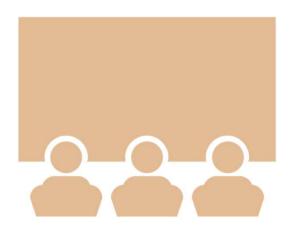
# DATA SCIENCE CAPSTONE PROJECT

#### **RUSHIKESH JAGDALE**

https://github.com/Rushi7171/DataScience

03/18/2024

#### Outline\_



- Executive Summary (3)
- Introduction (4)
- Methodology (6)
- Results (16)
- Conclusion (46)
- Appendix (47)

#### **Executive Summary\_**

- I collected SpaceX data, labeled successful landings, explored using SQL, visualization, and maps. Selected features, encoded variables, standardized data, optimized models, and visualized accuracy scores
- Four ML models (Logistic Regression, SVM, Decision Tree, KNN) achieved ~83.33% accuracy but tended to over-predict successful landings. More data required for improved accuracy.

#### Introduction



SpaceX Falcon 9 Rocket - The Verge

#### • Background:

- Commercial Space Age is Here
- Space X has best pricing (\$62 million vs. \$165 million USD)
- Largely due to ability to recover part of rocket (Stage 1)
- Space Y wants to compete with Space X

#### **Problem:**

 Space Y tasks us to train a machine learning model to predict successful Stage 1 recovery

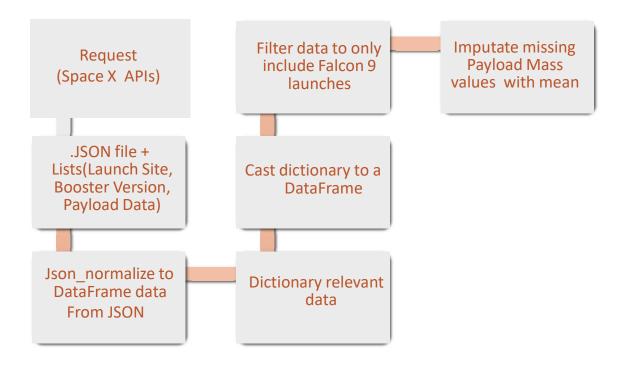
#### Methodology\_

- Data collection methodology: Combined data from SpaceX public API and SpaceX Wikipedia page
- Perform data wrangling- Classifying true landings as successful and unsuccessful otherwise
- Perform exploratory data analysis (EDA) using visualization and SQL
   Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models- Tuned models using GridSearchCV

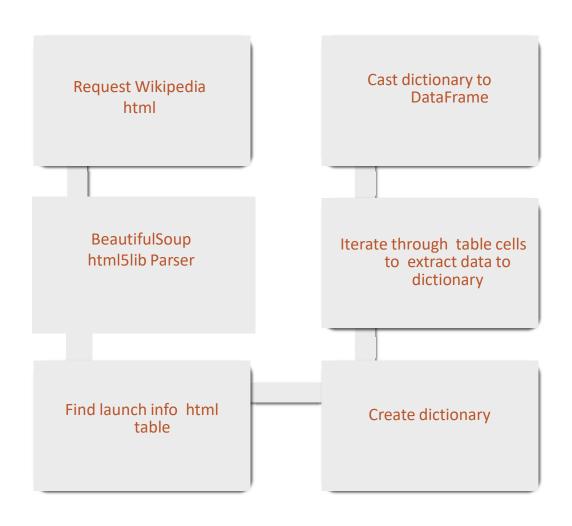
#### Data Collection Overview\_\_\_\_

- Data collection process involved a combination of API requests from Space X public API and web scraping data from a table in Space X's Wikipedia entry.
- The next slide will show the flowchart of data collection from API and the one after will show the flowchart of data collection from webscraping.
- Space X API Data Columns:
- FlightNumber, Date, BoosterVersion, PayloadMass, Orbit, LaunchSite,
   Outcome, Flights, GridFins, Reused, Legs, LandingPad, Block, ReusedCount,
   Serial, Longitude, Latitude
- Wikipedia Webscrape Data Columns: Flight No., Launch site, Payload, PayloadMass, Orbit, Customer, Launch outcome, Version Booster, Booster landing, Date, Time

