



SpaceX Rocket Recapture Modeling

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OUTLINE



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 - Visualization – Charts
 - Dashboard
- Discussion
 - Findings & Implications
- Conclusion
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EXECUTIVE SUMMARY



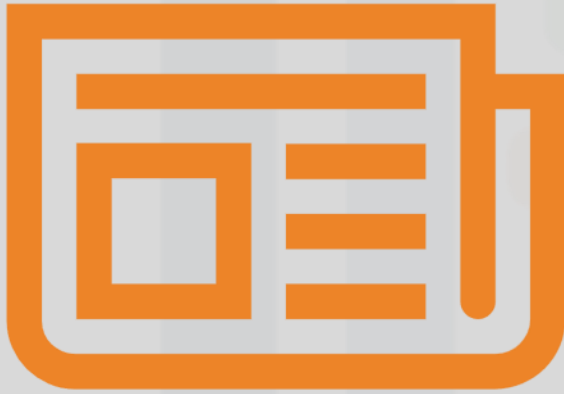
- We try to find features that predict the success, or failure, in landing the 1st stage of a Falcon 9 rocket.
- Captured relevant data for analysis.
 - Data collected from SpaceX API and Wikipedia.
 - Used SQL and Visualization to explore data.
 - Plotted geographic data to garner additional insight.
- Built dashboard to explore data granularly.
- Developed predictive models that accurately predict SpaceX Falcon 9 rocket landing failures, and (with less accuracy) successes.

INTRODUCTION



- We will use past data to predict whether the Falcon 9 rocket will land successfully.
- The Falcon 9 rocket has a stage that can be reused, upon successful landing.
- Reusing rocket parts allows for significant savings, a competitive advantage in the aeronautics industry.
- What factors are important to the successful recapture of stage 1 of a rocket?
 - Can these factors be used to successfully predict the success of rocket launches?

METHODOLOGY (Data Wrangling)



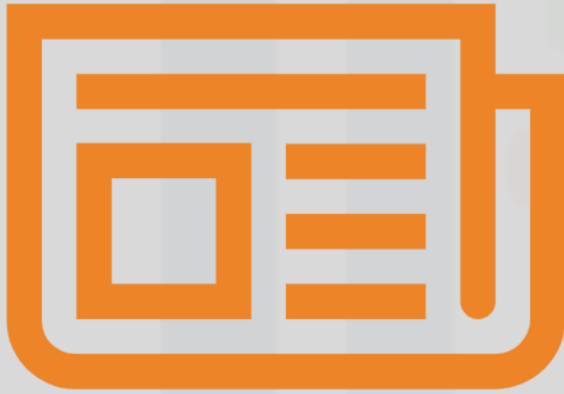
1. Import Libraries
2. Define Auxilliary Functions
3. Collect Data
4. Format Data

- Retrieved and parsed data from SpaceX API.
 - Imported libraries, defined helper functions, retrieved json response, parsed json into dataframe, dropped undesirable data, and imputed missing values.
- Scraped and parsed additional launch data from Wikipedia.
 - Imported libraries, defined helper functions, retrieved html data, parsed html data into dataframe columns.
- Wrangled data and uploaded files [here](#).
 - Imported libraries, defined auxillary functions, calculated launches by site/orbit, and developed training labels.

Using the `Outcome`, create a list where the element is zero if the corresponding row in `Outcome` is in the set `bad_outcome`; otherwise, it's one. Then assign it to the variable `landing_class`:

```
In [19]: # landing_class = 0 if bad_outcome
# landing_class = 1 otherwise
landing_class = []
for var in df['Outcome']:
    if var in bad_outcomes:
        landing_class.append(0)
    else:
        landing_class.append(1)
len(landing_class)
```

METHODOLOGY (EDA w/ SQL & Visuals)



- Developed SQL queries to explore data further.
- Used feature engineering to convert categorical features to numeric values & uploaded files [here](#).
- Displayed geographic data to identify insights not easily captured from tabular analyses.

