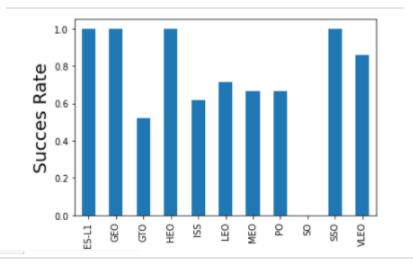
From the illustrated graph it can be seen that as the number of flights increases, so does the number of flights for each of the launch sites, the greater the probability that the stage 1 of the rocket achieves a successful landing.

Payload vs. Launch Site



From the graph it can be seen that when the payload mass is greater than 8000, most of the flights end in a success, in addition to the KSC LC 39A launch site all launches with a payload mass less than 4000 end in success. CCAFS SLC 40 launches with a payload mass less than 800 have the most diverse behavior within the chart.

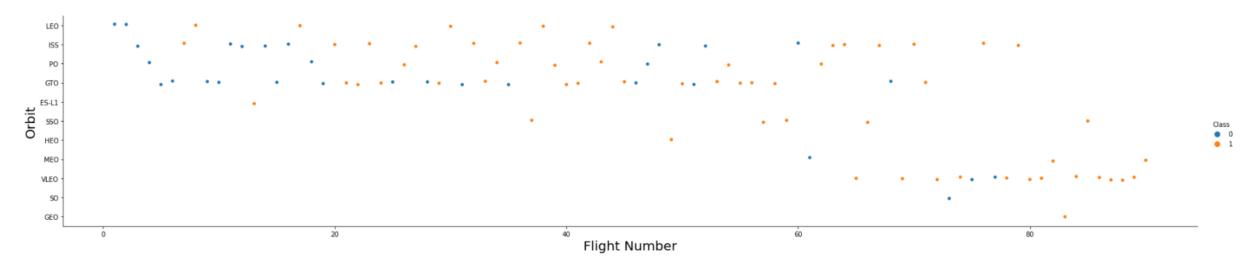
Success rate vs. Orbit type



	Success Rate	count
Orbit		
ES-L1	1.000000	1
GEO	1.000000	1
HEO	1.000000	1
SSO	1.000000	5
VLEO	0.857143	14
LEO	0.714286	7
MEO	0.666667	3
PO	0.666667	9
ISS	0.619048	21
GTO	0.518519	27
SO	0.000000	1

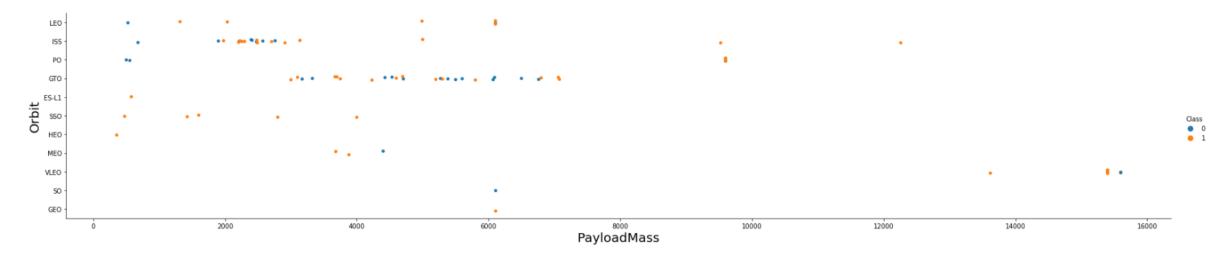
At first glance, one might think that the orbits with the highest success rate are ES-L1, GEO and HEO, however, only one launch has been made to each of these, which does not give us the necessary information to verify this statement. On the other hand, SSO and VLEO have success rates higher than 85% already having a considerable number of flights, so they could be identified as the orbits with the highest success rate

Flight Number vs. Orbit type



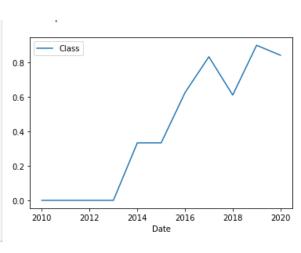
Nothing can be concluded from this graph at first glance, beyond the aforementioned relationship between the increase in the number of flights being proportional to the success rate.

Payload vs. Orbit type



For the LEO, ISS, PO orbits, the higher the payload, the higher the success rate, for the SSO orbit, all attempts have been successful, as has the ES-L1 orbit, so a relationship cannot be appreciated. with the payloadmass.