# Data Collection – SpaceX API



# Data Collection – Web Scraping

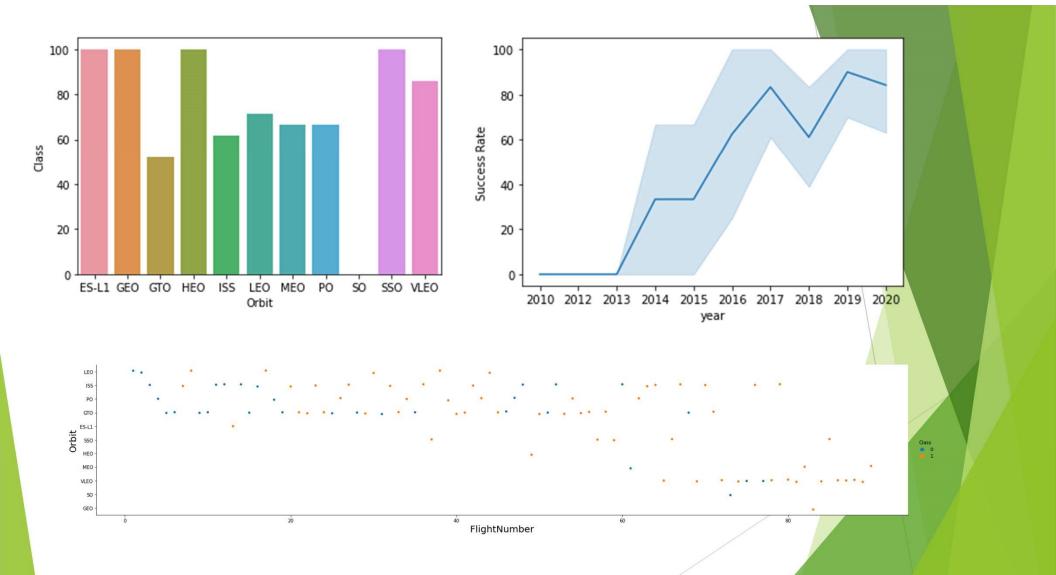


#### Data Wrangling\_\_\_\_

- Create a training label with landing outcomes where successful = 1 & failure = 0.
- Outcome column has two components: 'Mission Outcome' 'Landing Location'
- New training label column 'class' with a value of 1 if 'Mission Outcome' is True and 0 otherwise. Value Mapping:
- True ASDS, True RTLS, & True Ocean set to -> 1
- None None, False ASDS, None ASDS, False Ocean, False RTLS set to -> 0

#### EDA with Data Visualization\_\_\_\_\_

- Conducted thorough analysis on Flight Number, Payload Mass, Launch Site, Orbit, Class, and Year variables.
- Used diverse plots (scatter, line, bar) to explore relationships.
- Focused on key relationships like Flight Number vs. Payload Mass, Launch Site, and Orbit.
- Aiming to identify meaningful patterns for machine learning model training.
- Aimed to identify meaningful patterns for machine learning model training.
- Ensured data readiness for effective model development.



# EDA with SQL\_\_\_\_\_

Loaded data set into IBM DB2 Database.

Queried using SQL Python integration.

Queries were made to get a better understanding of the dataset.

Queried information about launch site names, mission outcomes, various pay load sizes of customers and booster versions, and landing outcomes

#### Task 6

List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less

```
%sql select booster_version from SPACEXDATASET where (mission_outcome like 'Success')
  AND (payload_mass_kg_BETWEEN 4000 AND 6000) AND (landing_outcome like 'Success (drone ship)')
 * ibm_db_sa://nxs27972:***@54a2f15b-5c0f-46df-8954-7e38e612c2bd.c1ogj3sd0tgtu0lqde00.databases.appdom
LUDB
Done.
```

#### booster\_version

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

#### Task 9

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

```
%sql select MONTHNAME(DATE) as Month, landing_outcome, booster_version, launch_site
from SPACEXDATASET where DATE like '2015%' AND landing_outcome like 'Failure (drone ship)'
```

\* ibm db sa://nxs27972:\*\*\*@54a2f15b-5c0f-46df-8954-7e38e612c2bd.clogj3sd0tgtu0lqde00.databases.appdomain.cloud:32733/ LUDB Done.

#### ]: MONTH landing\_outcome booster\_version launch\_site

January	Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
April	Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40

#### Task 10

Rank the count of successful landing\_outcomes between the date 2010-06-04 and 20'

%sql select landing\_outcome, count(\*) as count from SPACEXDATASET where Date >= '2010-06-04' AND Date <= '2017-03-20' GROUP by landing outcome ORDER BY count Desc

\* ibm\_db\_sa://nxs27972:\*\*\*@54a2f15b-5c0f-46df-8954-7e38e612c2bd.c1ogj3sd0tgt LUDB Done.