Display the names of the unique launch sites in the space mission

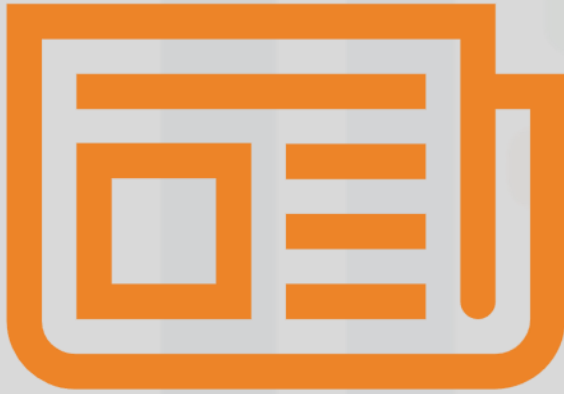
```
%sql select distinct "Launch_Site" from SPACEXTBL
```

```
* sqlite:///my_data1.db
```

Done.

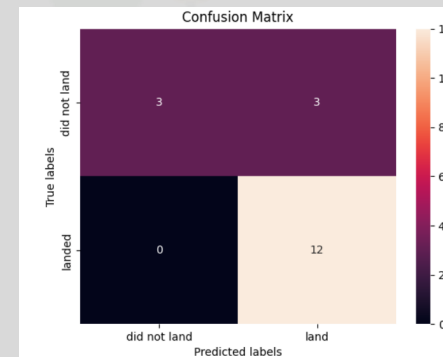
Launch_Site
CCAFS LC-40
VAFB SLC-4E
KSC LC-39A
CCAFS SLC-40

METHODOLOGY (Predictive Analysis)



1. Gather Model Input Data
2. Transform Data as Needed
3. Split Data into Test/Training Sets
4. Search for Optimal Parameters

- Standardized numeric values for consistency.
- Transformed data into format for model input.
- Split data into training and test sets.
- Used GridSearchCV to identify parameters that improve model performance.
 - Compared logistic regression, support vector machine, decision tree classifier, and k-nearest neighbors.
 - Calculated accuracy scores (R2) and plotted confusion matrices.



RESULTS



EDA WITH VISUALIZATION - FINDINGS & IMPLICATIONS

Findings

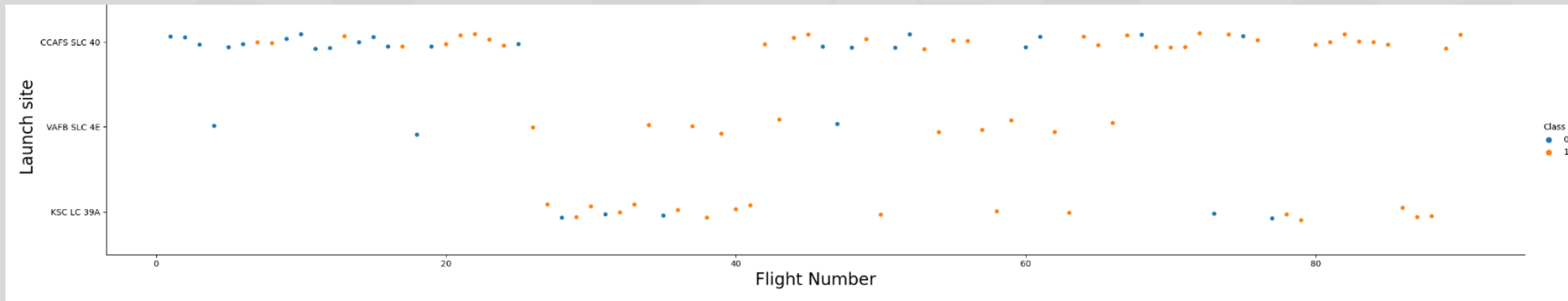
- Success varied with orbit
- Later launches showed greater success
- Launch sites varied in the payloads of rockets launched

Implications

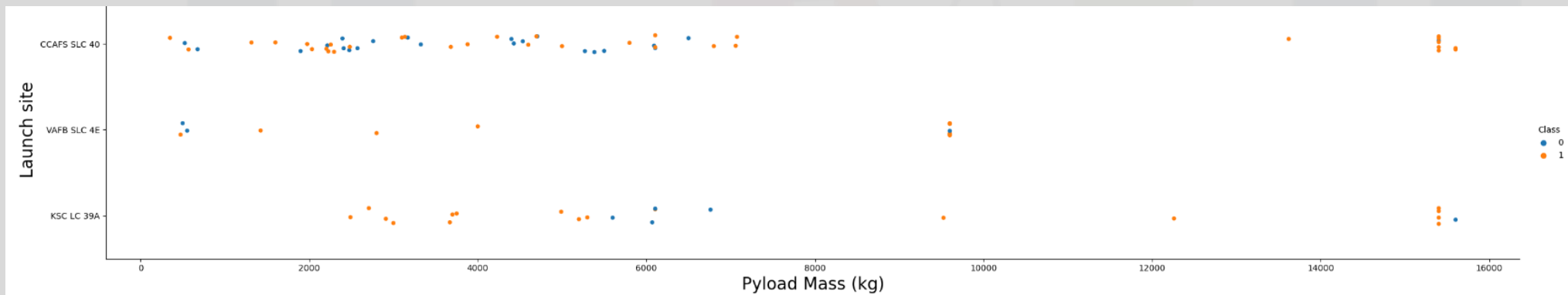
- Some orbits inherently difficult
- Over time, effective improvements were made
- Some sites are better suited for launches of large rockets