Launch success yearly trend



It can be seen that since 2013, the number of successful landings has increased, this is related to the experience and learnings obtained from previous missions, this supports the theory that the greater the number of flights, the greater the probability of success.

EDA with SQL

All launch site names

• The "DISTINCT" function is used to find the unique values in the requested column

%sql SELECT DISTINCT LAUNCH_SITE FROM SPACEXTBL

* ibm_db_sa://wfd01464:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/bludb Done.

launch site

CCAFS LC-40

CCAFS SLC-40

KSC LC-39A

VAFB SLC-4E

Launch site names begin with `CCA`

• The combination of the "WHERE" and "LIKE" functions is used to make sure we obtain the records of the release sites that begin with CCA, also the "LIMIT" function is used to only obtain 5 records.

Out[14]:

: [DATE	timeutc_	booster_version	launch_site	payload	payload_masskg_	orbit	customer	mission_outcome	landing_outcome
- 1	2010- 06-04	18:45:00	F9 v1.0 B0003	CCAFS LC- 40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
	2010- 12-08	15:43:00	F9 v1.0 B0004	CCAFS LC- 40	Dragon demo flight C1, two CubeSats, barrel of Brouere cheese	0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
- 1	2012- 05-22	07:44:00	F9 v1.0 B0005	CCAFS LC- 40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
	2012- 10-08	00:35:00	F9 v1.0 B0006	CCAFS LC- 40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
- 1	2013- 03-01	15:10:00	F9 v1.0 B0007	CCAFS LC- 40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

Total payload mass

• The "SUM" function is used to obtain the sum of the payload, and the "WHERE" function is used to specify the customer. The "GROUP BY" function is used to make the SUM function work.

```
In [18]:  
%%sql SELECT customer, SUM(payload_mass_kg_) as total_payload  
FROM SPACEXTBL  
WHERE customer = 'NASA (CRS)'  
GROUP BY customer

* ibm_db_sa://wfd01464:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/bludb  
Done.

Out[18]:  
customer  
total_payload  
NASA (CRS)  
45596
```

Average payload mass by F9 v1.1

 The "AVG" function is used to obtain the average of the payload and the "WHERE" function is used to specify the customer, the "GROUP BY" function is used to make the "AVG" function work.

* ibm_db_sa://wfd01464:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.c1ogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/bludb Done.

Out[23]:

booster_version	avg_payload
F9 v1.1	2928

First successful ground landing date

2015-12-22

• The "MIN" function is used to find the minimum date, and the where function to fulfill the condition of successful landing.

Successful drone ship landing with payload between 4000 and 6000

The "DISTINCT" function is used to find the unique values in the requested column, along with this the "WHERE" function is used both to ensure the specified landing condition and to find the payload within the requested range, it should be noted that this is the same could be achieved using the "BETWEEN" function

```
In [16]: %%sql SELECT DISTINCT BOOSTER_VERSION
FROM SPACEXTBL
WHERE landing_outcome = 'Success (drone ship)'
AND payload_mass_kg_ >4000
AND payload_mass_kg_ <6000

* ibm_db_sa://wfd01464:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/bludb
Done.

Out[16]: booster_version
F9 FT B1021.2
F9 FT B1031.2
```

F9 FT B1022 F9 FT B1026

Total number of successful and failure mission outcomes

• The "COUNT" function is used to count the number of landings in each category (successful or failure)



Boosters carried maximum payload

• In this case, a sub-query must be used to find the maximum payload condition, then the "WHERE" function is used to find the BOOSTER_VERSIONS that meet the maximum payload condition.

Task 8

List the names of the booster versions which have carried the maximum payload mass. Use a subquery

```
In [36]: %%sql SELECT DISTINCT BOOSTER_VERSION
          FROM SPACEXTBL
         WHERE payload mass kg = (SELECT MAX(payload mass kg ) FROM SPACEXTBL)
             * ibm_db_sa://wfd01464:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/bludb
            Done.
             booster version
               F9 B5 B1048.4
               F9 B5 B1048.5
               F9 B5 B1049.4
               F9 B5 B1049.5
               F9 B5 B1049.7
               F9 B5 B1051.3
               F9 B5 B1051.4
               F9 B5 B1051.6
               F9 B5 B1056.4
               F9 B5 B1058.3
               F9 B5 B1060.2
               F9 B5 B1060.3
```

2015 launch records

 In this query the "WHERE" function is used to fulfill the landing_outcome condition, and the "WHERE" and "YEAR" function for the condition about the year 2015

