



IBM Developer
SKILLS NETWORK

Winning Space Race with Data Science

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EXECUTIVE SUMMARY

CAPABILITIES & SERVICES

SpaceX offers competitive pricing for its Falcon 9 and Falcon Heavy launch services. Modest discounts are available, for contractually committed, multi-launch purchases. SpaceX can also offer crew transportation services to commercial customers seeking to transport astronauts to alternate LEO destinations.

PRICE	FALCON 9
STANDARD PAYMENT PLAN (THROUGH 2022)	\$62 M UP TO 5.5 mT TO GTO
DESTINATION	PERFORMANCE*
LOW EARTH ORBIT (LEO)	22,800 kg 50,265 lbs
GEOSYNCHRONOUS TRANSFER ORBIT (GTO)	8,300 kg 18,300 lbs
PAYLOAD TO MARS	4,020 kg 8,860 lbs

*Performance represents max capability on fully expendable vehicle.



- SpaceY is a new commercial rocket launch provider who wants to bid against SpaceX.
- SpaceX advertises launch services starting at \$62 million for missions that allow some fuel to be reserved for landing the 1st stage rocket booster, so that it can be reused.
- SpaceX public statements indicate a 1st stage Falcon 9 booster to cost upwards of \$15 million to build without including R&D cost recoupment or profit margin.
- Given mission parameters such as payload mass and desired orbit, the models produced in this report were able to predict the first stage rocket booster landing successfully with an accuracy level of 83.3%.
- As a result, SpaceY will be able to make more informed bids against SpaceX by using 1st stage landing predictions as a proxy for the cost of a launch.

INTRODUCTION: BACKGROUND



- This report has been prepared as part of the Applied Data Science Capstone course.*
- In this capstone, I take the role of a data scientist working for a new rocket company called SpaceY.
- With the help of the data science findings and models in this report, SpaceY will be able to make more informed bids against SpaceX for a rocket launch.

* 10th course in the [IBM Data Science Professional Certification](#)

INTRODUCTION: BUSINESS PROBLEM



Compilation of early attempts at propulsive landing prior to the first success in 2016

- SpaceX advertises Falcon 9 rocket launches with a cost of 62 million dollars when the first stage of their rockets can be reused.
- The first stage is estimated to cost upwards of 15 million to build without including R&D cost recoupment or profit margin.
- Sometimes SpaceX will sacrifice the first stage due to mission parameters such as payload, orbit, and customer.
- Therefore this report aims to accurately predict the likelihood of the first stage rocket landing successfully as a proxy for the cost of a launch.

Section 1

Methodology

METHODOLOGY

For this report the data science methodology used can be outlined as such:

1. Data collection
2. Data wrangling
3. Exploratory data analysis
4. Data visualization
5. Model development
6. Reporting results to stakeholders

METHODOLOGY

Past launches | [view](#) | [edit](#)

2010 to 2013 | [view](#) | [edit](#)

Flight No.	Date and time (UTC)	Version, Booster	Launch site	Payload ⁽¹⁾	Payload mass	Orbit	Customer	Launch outcome	Booster landing
1	4 June 2010, 18:45	F9 v1.0 ⁽¹⁾ B0003 ⁽²⁾	CCAFS, SLC-40	Dragon Spacecraft Qualification Unit		LEO	SpaceX	Success	Failure ⁽¹⁾⁽³⁾ (parachute)
First flight of Falcon 9 v1.0 ⁽¹⁾ Used a test-payload version of Dragon capsule which was not designed to separate from the second stage (now defunct) Attempted to recover the first stage by parachuting it into the ocean, but it burned up on reentry before the parachutes even got to deploy. ⁽¹⁾⁽³⁾									
2	8 December 2010, 15:45 ⁽¹⁾	F9 v1.0 ⁽¹⁾ B0004 ⁽²⁾	CCAFS, SLC-40	Dragon demo flight C1 (Dragon C101)		LEO (ISS)	NASA (COTS) RPO	Success ⁽¹⁾	Failure ⁽¹⁾⁽³⁾ (parachute)
Maiden flight of SpaceX's Dragon capsule, consisting of over 9 hours of testing thruster maneuvering and then reentry. ⁽¹⁾ Attempted to recover the first stage by parachuting it into the ocean, but it disintegrated upon reentry again before the parachutes were deployed. ⁽¹⁾⁽³⁾ (see details below) It also included two CubeSats ⁽¹⁾⁽³⁾ and a wheel of silicone cheese. Before the launch, SpaceX discovered that there was a crack in the nozzle of the 2nd stage's Merlin vacuum engine. So Elon just had them cut off the end of the nozzle with a pair of shears and launched the rocket a few days later. After SpaceX had trimmed the nozzle, NASA was notified of the change and they agreed to it. ⁽¹⁾⁽³⁾									
3	22 May 2012, 07:44 ⁽¹⁾⁽³⁾	F9 v1.0 ⁽¹⁾ B0005 ⁽²⁾	CCAFS, SLC-40	Dragon demo flight C2 ⁽¹⁾⁽³⁾ (Dragon C102)	525 kg (1,157 lb) ⁽²⁾⁽³⁾	LEO (ISS)	NASA (COTS)	Success ⁽¹⁾⁽³⁾	No attempt
The Dragon spacecraft demonstrated a series of tests before it was allowed to approach the International Space Station. Two days later, it became the first commercial spacecraft to boost the ISS. ⁽¹⁾⁽³⁾ (see details below)									
4	8 October 2012, 00:35 ⁽²⁾	F9 v1.0 ⁽¹⁾ B0006 ⁽²⁾	CCAFS, SLC-40	SpaceX CRS-1 ⁽¹⁾⁽³⁾ (Dragon C103) Orion-OS ⁽²⁾⁽³⁾	4,700 kg (10,400 lb) 172 kg (379 lb) ⁽²⁾⁽³⁾	LEO (ISS) LEO	NASA (CRS) Orion	Success Partial failure ⁽²⁾⁽³⁾	No attempt
CRS-1 was successful, but the secondary payload was inserted into an abnormally low orbit and subsequently lost. This was due to one of the nine Merlin engines shutting down during the launch, and NASA declining a second re-ignition, as per ISS visiting vehicle safety rules. The primary payload owner is contractually allowed to decline a second re-ignition. NASA stated that this was because SpaceX could not guarantee a high enough likelihood of the second stage completing the second burn successfully which was required to avoid any risk of secondary payload's collision with the ISS. ⁽²⁾⁽³⁾⁽³⁾									
	1 March 2013	F9 v1.0 ⁽¹⁾	CCAFS	SpaceX CRS-2 ⁽²⁾⁽³⁾	4,877 kg	LEO	NASA	Success	No attempt

What the first page of launch data looked like on Wikipedia prior to web scraping

- Data Collection

- API

- Acquired historical launch data from Open Source REST API for SpaceX

- Requested and parsed the SpaceX launch data using the GET request
 - Filtered the dataframe to only include Falcon 9 launches
 - Replaced missing payload mass values from classified missions with mean

- Web Scraping

- Acquired historical launch data from Wikipedia page 'List of Falcon 9 and Falcon Heavy Launches'

- Requested the Falcon9 Launch Wiki page from its Wikipedia URL
 - Extracted all column/variable names from the HTML table header
 - Parsed the table and converted it into a Pandas data frame

Note: Falcon 9 launch dataset was limited to launches before December 7, 2020 per instructions.