EDA VISUALIZATIONS

Flight Number and Launch Site Scatter

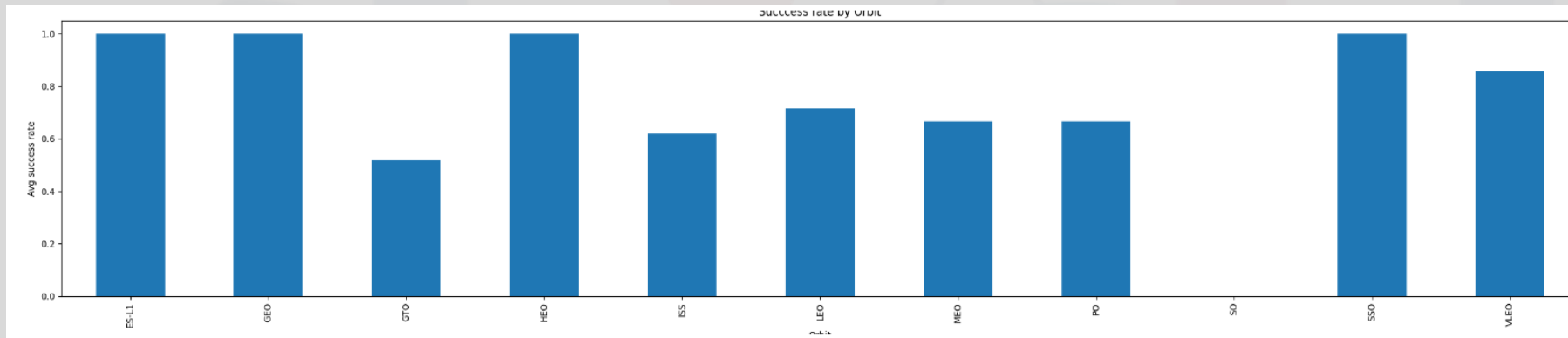


Payload and Launch Site Scatter

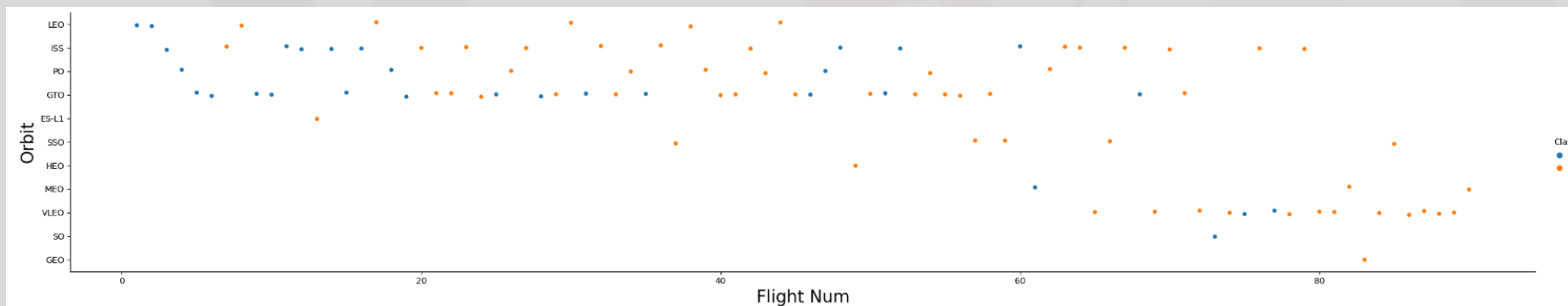


EDA VISUALIZATIONS

Success rate vs Orbit bar chart

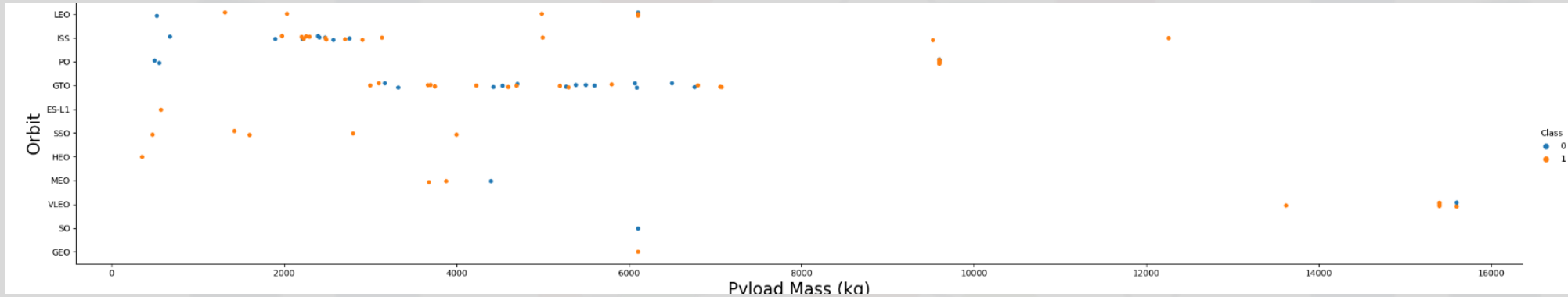


Flight Number vs Orbit Scatter

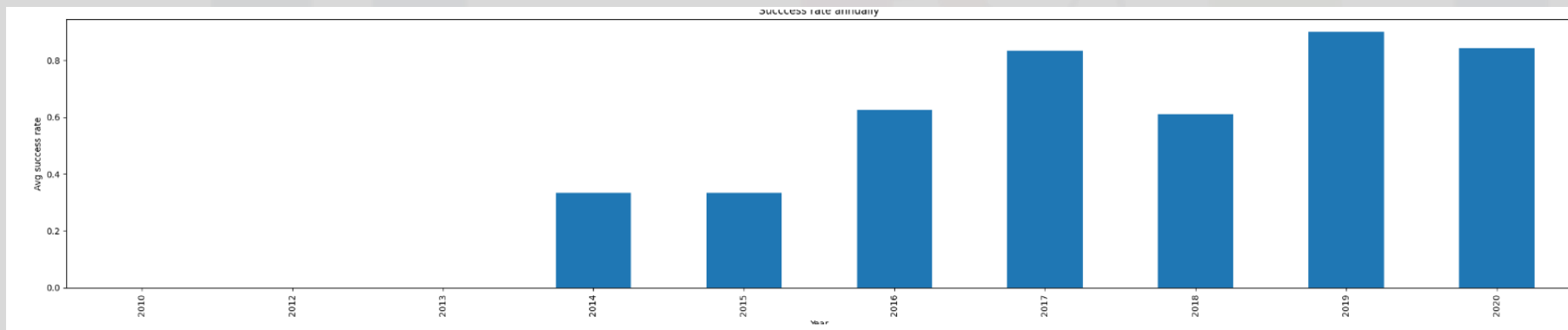


EDA VISUALIZATIONS

Payload vs Orbit scatter



Launch Success Yearly Trend



EDA WITH SQL - FINDINGS & IMPLICATIONS

Findings

- 4 launch sites, though some are similarly named
- Calculated average mass of payload carried by booster
- Calculated number of success and failure outcomes

Implications

- Different sites used for different launch types
- Payload may be correlated with other features (e.g. orbit)
- Variations in data can illuminate impactful features

SQL EXPLORATORY DATA ANALYSIS

Find distinct launch sites

Display the names of the unique launch sites in the space mission

```
%sql select distinct "Launch_Site" from SPACEXTBL
```

```
* sqlite:///my_data1.db  
Done.
```

Launch_Site

CCAFS LC-40

VAFB SLC-4E

KSC LC-39A

CCAFS SLC-40

5 sites that begin with 'CCA'

Display 5 records where launch sites begin with the string 'CCA'

```
%sql select * from SPACEXTBL where "Launch_Site" like "CCA%" limit 5
```

```
* sqlite:///my_data1.db  
Done.
```

Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG_	Orbit	Customer	Mission_Outcome	Landing_Outcome
04-06-2010	18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
08-12-2010	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)
22-05-2012	07:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
08-10-2012	00:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
01-03-2013	15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