Task 9

F9 v1.1 B1012 CCAFS LC-40

2015-04-14 F9 v1.1 B1015 CCAFS LC-40

List the failed landing outcomes in drone ship, their booster versions, and launch site names for the in year 2015

```
In [41]: %%sql SELECT DATE, BOOSTER_VERSION, LAUNCH_SITE
FROM SPACEXTBL
WHERE LANDING_OUTCOME = 'Failure (drone ship)'
AND YEAR(DATE) = 2015

* ibm_db_sa://wfd01464:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/bludb
Done.

Out[41]: DATE booster_version launch_site
```

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

• The "COUNT" function is used to be able to do a proper ranking, together with a "WHERE" and "BETWEEN" function to meet the established requirements.

```
4]: %sql SELECT LANDING_OUTCOME, COUNT(*) AS outcome_counts
FROM SPACEXTBL
WHERE DATE BETWEEN '2010-06-04' AND '2017-03-20'
GROUP BY LANDING_OUTCOME
ORDER BY count(*) DESC

* ibm_db_sa://wfd01464:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/bludb
Done.

E[44]: landing_outcome outcome_counts

No attempt 10

Failure (drone ship) 5
```

44]:	landing_outcome	outcome_counts
	No attempt	10
	Failure (drone ship)	5
	Success (drone ship)	5
	Controlled (ocean)	3
	Success (ground pad)	3
	Failure (parachute)	2
	Uncontrolled (ocean)	2
	Precluded (drone ship)	1

Interactive map with Folium

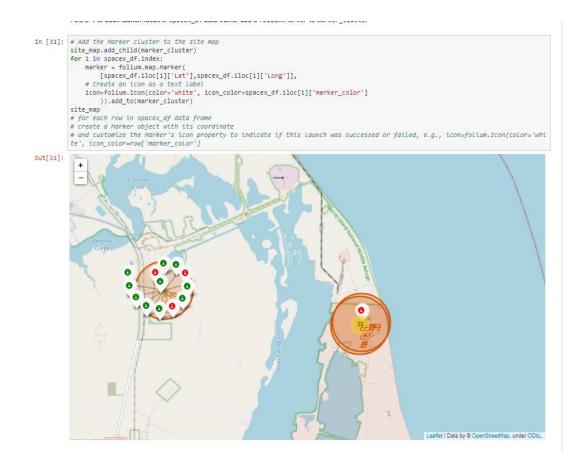
Launch Sites

- In this map you can see the launch sites studied and their geographic distributions.
- As can be seen, they are all found in the United States, specifically in Los Angeles and Florida, in both cases areas that allow landings in the ocean.

```
# Initial the map
site_map = folium.Map(location=nasa_coordinate, zoom_start=5)
# For each Launch site, add a Circle object based on its coordinate (Lat, Long) values. In addition, add Launch site nam
for lon,lat,site in zip(launch sites df.Long, launch sites df.Lat, launch sites df['Launch Site']):
    # Create a blue circle at NASA Johnson Space Center's coordinate with a popup label showing its name
    circle = folium.Circle([lat,lon], radius=1000, color='#d35400', fill=True).add_child(folium.Popup(site))
    # Create a blue circle at NASA Johnson Space Center's coordinate with a icon showing its name
    marker = folium.map.Marker(
        [lat,lon],
    # Create an icon as a text Label
    icon=DivIcon(
        icon size=(20,20),
        icon anchor=(0,0),
        html='<div style="font-size: 12; color:#d35400;"><b>%s</b></div>' % site,
   site_map.add_child(circle)
    site map.add child(marker)
                                                           United States
```