METHODOLOGY

Landing Outcomes

sample size = 90

□ = Class 0

□ = Class 1

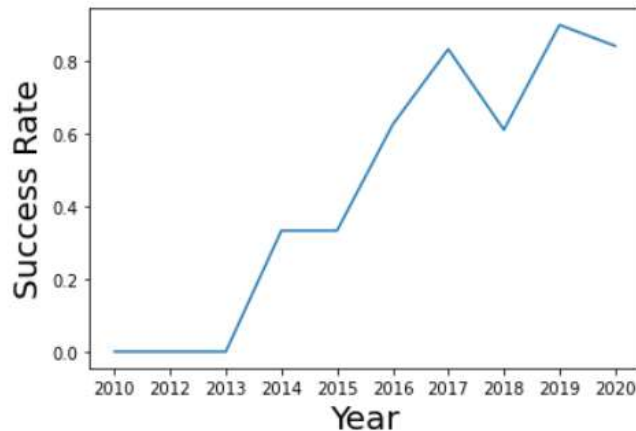
True ASDS	41
None None	19
True RTLS	14
False ASDS	6
True Ocean	5
None ASDS	2
False Ocean	2
False RTLS	1

- Data Wrangling

- Explored data to determine the label for training supervised models
 - Calculated the number of launches on each site
 - Calculated the number and occurrence of each orbit
 - Calculated the number and occurrence of mission outcome per orbit type
- Created a landing outcome training label from 'Outcome' column
 - Training label: 'Class'
 - Class = 0; first stage booster did not land successfully
 - None None; not attempted
 - None ASDS; unable to be attempted due to launch failure
 - False ASDS; drone ship landing failed
 - False Ocean; ocean landing failed
 - False RTLS; ground pad landing failed
 - Class = 1; first stage booster landed successfully
 - True ASDS; drone ship landing succeeded
 - True RTLS; ground pad landing succeeded
 - True Ocean; ocean landing succeeded

METHODOLOGY

```
In [11]: # Plot a line chart with x axis to be the extracted year and y axis to be the success rate
df1=pd.DataFrame(Extract_year(df['Date']),columns=['year'])
df1['Class']=df['Class']
sns.lineplot(data=df1, x=np.unique(Extract_year(df['Date'])), y=df1.groupby('year')['Class'].mean())
plt.xlabel("Year", fontsize=20)
plt.ylabel("Success Rate", fontsize=20)
plt.show()
```



screenshot of Year vs. Success rate* plot
*(of 1st stage booster landing)

- Exploratory Data Analysis (EDA)

- EDA with SQL

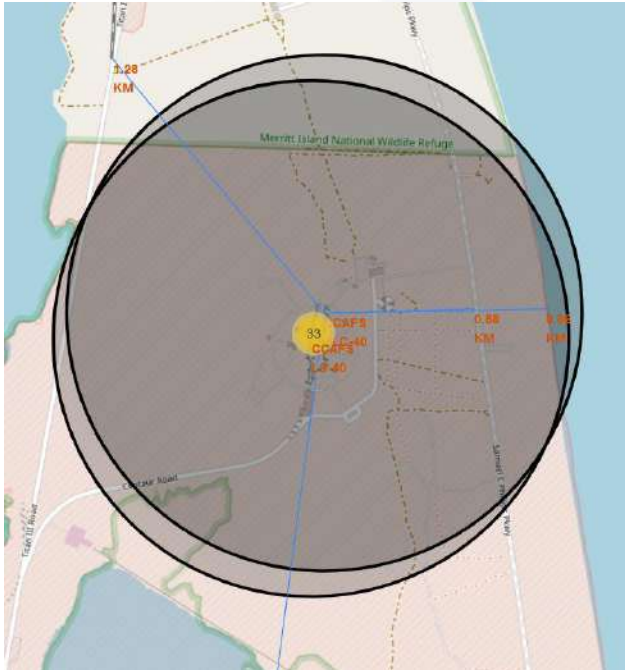
- Loaded data into an IBM DB2 instance
 - Ran SQL queries to display and list information about
 - Launch sites
 - Payload masses
 - Booster versions
 - Mission outcomes
 - Booster landings

- EDA with visualization

- Read the dataset into a Pandas dataframe
 - Used Matplotlib and Seaborn visualization libraries to plot
 - FlightNumber x PayloadMass †
 - FlightNumber x LaunchSite †
 - Payload x LaunchSite †
 - Orbit type x Success rate
 - FlightNumber x Orbit type †
 - Payload x Orbit type †
 - Year x Success rate

† = with Class overlayed (1st stage booster landing outcome)

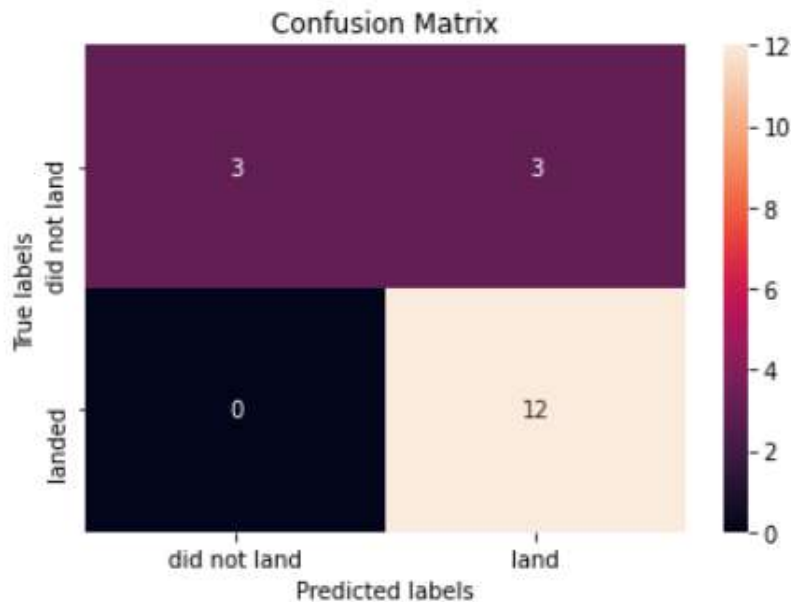
METHODOLOGY



Screenshot of interactive Folium map showing proximity from CCAFS SLC-40 launch site to nearby railway, highway, and coastline