SQL EXPLORATORY DATA ANALYSIS

Total Payload Mass

Display the total payload mass carried by boosters launched by NASA (CRS)

```
%sql select sum("PAYLOAD_MASS_KG_") from SPACEXTBL where "Customer" = "NASA (CRS)"
```

```
* sqlite:///my_data1.db  
Done.
```

sum("PAYLOAD_MASS_KG_")

45596

Average Payload Mass by F9 v1.1

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

```
%sql select sum("PAYLOAD_MASS_KG_")/count("PAYLOAD_MASS_KG_") as avgMassf9 from SPACEXTBL where "Booster_Version" = "F9 v1.1"
```

```
* sqlite:///my_data1.db  
Done.
```

avgMassf9

2928

SQL EXPLORATORY DATA ANALYSIS

First Successful Ground Pad Landing

Task 5

List the date when the first succesful landing outcome in ground pad was acheived.

Hint: Use min function

```
%sql select min("Date") from SPACEXTBL where "Mission_Outcome" = "Success" and lower("Landing_Outcome") like "%ground pad%"
```

```
* sqlite:///my_data1.db
```

Done.

min("Date")

01-05-2017

Boosters Landed on Drone Ships with Payload From 4-6k

List the names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000

```
%sql select distinct "Booster_Version" from SPACEXTBL where "Mission_Outcome" = "Success" and lower("Landing_Outcome") like "%drone ship%" and "PAYLOAD_MASS_KG_" between 4000 and 6000
```

```
* sqlite:///my_data1.db
```

Done.

Booster_Version

F9 FT B1020

F9 FT B1022

F9 FT B1026

F9 FT B1021.2

F9 FT B1031.2

SQL EXPLORATORY DATA ANALYSIS

Total Number of Success/Failure Outcomes

List the total number of successful and failure mission outcomes

```
%sql select "Mission_Outcome", count(*) from SPACEXTBL group by "Mission_Outcome"
```

```
* sqlite:///my_data1.db
```

Done.

Mission_Outcome	count(*)
Failure (in flight)	1
Success	98
Success	1
Success (payload status unclear)	1

Which Boosters Carried Maximum Payloads

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

```
%sql select "Booster_Version" from SPACEXTBL where "PAYLOAD_MASS_KG" = (select max("PAYLOAD_MASS_KG") from SPACEXTBL)
```

```
* 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

SQL EXPLORATORY DATA ANALYSIS

2015 Launch Records

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.

Note: SQLite does not support monthnames. So you need to use substr(Date, 4, 2) as month to get the months and substr(Date,7,4)='2015' for year.

```
%sql select substr("Date",4,2) as month, "Landing _Outcome", "Booster_Version", "Launch_Site" from SPACEXTBL where substr("Date",7,4)="2015" and lower("Landing _Outcome") like "%Failure%drone"
```