#### landing\_outcome COUNT

	<b>3</b>
10	No attempt
5	Failure (drone ship)
5	Success (drone ship)
3	Controlled (ocean)
3	Success (ground pad)
2	Failure (parachute)
2	Uncontrolled (ocean)
1	Precluded (drone ship)

## Build an interactive map with Folium:

Folium maps display Launch Sites, successful and unsuccessful landings, and proximity examples to key locations: Railway, Highway, Coast, and City.

The maps provide insight into the rationale behind the selection of launch site locations.

They also visualize successful landings in relation to their geographic location, aiding in location analysis.



#### Build a Dashboard with Plotly Dash

The dashboard features a pie chart and a scatter plot.

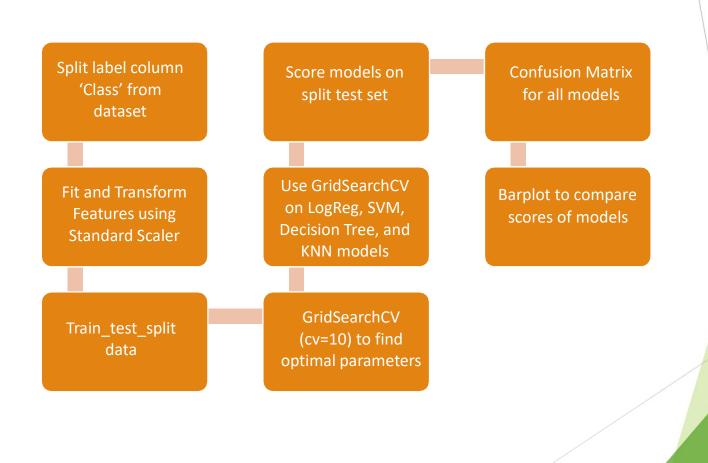
The pie chart allows selection to display the distribution of successful landings across all launch sites or individual launch site success rates.

The scatter plot allows selection of either all sites or individual sites, with a slider for payload mass ranging from 0 to 10000 kg.

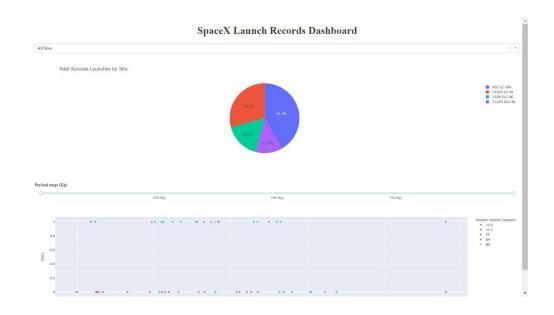
The pie chart visualizes launch site success rates.

The scatter plot helps in observing variations in success rates across launch sites, payload mass, and booster version categories.

## Predictive analysis (Classification)



#### Results

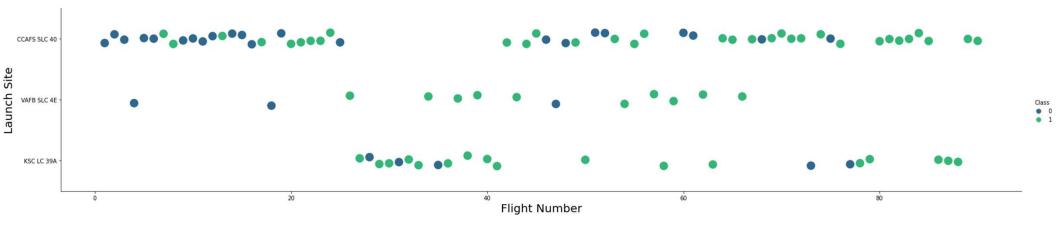


This preview showcases the Plotly dashboard, which includes the outcomes of Exploratory Data Analysis (EDA) through visualization and SQL, an Interactive Map using Folium, and the results of our model with an accuracy of approximately 83%