- Data Visualization
 - Launch Sites Location Analysis
 - Used Python interactive mapping library called Folium
 - Marked all launch sites on a map
 - Marked the successful/failed launches for each site on map
 - Calculated the distances between a launch site to its proximities
 - Railways
 - Highways
 - Coastlines
 - Cities
 - Launch Records Dashboard
 - Used Python interactive dashboarding library called Plotly Dash to enable stakeholders to explore and manipulate data in an interactive and real-time way
 - Pie chart showing success rate
 - Colorcoded by launch site
 - Scatter chart showing payload mass vs. landing outcome
 - Colorcoded by booster version
 - With range slider for limiting payload amount
 - Drop-down menu to choose between all sites and individual launch sites
 - Deployed to Heroku static web app hosting service
<https://ibm-applied-data-science-capst.herokuapp.com/>

METHODOLOGY



Confusion matrix of logistic regression model, showing 15 correct predictions and 3 false positives

- Predictive Analysis (Model development)
 - Imported libraries and defined function to create confusion matrix
 - Pandas
 - Numpy
 - Matplotlib
 - Seaborn
 - Sklearn
 - Loaded the dataframe created during data collection
 - Created a column for our training label 'Class' created during data wrangling
 - Standardized the data
 - Split the data into training data and test data
 - Fit the training data to various model types
 - Logistic Regression
 - Support Vector Machine
 - Decision Tree Classifier
 - K Nearest Neighbors Classifier
 - Used a cross-validated grid-search over a variety of hyperparameters to select the best ones for each model
 - Enabled by Scikit-learn library function GridSearchCV
 - Evaluated accuracy of each model using test data to select the best model

RESULTS

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



The background of the slide is an abstract composition. It features a solid blue area on the left side, which transitions into a dynamic pattern of diagonal streaks in shades of blue, red, and teal on the right. These streaks have a textured, almost woven appearance. Overlaid on this pattern is a faint, light blue grid that recedes into the background, creating a sense of depth and digital structure.

Section 2

Insights drawn from EDA

RESULTS: EDA WITH SQL

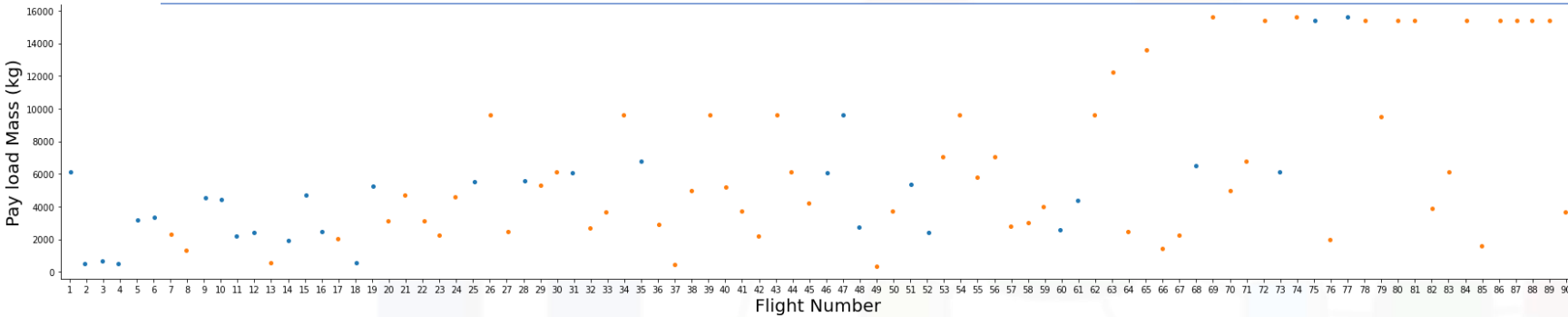
The team at SpaceY had some very specific questions to answer with SQL:

- What launch sites has SpaceX used?
 - CCAFS LC-40
 - CCAFS SLC-40
 - KSC LC-39A
 - VAFB SLC-4E
- Examine launch site and date records where launch sites begin with the string 'CCA', do they overlap?
 - Last launch from CCAFS LC-40 was 2016-08-14
 - First launch from CCAFS SLC-40 was 2017-12-15
 - [Wikipedia confirms Cape Canaveral Space Launch Complex 40 was renamed in 2017](#)
- Display the total payload mass carried by boosters launched by NASA (CRS)
 - 45,596 KG, total
- Display average payload mass carried by booster version F9 v1.1
 - 340 KG, average
- List the date when the first successful landing outcome in ground pad was achieved.
 - 2015-12-22, more than 5 years after the first Falcon 9 launch on 2010-06-04

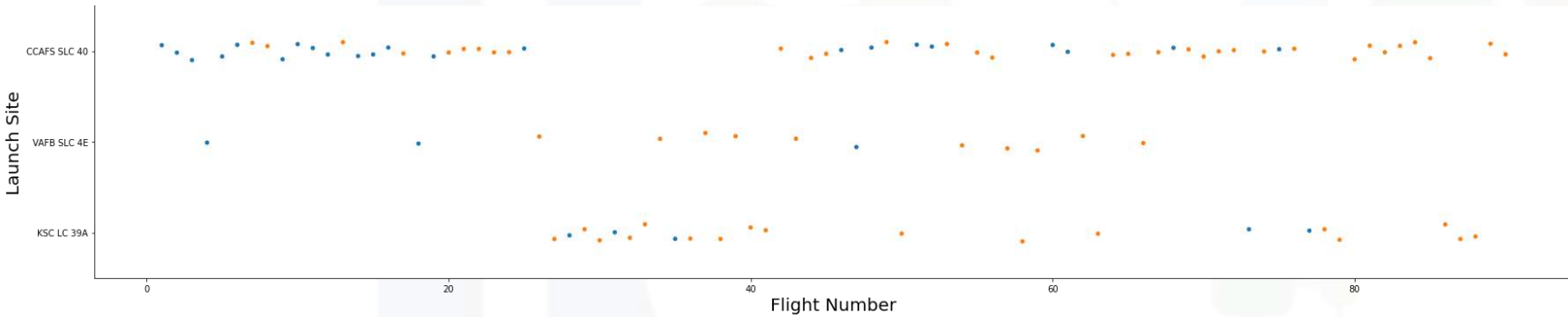
RESULTS: EDA WITH SQL (CONTINUED)