```
* sqlite:///my_data1.db
```

```
Done.
```

month	Landing _Outcome	Booster_Version	Launch_Site
01	Failure (drone ship)	F9 v1.1 B1012	CCAFS LC-40
04	Failure (drone ship)	F9 v1.1 B1015	CCAFS LC-40

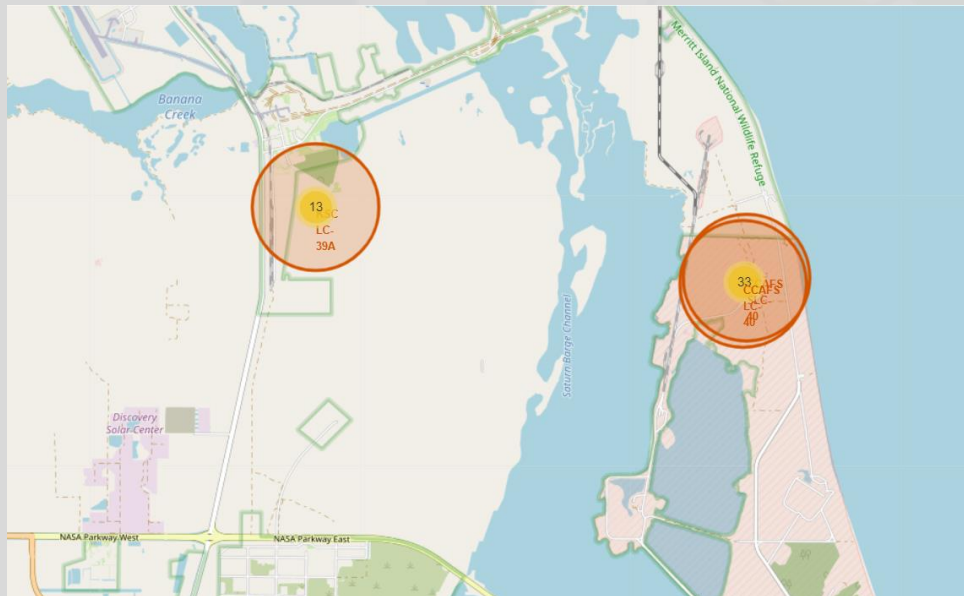
Rank Success Count Between Dates

Rank the count of successful landing_outcomes between the date 04-06-2010 and 20-03-2017 in descending order.

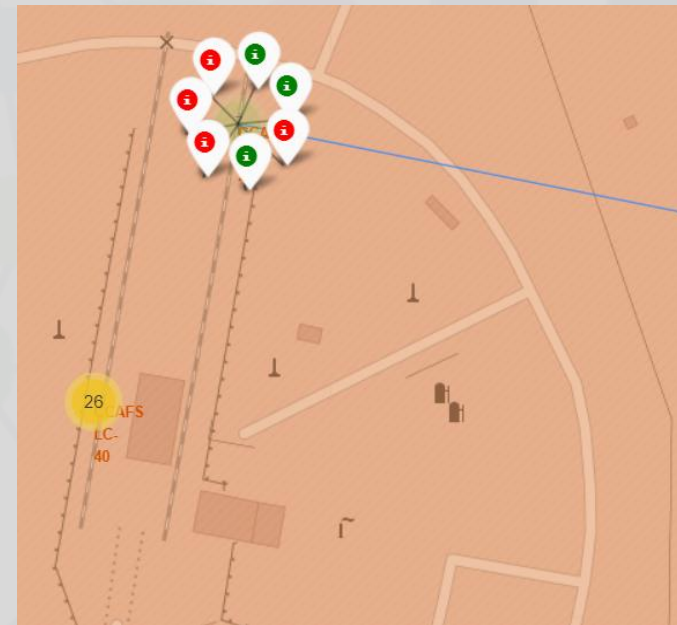
```
%sql select substr("Date",4,2) as month, "Landing _Outcome", "Booster_Version", "Launch_Site" from SPACEXTBL where "Date" > '04-06-2010' and "Date" < '20-03-2017'