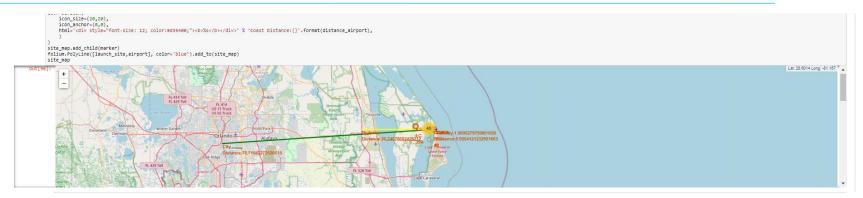
Launch Site Outcomes

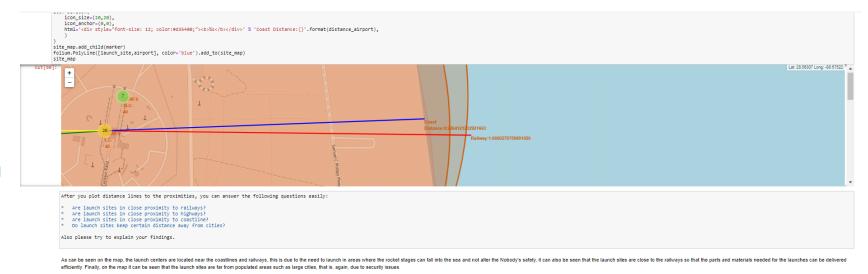
 In this map, in addition to being able to see the geographical distribution of the launch sites, it is also possible to observe the number of successful and failed launches in each one, these are observed in the colors of the icons indicated in each launch site, where the icons Green symbolizes a successful launch, and a red icon symbolizes a failed launch.



<Folium map screenshot 3>

 In the graphical map you can see relevant areas near the launch site, where it is identified that the populated areas are far from it due to security issues, also it is distinguished that the launch sites are in coastal areas so that the remnants rockets can land or fall directly into the ocean and cause no collateral damage.

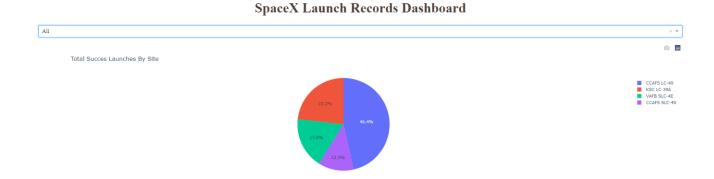




Build a Dashboard with Plotly Dash

SPACEX Launch Dashboard – Pie Chart

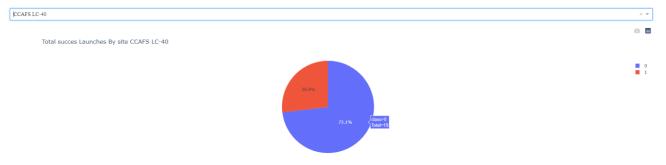
This graph tells us directly which are the launch sites that show the highest number of successful launches, and their proportion with respect to the total number of successful launches.



SPACEX Launch Dashboard – Most Successful Launch Site

From the illustrated figure it is distinguished that the CCAFS LC-40 launch site has a success rate of 73.1% with a total of 19 launches with successful results.

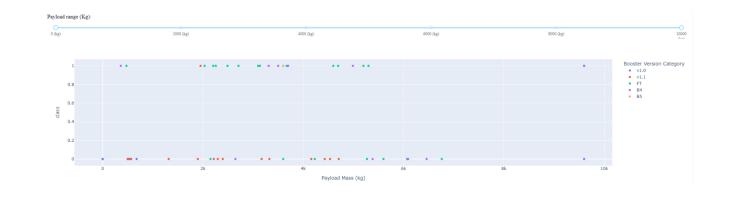
SpaceX Launch Records Dashboard

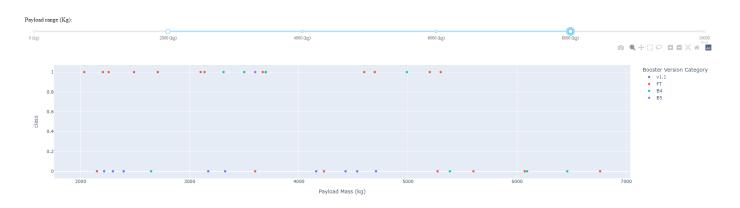


SPACEX Launch Dashboard – Scatter Plot

It can be determined that for all launch sites, if the payload is between 6000 and 9000 kg, the mission is very likely to be a success.

For the versions of Booster v1.0 and v1.1, it is distinguished that most missions were a failure, regardless of the mission payload, for the FT version, most missions with payloads between 2000 and 4000 kg were successful, while when the payload was outside these ranges, its behavior was varied. For the version of Booster B4, it is observed that it has a favorable behavior between 3000 and 5000 kg of payload, while for the rest of the range of payloads, it shows a rather negative behavior, finally, the version of booster B5 has only one launch, so not much can be concluded from its behavior.