- List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
 - F9 FT B1021.1
 - F9 FT B1023.1
 - F9 FT B1029.2
 - F9 FT B1038.1
 - F9 B4 B1042.1
 - F9 B4 B1045.1
 - F9 B5 B1046.1
- Rank the count of landing outcomes between the date 2010-06-04 and 2017-03-20, in descending order
 - 10 - No attempt5 - Failure (drone ship)
 - 5 - Success (drone ship)
 - 3 - Controlled (ocean)
 - 3 - Success (ground pad)
 - 2 - Failure (parachute)
 - 2 - Uncontrolled (ocean)
 - 1 - Precluded (drone ship)
- List the names of the booster_versions which have carried the maximum payload mass.
 - 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
- List the failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015
 - Failure (drone ship) F9 v1.1 B1012 CCAFS LC-40
 - Failure (drone ship) F9 v1.1 B1015 CCAFS LC-40
- List the total number of successful and failure mission outcomes
 - 1 - Failure (in flight)
 - 99 - Success
 - 1 - Success (payload status unclear)

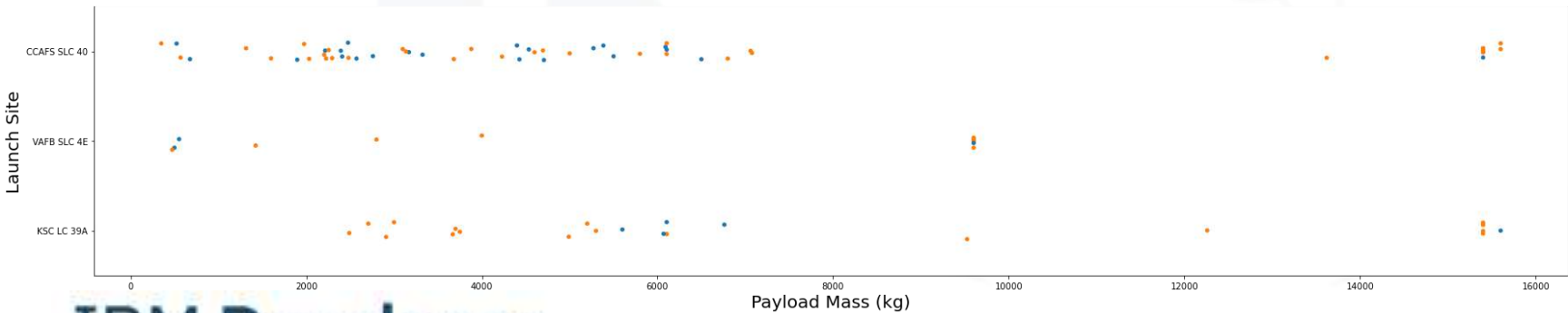
RESULTS: EDA WITH VISUALIZATION



FlightNumber x PayloadMass,
1st stage landing success positively
correlated with continuous launch
attempts, while negatively correlated
with payload mass

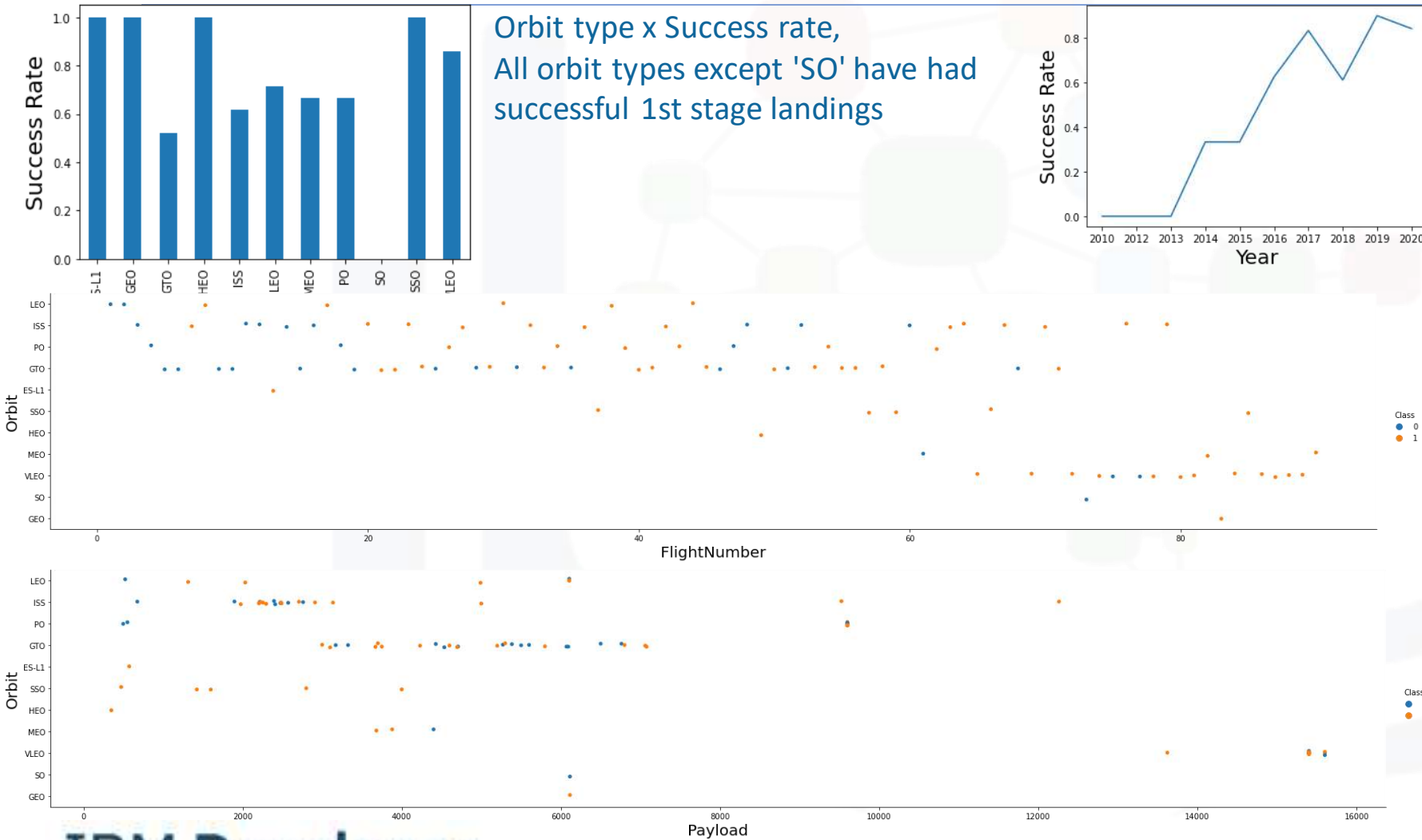


FlightNumber x LaunchSite,
CCAFS SLC 40 appears to have been
where most of the early 1st stage landing
failures took place