► EXPLORATORY DATA ANALYSIS WITH SEABORN PLOTS

#### Flight Number Vs Launch Site

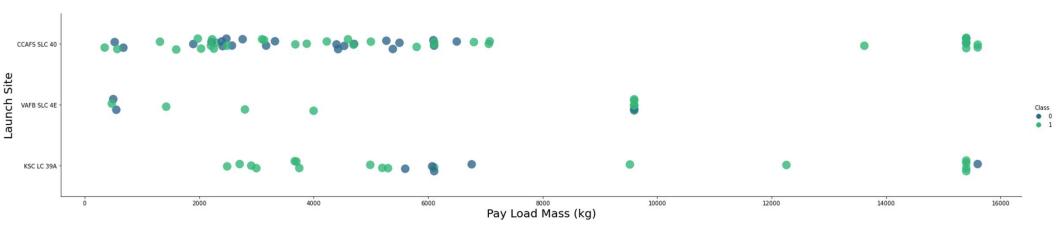


Green indicates successful launch; Purple indicates unsuccessful launch.

The graph indicates a gradual rise in success rates over time, as depicted by the increase in Flight Number. There seems to be a notable breakthrough around flight number 20, leading to a significant spike in success rates.

Additionally, it is evident that CCAFS is the primary launch site, given its higher volume compared to others.

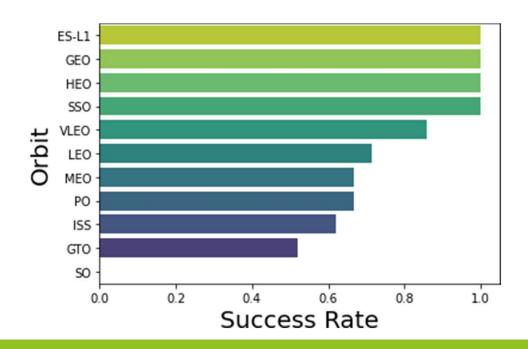
#### Payload Vs Launch Site



Green indicates successful launch; Purple indicates unsuccessful launch.

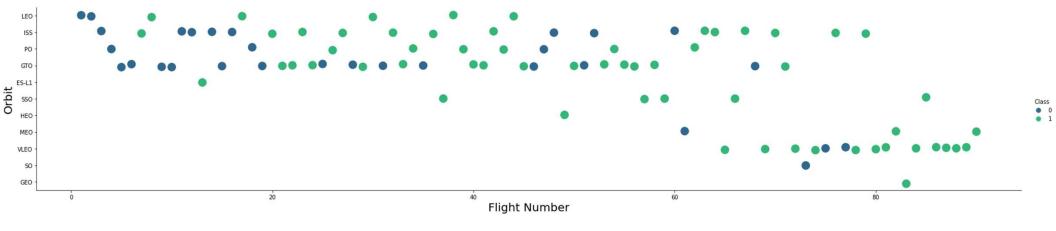
The analysis reveals that the majority of payload masses fall within the range of 0-6000 kg. Furthermore, it appears that different launch sites employ varying payload masses for their missions.

#### Success rate Vs Orbit type



ES-L1, GEO, and HEO, each with a sample size of 1, achieved a 100% success rate. Similarly, SSO, with a sample size of 5, also attained a 100% success rate. VLEO, with 14 attempts, demonstrated a respectable success rate. In contrast, SO, with a single attempt, experienced a 0% success rate. Notably, GTO, with the largest sample size of 27, achieved a success rate of approximately 50%

#### Flight Number Vs Orbit type

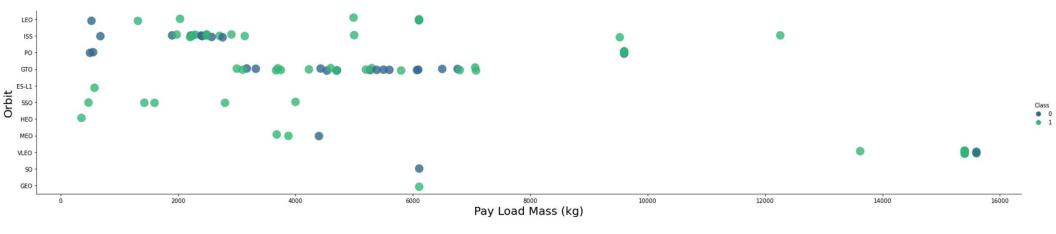


Green indicates successful launch; Purple indicates unsuccessful launch.

SpaceX's launch orbit preferences have changed over time, correlating with Flight Number and Launch Outcome.

Initially focusing on LEO missions, success rates were moderate. Recent launches show a return to VLEO, suggesting better performance in lower or Sun-synchronous orbits.