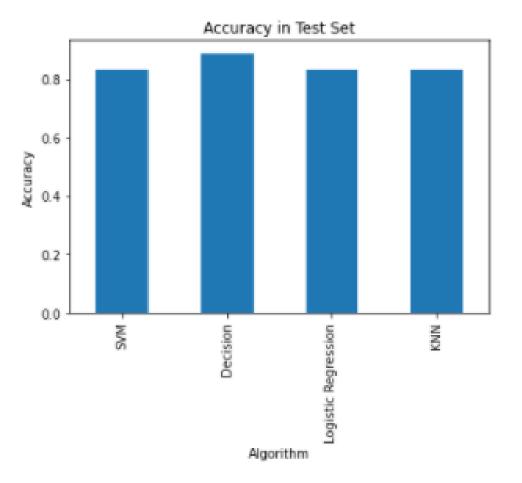




Predictive analysis (Classification)

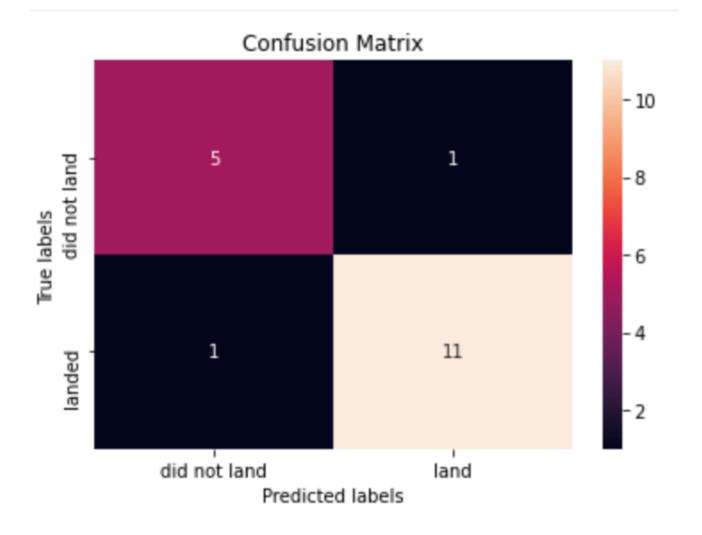
Classification Accuracy

As seen in the image, the model with the best performance in the test set corresponds to a decision tree, so this is the selected model.



Confusion Matrix

As can be seen in the confusion matrix, the algorithm is able to correctly determine the classes 88% of the time, however, in the cases where it is wrong (2), it shows a false positive and a false negative, where the The first of these is the most dangerous case, since the algorithm indicates that the landing would be successful, and finally it is not, which would cause millions of dollars in losses.

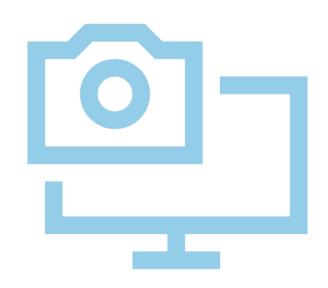


CONCLUSION



- It is possible to determine whether the Falcon 9 phase 1 landing will end successfully with reasonably high precision.
- Many of the variables obtained from various data sources were used to make the prediction of the selected model more accurate.
- Although the developed model presents good results, these could always be improved by adding various variables and sources of information, such as the weather on the day the launch was made, or the wind speed.
- The visualizations made were of the utmost importance to understand the behavior of the variables and their relationship with the dependent variable to be predicted.

APPENDIX: Git Repository



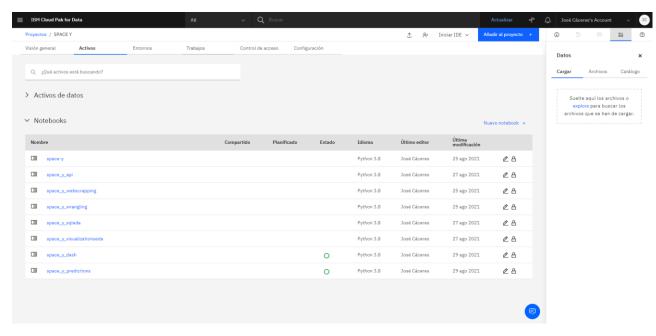
• The link to the repository is at the link below.

https://github.com/joseMCV/space-y

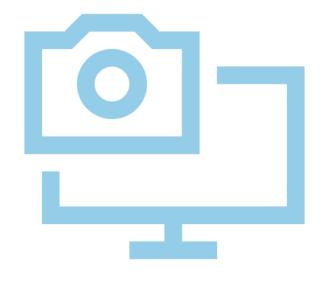
APPENDIX: Plataform



 All the codes created for this work were developed on the IBM platform, in the Watson Studio tool.



APPENDIX: Libraries



- The python libraries used for this project are listed below:
 - Pandas
 - Numpy
 - Sklearn
 - Folium
 - Dash
 - Requests
 - BeautifullSoup
 - Matplotlib
 - Plotly