PayloadMass x LaunchSite,
CCAFS SLC 40 and KSC LC 39A appear to
be favored for heavier payloads

RESULTS: EDA WITH VISUALIZATION (CONTINUED)



Orbit type x Success rate,
All orbit types except 'SO' have had
successful 1st stage landings

Year x Success rate,
success rate trending positively on a
yearly basis since 2013

FlightNumber x Orbit type,
flight number positively correlated with
1st stage recovery for all orbit types

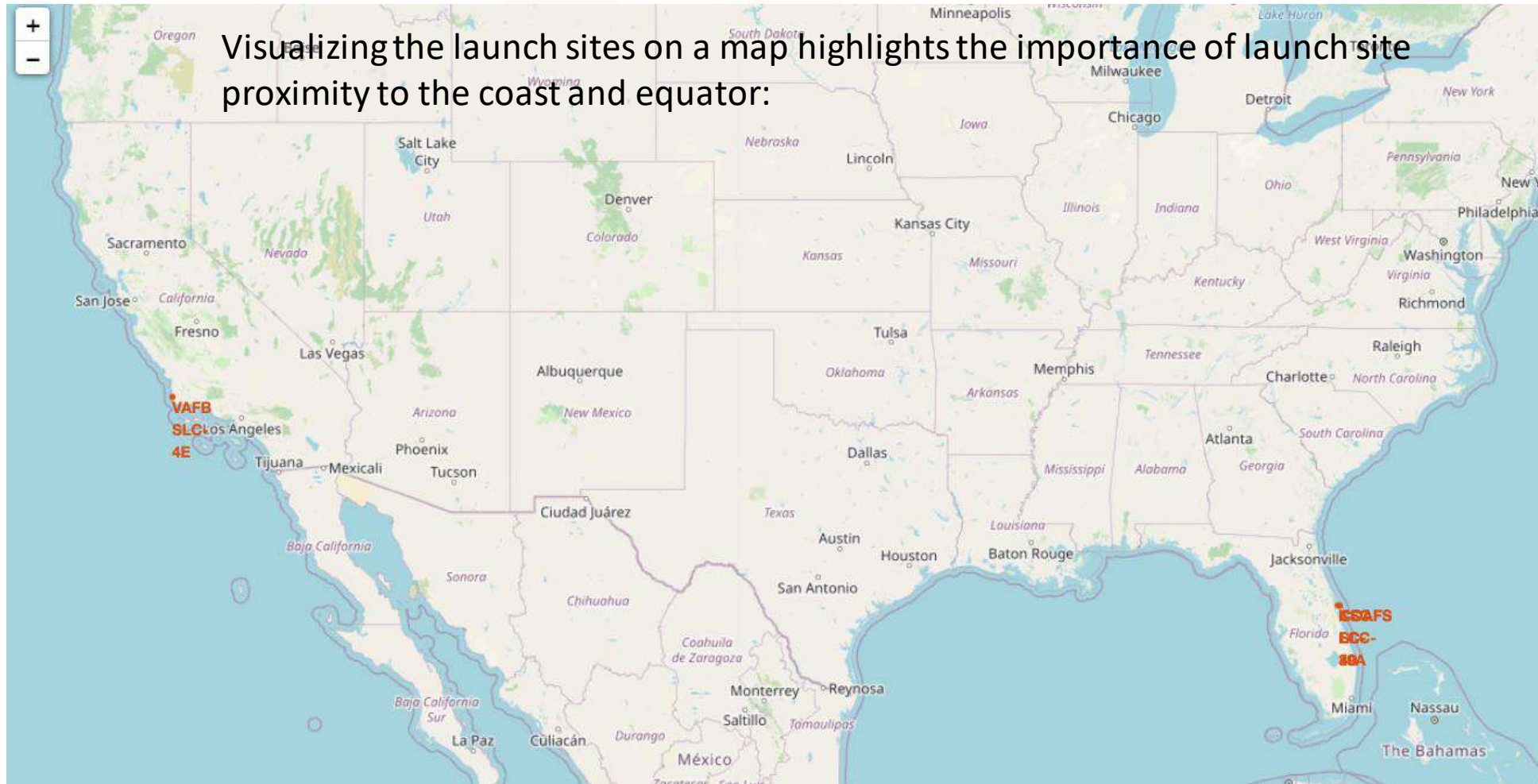
PayloadMass x Orbit type,
heavier payloads have a negative
influence on GTO orbits and positive
influence on ISS orbits

A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The image is a composite of a solid blue background on the left and a satellite photograph of Earth on the right. The Earth's surface is dark, with numerous bright yellow and orange lights representing cities and urban areas. The horizon of the Earth is visible as a thin, curved line separating the dark surface from the deep blue of space.