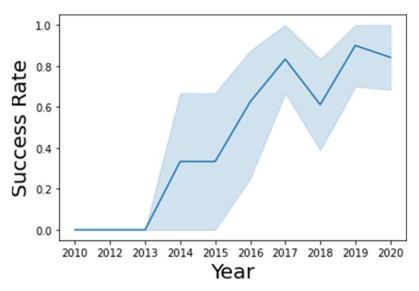
#### Payload Vs Orbit type



Green indicates successful launch; Purple indicates unsuccessful launch.

Payload mass appears to correlate with the orbit type, with Low Earth Orbit (LEO) and Sun-Synchronous Orbit (SSO) exhibiting relatively low payload masses. Conversely, the Very Low Earth Orbit (VLEO), which is among the most successful orbits, tends to have payload mass values at the higher end of the range.

#### **Launch Success Yearly Trend**



95% confidence interval (light blue shading)

Success rates have generally shown an upward trend since 2013, albeit with a slight dip observed in 2018. In recent years, success rates have stabilized at around 80%.

# **EDA** with SQL

EXPLORATORY DATA ANALYSIS WITH SQL DB2 INTEGRATED IN PYTHON WITH SQLALCHEMY

#### **All Launch Site Names**

```
In [4]: %%sql
SELECT UNIQUE LAUNCH_SITE
FROM SPACEXDATASET;

* ibm_db_sa://ftb12020:***@0c77d6f;
Done.

Out[4]: launch_site
CCAFS LC-40
CCAFS SLC-40
CCAFSSLC-40
KSC LC-39A
VAFB SLC-4E
```

- Retrieve unique launch site names from the database.
- There are likely data entry errors, as "CCAFS SLC-40" and "CCAFSSLC-40" likely represent the same launch site.
- "CCAFS LC-40" was the previous name. Therefore, there are likely only three unique launch site values:
- "CCAFS SLC-40", "KSC LC-39A", and "VAFB SLC-4E".

# Launch Site Names Beginning with `CCA`

 $* ibm\_db\_sa://ftb12020:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludbDone.$ 

Out[5]:

DATE	timeutc_	booster_version	launch_site	payload	payload_masskg_	orbit	customer	mission_outcome	landing_outcome
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)
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
2013- 03-01	15:10:00	F9 v1.0 B0007	CCAFS LC- 40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

Retrieve the first five entries from the database where the Launch Site name begins with "CCA".

#### **Total Payload Mass from NASA**

```
%%sql
SELECT SUM(PAYLOAD_MASS__KG_) AS SUM_PAYLOAD_MASS_KG
FROM SPACEXDATASET
WHERE CUSTOMER = 'NASA (CRS)';
```

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-81f8-86 Done.

```
sum_payload_mass_kg
45596
```

This query sums the total payload mass in kg where NASA was the customer.

CRS stands for Commercial Resupply
Services which indicates that these payloads
were sent to the International Space Station
(ISS).

## Average Payload Mass by F9 v1.1

```
%%sql
SELECT AVG(PAYLOAD_MASS__KG_) AS AVG_PAYLOAD_MASS_KG
FROM SPACEXDATASET
WHERE booster_version = 'F9 v1.1'
```

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-81f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-80f8-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-48a9-8@0c77d6f2-5da9-8@0c77d6f2-8@0c77d66f2-8@0c77d66f2-8@0c77d6f2-8@0c77d66f2-8@0c77d66f2-8@0c77d66f2-

avg\_payload\_mass\_kg

This query calculates the average payload mass or launches which used booster version F9 v1.1

Average payload mass of F9 1.1 is on the low end of our payload mass range

## First Successful Ground Pad Landing Date

```
%%sql
SELECT MIN(DATE) AS FIRST_SUCCESS
FROM SPACEXDATASET
WHERE landing_outcome = 'Success (ground pad)';
```

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-81 Done.

first\_success

2015-12-22

This query returns the first successful ground pad landing date.

First ground pad landing wasn't until the end of 2015.

Successful landings in general appear starting 2014.

## Successful Drone Ship Landing with Payload 4000 and 6000

%%sql SELECT booster version FROM SPACEXDATASET WHERE landing\_outcome = 'Success (drone ship)' AND payload\_mass\_kg\_ BETWEEN 4001 AND 5999;

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.database Done.