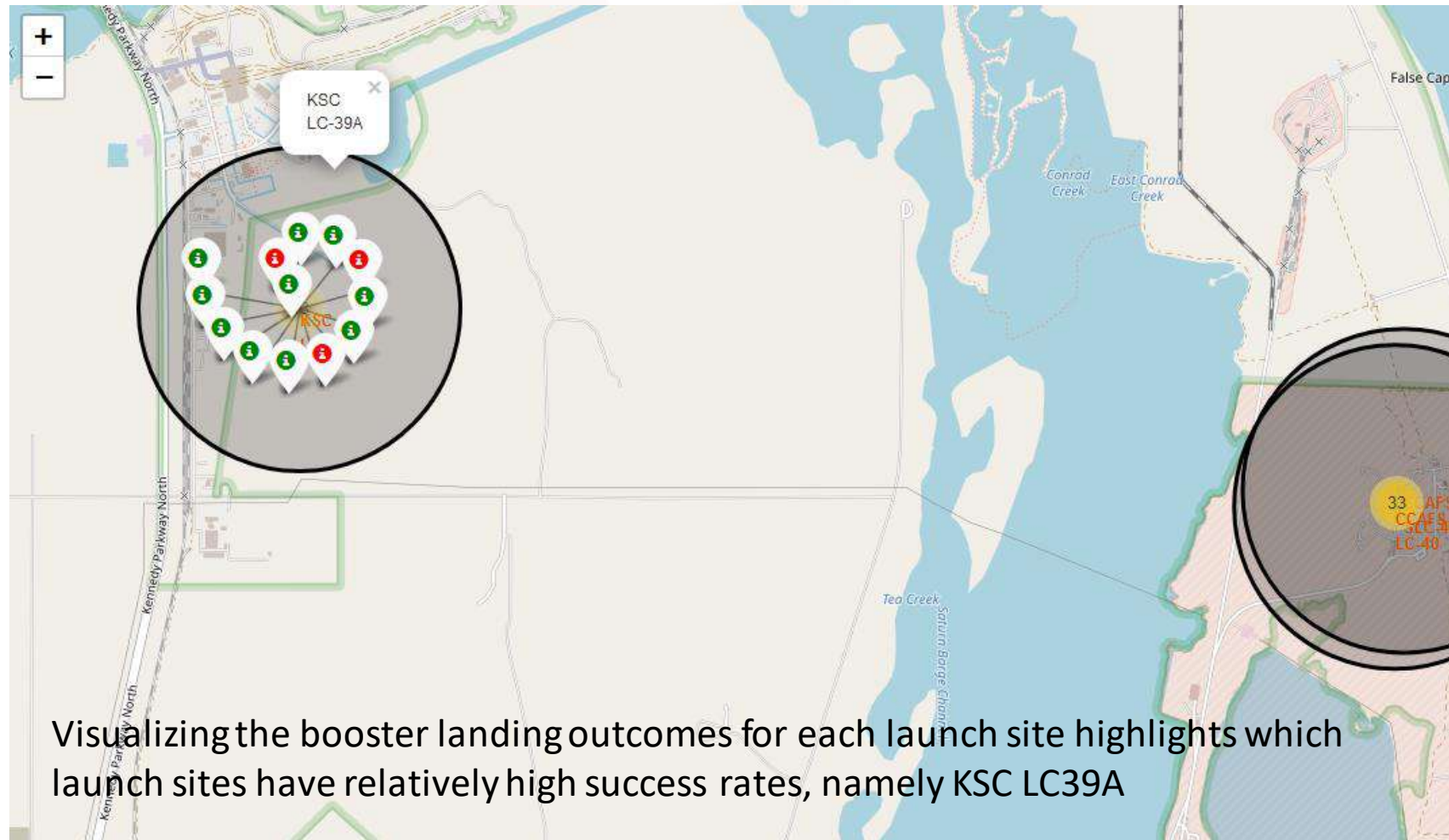
Section 4

Launch Sites Proximities Analysis

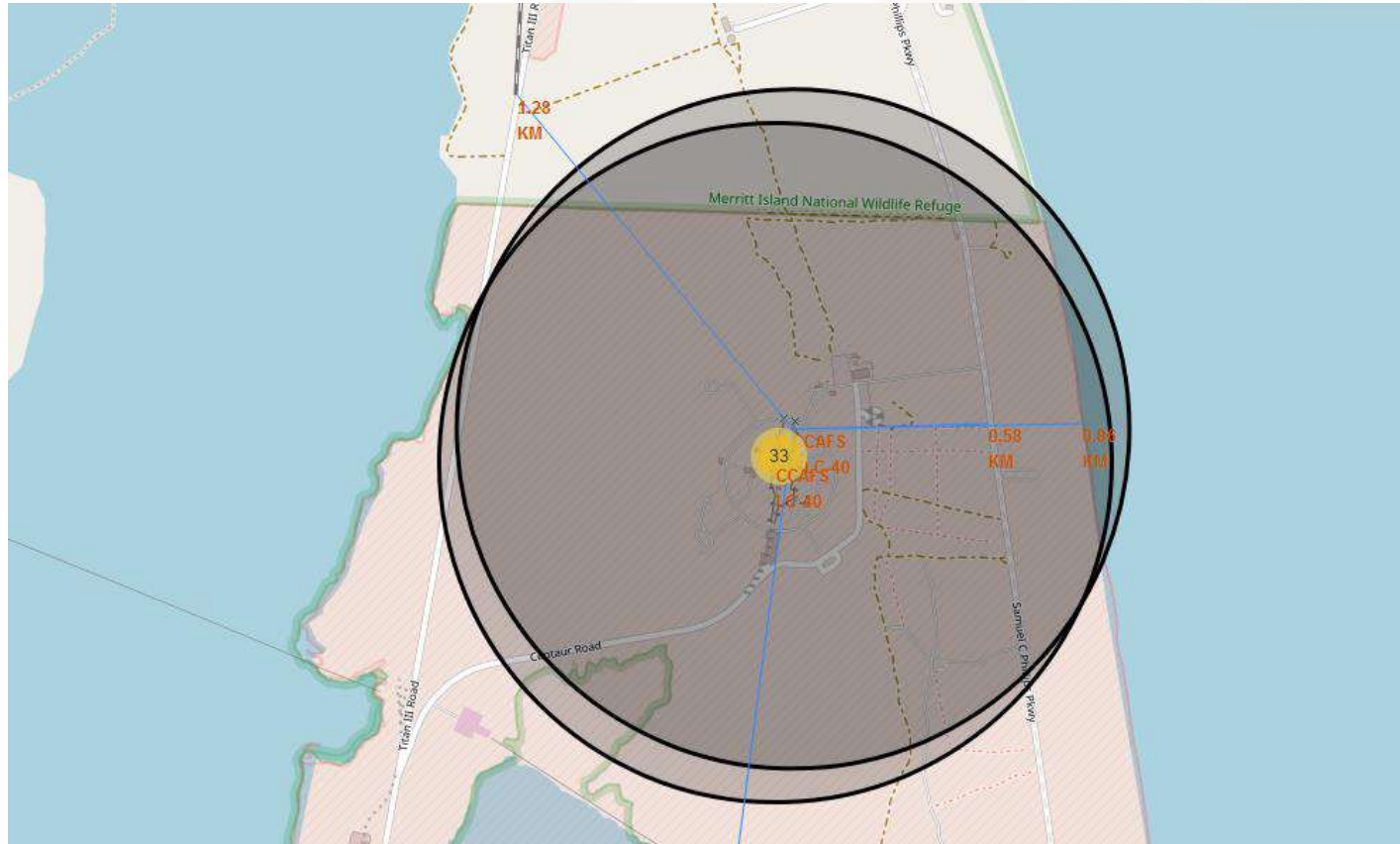
RESULTS: LAUNCH SITE LOCATION ANALYSIS



RESULTS: LAUNCH SITE LOCATION ANALYSIS (CONTINUED)



RESULTS: LAUNCH SITE LOCATION ANALYSIS (CONTINUED)



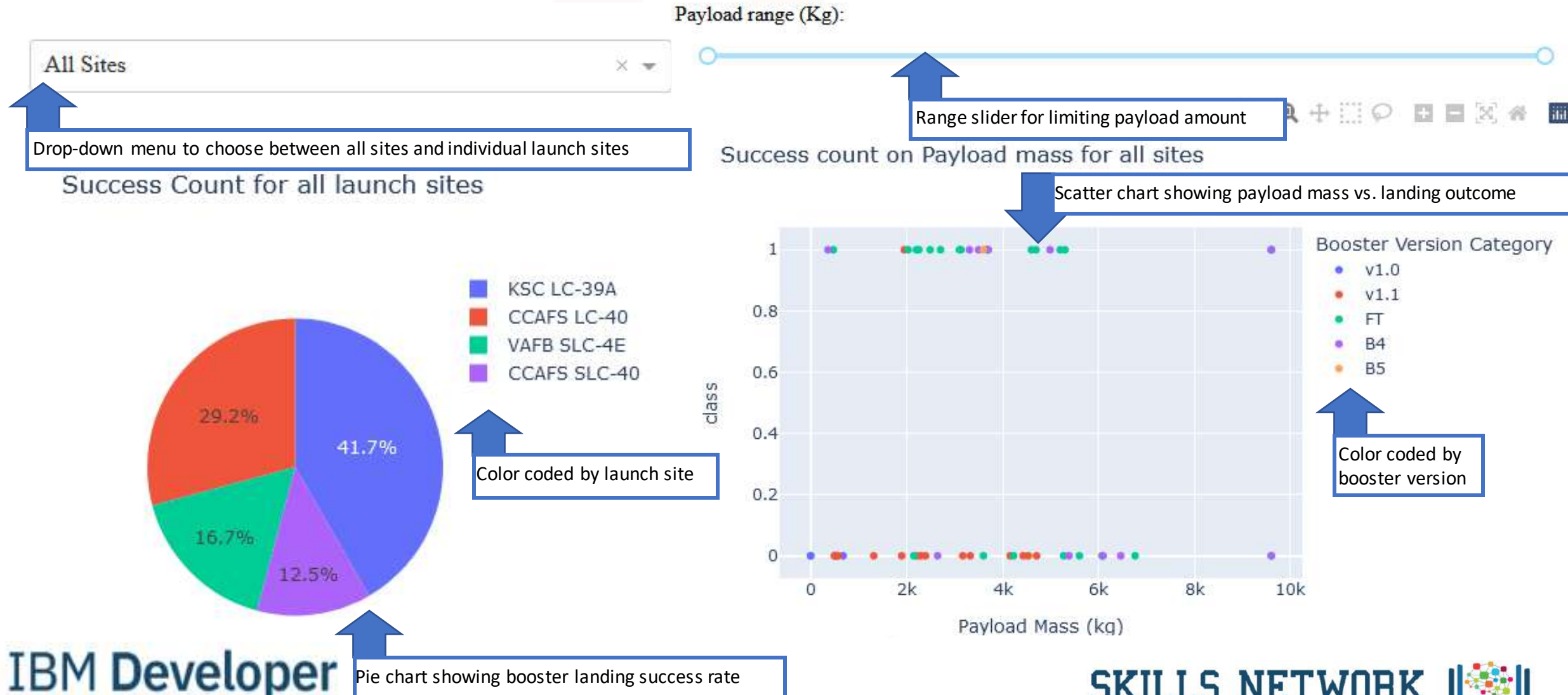
- Visualizing the railway, highway, coastline, and city proximities for each launch site allows us to see how close each is, for example:
- Proximities for CCAFS SLC-40:
 - railway: 1.28 km
 - transporting heavy cargo
 - highway: 0.58 km
 - transporting personnel and equipment
 - coastline: 0.86 km
 - optionality to abort launch and attempt water landing
 - minimizing risk from falling debris
 - city: 51.43 km
 - minimizing danger to population dense areas.



Section 5

Build a Dashboard with Plotly Dash

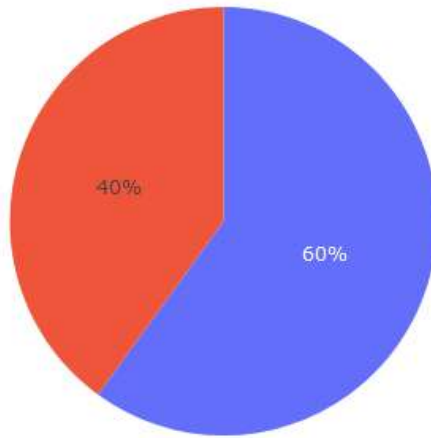
RESULTS: LAUNCH RECORDS DASHBOARD



RESULTS: LAUNCH RECORDS DASHBOARD (CONTINUED)

VAFB SLC-4E

Total Success Launches for site VAFB SLC-4E



■ 0
■ 1

Example dashboard view:

Booster landing success rate for VAFB SLC-4E

- Explore the dashboard yourself:

- <https://ibm-applied-data-science-capst.herokuapp.com/>
- Enabling stakeholders to explore and manipulate the data in an interactive and real-time way