#### booster version

F9 FT B1022

F9 FT B1026

F9 FT B1021.2

F9 FT B1031.2

This query returns the four booster versions that had successful drone ship landings and a payload mass between 4000 and 6000 noninclusively.

#### **Total Number of Each Mission Outcome**

%%sql
SELECT mission\_outcome, COUNT(\*) AS no\_outcome
FROM SPACEXDATASET
GROUP BY mission\_outcome;

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-

mission_outcome	no_outcome
Failure (in flight)	1
Success	99
Success (payload status unclear)	1

This query returns a count of each mission outcome.

SpaceX appears to achieve its mission outcome nearly 99% of the time.

This means that most of the landing failures are intended.

Interestingly, one launch has an unclear payload status and unfortunately one failed in flight.

# **Boosters that Carried Maximum Payloa**

```
%%sql
SELECT booster_version, PAYLOAD_MASS__KG_
FROM SPACEXDATASET
WHERE PAYLOAD_MASS__KG_ = (SELECT MAX(PAYLOAD_MASS__KG_) FROM SPACEXDATASET);
```

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1 Done.

booster_version	payload_masskg_
F9 B5 B1048.4	15600
F9 B5 B1049.4	15600
F9 B5 B1051.3	15600
F9 B5 B1056.4	15600
F9 B5 B1048.5	15600
F9 B5 B1051.4	15600
F9 B5 B1049.5	15600
F9 B5 B1060.2	15600
F9 B5 B1058.3	15600
F9 B5 B1051.6	15600
F9 B5 B1060.3	15600
F9 B5 B1049.7	15600

This query returns the booster versions that carried the highest payload mass of 15600 kg.

These booster versions are very similar and all are of the F9 B5 B10xx.x variety.

This likely indicates payload mass correlates with the booster version that is used.

# 2015 Failed Drone Ship Landing Records

#### %%sql

SELECT MONTHNAME(DATE) AS MONTH, landing\_outcome, booster\_version, PAYLOAD\_MASS\_\_KG\_, launch\_site FROM SPACEXDATASET

WHERE landing\_outcome = 'Failure (drone ship)' AND YEAR(DATE) = 2015;

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.app Done.

MONTH	landing_outcome	booster_version	payload_masskg_	launch_site
January	Failure (drone ship)	F9 v1.1 B1012	2395	CCAFS LC-40
April	Failure (drone ship)	F9 v1.1 B1015	1898	CCAFS LC-40

This query returns the Month, Landing Outcome, Booster Version, Payload Mass (kg), and Launch site of 2015 launches where stage 1 failed to land on a drone ship.

There were two such occurrences.

# Ranking Counts of Successful Landings Between 2010-06-04 and 2017-03-20

```
%%sql
SELECT landing_outcome, COUNT(*) AS no_outcome
FROM SPACEXDATASET
WHERE landing_outcome LIKE 'Succes%' AND DATE BETWEEN '2010-06-04' AND '2017-03-20'
GROUP BY landing_outcome
ORDER BY no_outcome DESC;
```

\* ibm\_db\_sa://ftb12020:\*\*\*@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg Done.

landing_outcome	no_outcome
Success (drone ship)	5
Success (ground pad)	3

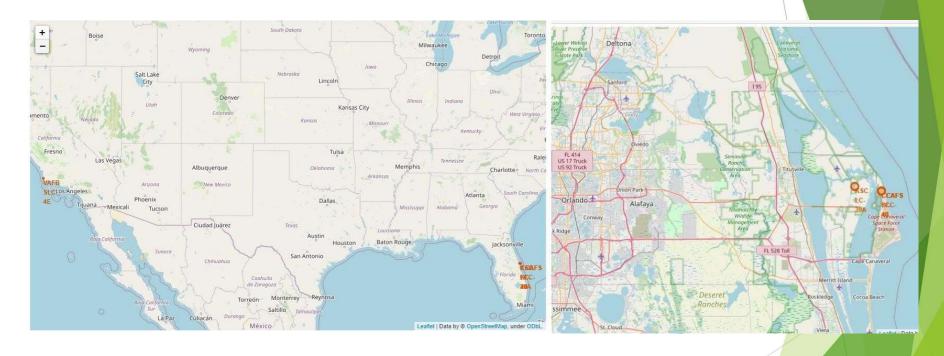
This query returns a list of successful landings and between 2010-06-04 and 2017-03-20 inclusively.