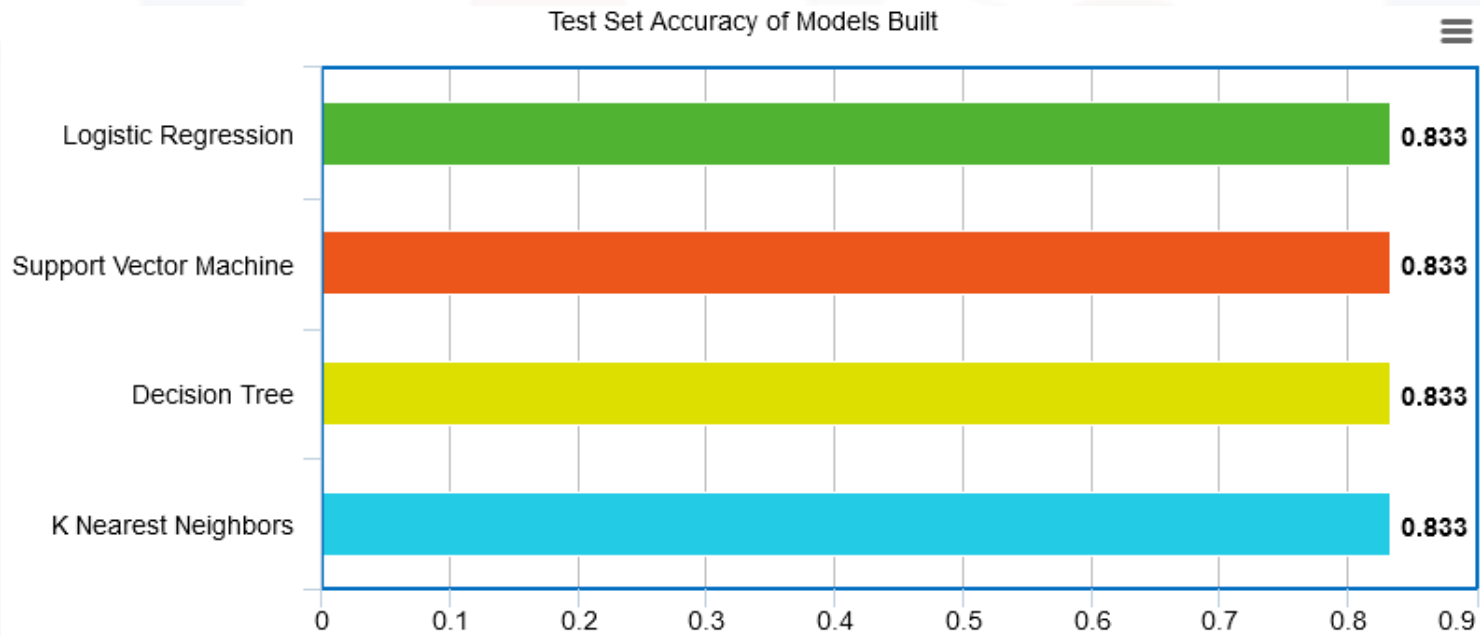
- Dashboard observations:

- FAFB SLC-4E had the heaviest successful booster landing success
- KSC LC-39A has the highest booster landing success rate
- Payloads < 5,300 kg had the highest booster landing success rate
- Payloads > 5,300 kg had the lowest booster landing success rate

Section 6

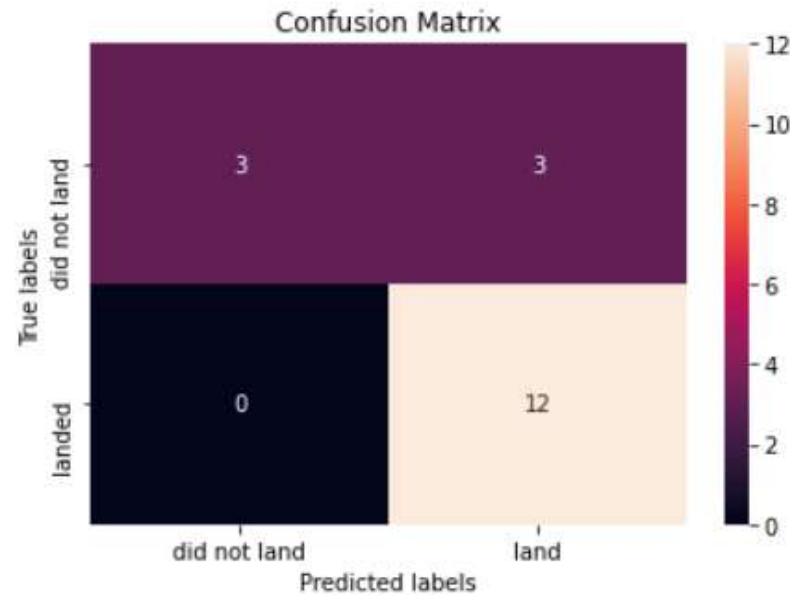
Predictive Analysis (Classification)

RESULTS: CLASSIFICATION ACCURACY



- Each of the four models built came back with the same accuracy score, 83.33%

RESULTS: CONFUSION MATRIX



- The confusion matrices of the best performing models (4-way-tie) are the same
- The major problem is false positives as evidenced by the models incorrectly predicting the 1st stage booster to land in 3 out of 18 samples in the test set

CONCLUSION

- Using the models from this report SpaceY can predict when SpaceX will successfully land the 1st stage booster with 83.3% accuracy
- SpaceX public statements indicate the 1st stage booster costs upwards of \$15 million to build
- This will enable SpaceY to make more informed bids against SpaceX, since they will have a good idea when to expect the SpaceX bid to include the cost of a sacrificed 1st stage booster
- With a list price of \$62 million per launch, sacrificing the \$15+ million 1st stage, would put the SpaceX bid at upwards of \$77 million
- Biggest opportunities going forward to make even more informed bids:
 - Freeze the best performing combination of model and hyperparameters and re-fit using the whole dataset instead of just the training data
 - Potentially better than using only part of the data to fit the model, but you would no longer be able to measure the accuracy of the resulting model
 - Incorporate additional launch data to the dataset and model as it becomes available
 - Subdivide the current model into two models
 - Predict if SpaceX will ATTEMPT to land the 1st stage
 - Predict if SpaceX will SUCCEED in their attempt
 - Create a related model that predicts if SpaceX will launch using a previously-flown 1st stage booster
 - Would enable SpaceY to take into account when the SpaceX bid would likely include a discount

APPENDIX

- Notebooks to recreate dataset, analysis, and models:
 - <https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/Hands-on%20Lab%20Complete%20the%20Data%20Collection%20API%20Lab.ipynb>
 - <https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/Hands-on%20Lab%20Data%20Collection%20with%20Web%20Scraping.ipynb>
 - <https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/Hands-On%20Lab%20Data%20Wrangling.ipynb>
 - <https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/Hands-on%20Lab%20Complete%20the%20EDA%20with%20SQL.ipynb>
 - <https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/jupyter-labs-eda-dataviz.ipynb>
 - https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/lab_jupyter_launch_site_location.ipynb
 - https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/spacex_dash_app.py
 - <https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/Hands-on%20Lab%20Build%20an%20Interactive%20Dashboard%20with%20Plotly%20Dash.ipynb>
 - https://github.com/brt-h/Applied-Data-Science-Capstone/blob/main/SpaceX_Machine%20Learning%20Prediction_Part_5.ipynb
- Acknowledgments
 - Thank you to Joseph Santarcangelo at IBM for creating the course and materials
 - Thank you to Lakshmi Holla at IBM for assisting me with questions and troubleshooting
- References
 - <https://aviationweek.com/defense-space/space/podcast-interview-spacexs-elon-musk>
 - Interview with Elon Musk where he discloses the 1st stage booster to cost upwards of \$15 million
 - <https://datascience.stackexchange.com/a/33050>
 - Explanation of why you would rebuild your model using the full dataset
 - <https://www.spacex.com/vehicles/falcon-9/>
 - Source of SpaceX's advertised \$62 million launch price

Thank you!