There are two types of successful landing outcomes: drone ship and ground pad landings.

There were 8 successful landings in total during this time period

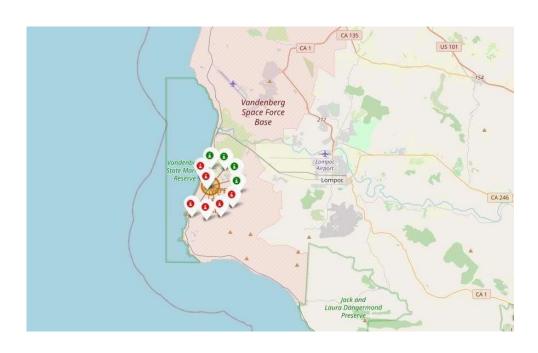
# Interactive Map with Folium

## **Launch Site Locations**



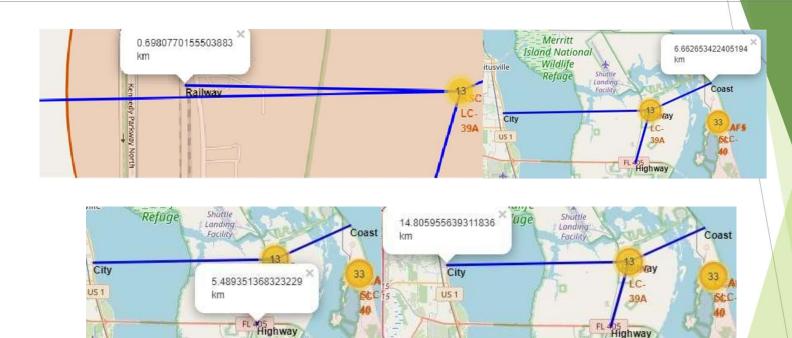
The left map shows all launch sites relative to the US map, while the right map focuses on the two Florida launch sites due to their proximity, all of which are situated near the ocean

## **Color-Coded Launch Markers**



Clusters on the Folium map are clickable, revealing individual successful (green icon) and failed (red icon) landings. For instance, VAFB SLC-4E exhibits 4 successful landings and 6 failed landings in this example.

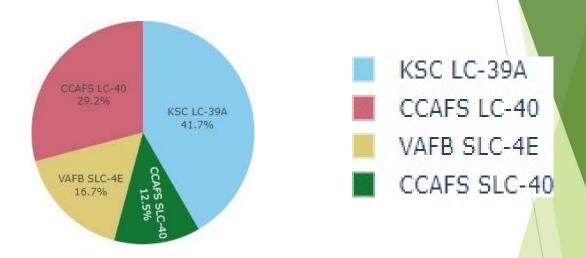
## **Key Location Proximities**



Launch sites like KSC LC-39A are positioned near railways for supply transportation, highways for accessibility, and coastlines to mitigate risks of launch failures in densely populated areas.

# Build a Dashboard with Plotly Dash

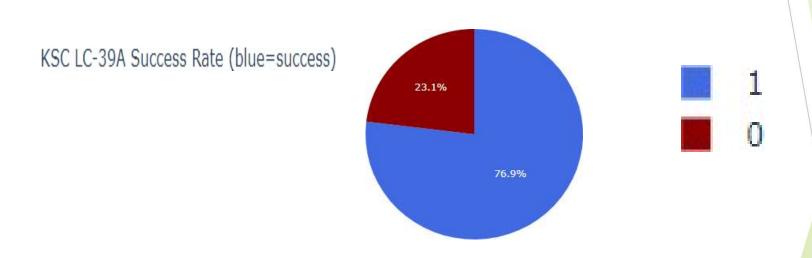
## Successful Launches Across Launch Site



CCAFS LC-40, now known as CCAFS SLC-40, and KSC have an equal number of successful landings. However, most successful landings at CCAFS occurred before the name change.

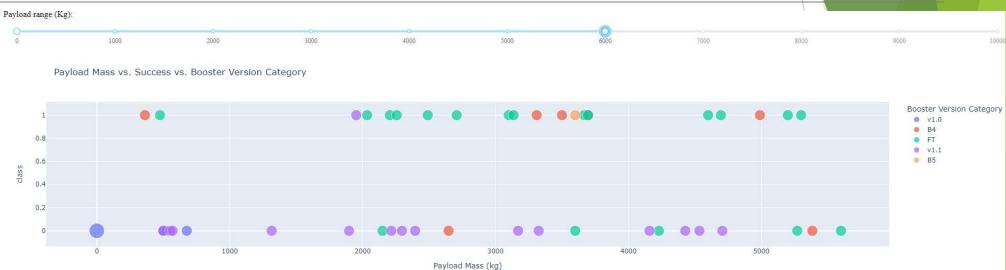
VAFB has the smallest proportion of successful landings, possibly due to a smaller sample size and the increased difficulty of launching on the west coast.

## **Highest Success Rate Launch Site**



KSC LC-39A boasts the highest success rate, with 10 successful landings and 3 failed landings.

## Payload Mass vs. Success vs. Booster Category

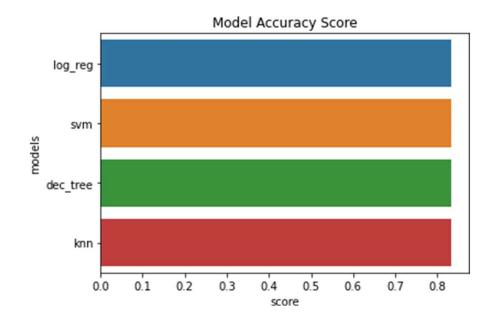


- The Plotly dashboard's payload range selector is set from 0 to 10000 instead of the max payload
  of 15600.
- "Class" indicates 1 for successful landings and 0 for failures.
- The scatter plot includes booster version category as color and number of launches as point size.
- Within the 0 to 6000 payload range, two failed landings with payloads of zero kg are observed, which is noteworthy.

## Predictive Analysis (Classification)

GRIDSEARCHCV(CV=10) ON LOGISTIC REGRESSION, SVM, DECISION TREE, AND KNN

## **Classification Accuracy**



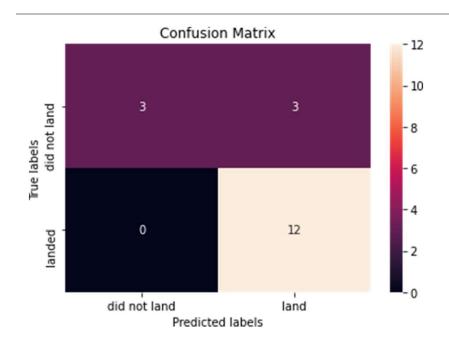
All models achieved an 83.33% accuracy rate on the test set.

However, the test set's small size, consisting of only 18 samples, can result in considerable accuracy variance.

The Decision Tree Classifier model showed notable variability in accuracy across repeated runs.

To reliably determine the best model, acquiring more data is recommended.

## **Confusion Matrix**



Correct predictions are on a diagonal from top left to bottom right.

Observation: The models tend to overpredict successful landings, as indicated by the false positives.

#### **Confusion Matrix Analysis:**

True Positive (TP): Models predicted 12 successful landings correctly.

True Negative (TN): Models predicted 3 unsuccessful landings correctly.

False Positive (FP): Models incorrectly predicted 3 successful landings when the true label was unsuccessful landings.

False Negative (FN): Not specified in the provided text.

### CONCLUSION

- Task: Develop a machine learning model for Space Y to predict successful Stage 1 landings and potentially save \$100 million USD.
- Data Sources: Utilized data from a public SpaceX API and web scraping SpaceX
   Wikipedia page.
- Data Handling: Created data labels and stored data into a DB2 SQL database.
- Visualization: Developed a dashboard for visualization purposes.
- Machine Learning Model: Built a machine learning model with an accuracy of 83%.
- Application: Allon Mask of SpaceY can use this model to predict with relatively high accuracy whether a launch will have a successful Stage 1 landing before launch, aiding in decision-making.
- Future Improvements: Suggested collecting more data to further improve accuracy and determine the best machine learning model.

## **APPENDIX**

GitHub repository URL: <a href="https://github.com/Rushi717171/DataScience">https://github.com/Rushi717171/DataScience</a>

Instructors: Rav Ahuja, Alex Aklson, Aije Egwaikhide, Svetlana Levitan, Romeo Kienzler, Polong Lin, Joseph Santarcangelo, Azim Hirjani, Hima Vasudevan, Saishruthi Swaminathan, Saeed Aghabozorgi, Yan Luo

Special Thanks to All Instructors:

https://www.coursera.org/professional-certificates/ibm-data-science?#instructors