```

FOLIUM RESULTS

All launch records are associated with a site

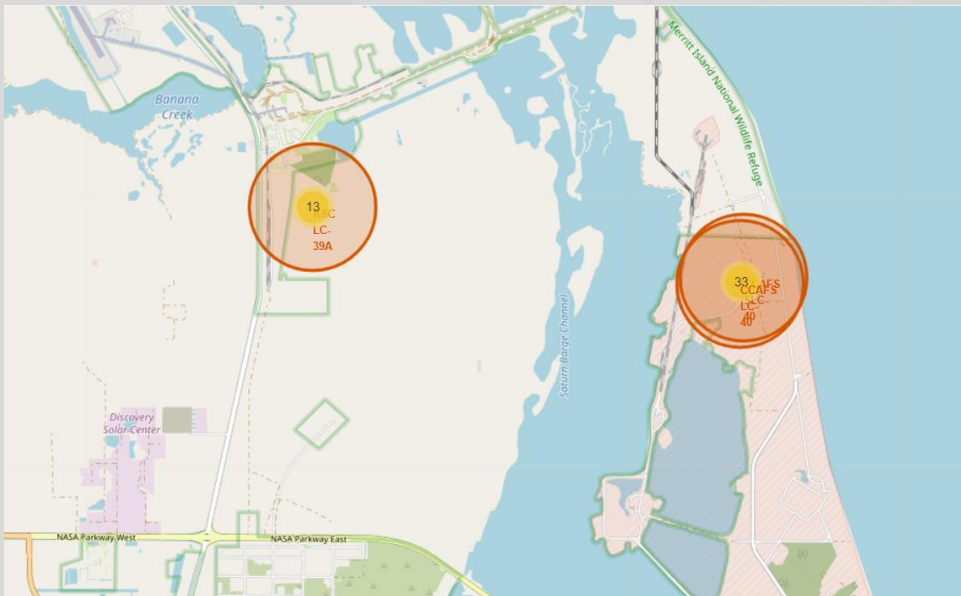


Markers show multiple launches per each site



FOLIUM RESULTS

Success rates vary by launch site



Launch sites close to coasts, highways, and rail



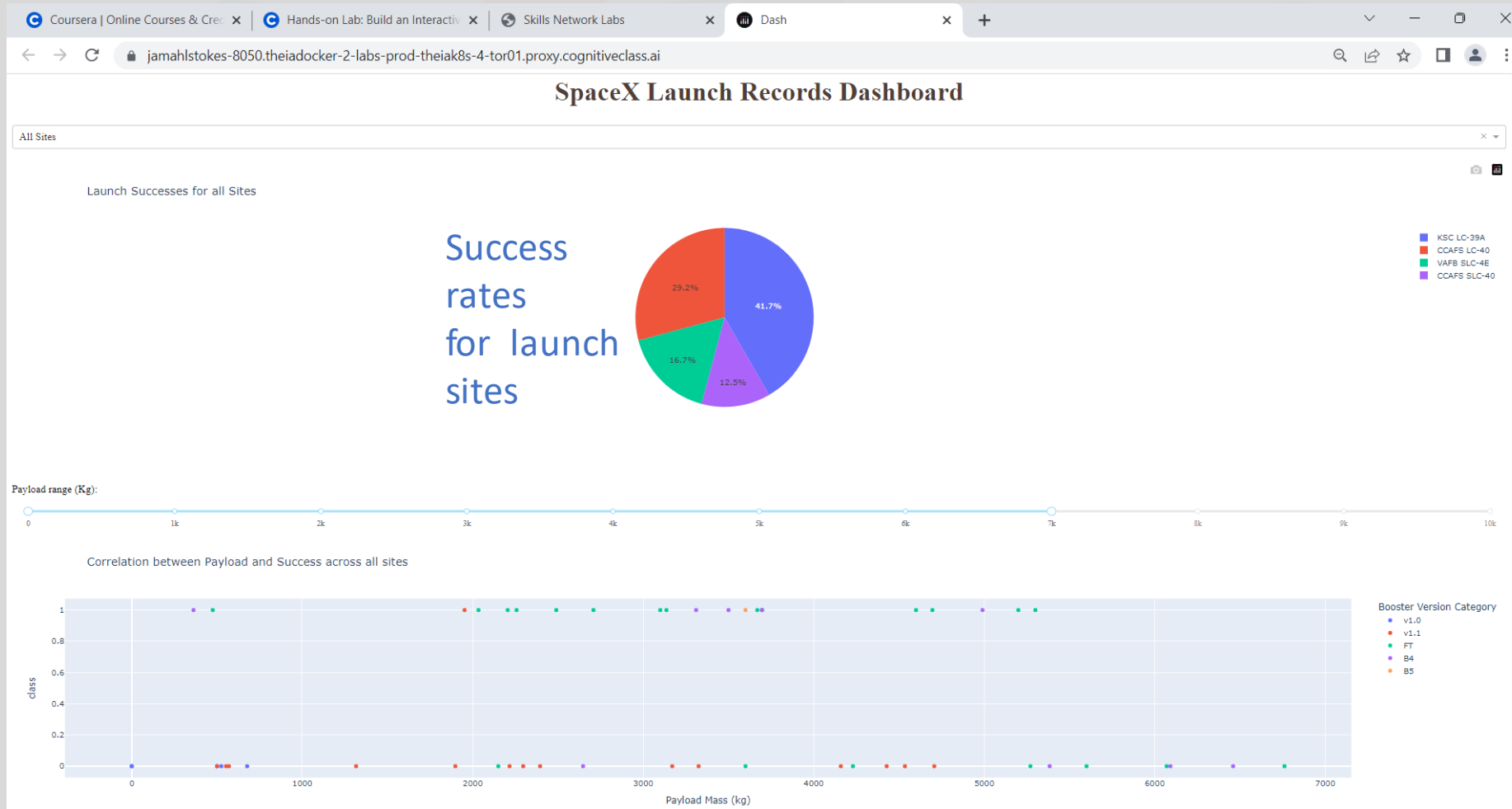
DASHBOARD



The dashboard uploaded [here](#) was developed to examine the success rates across launch sites and to explore how launch sites varied in the payload mass that rockets carried.

Results showed that payload mass and launch site were correlated with launch and re-landing success.

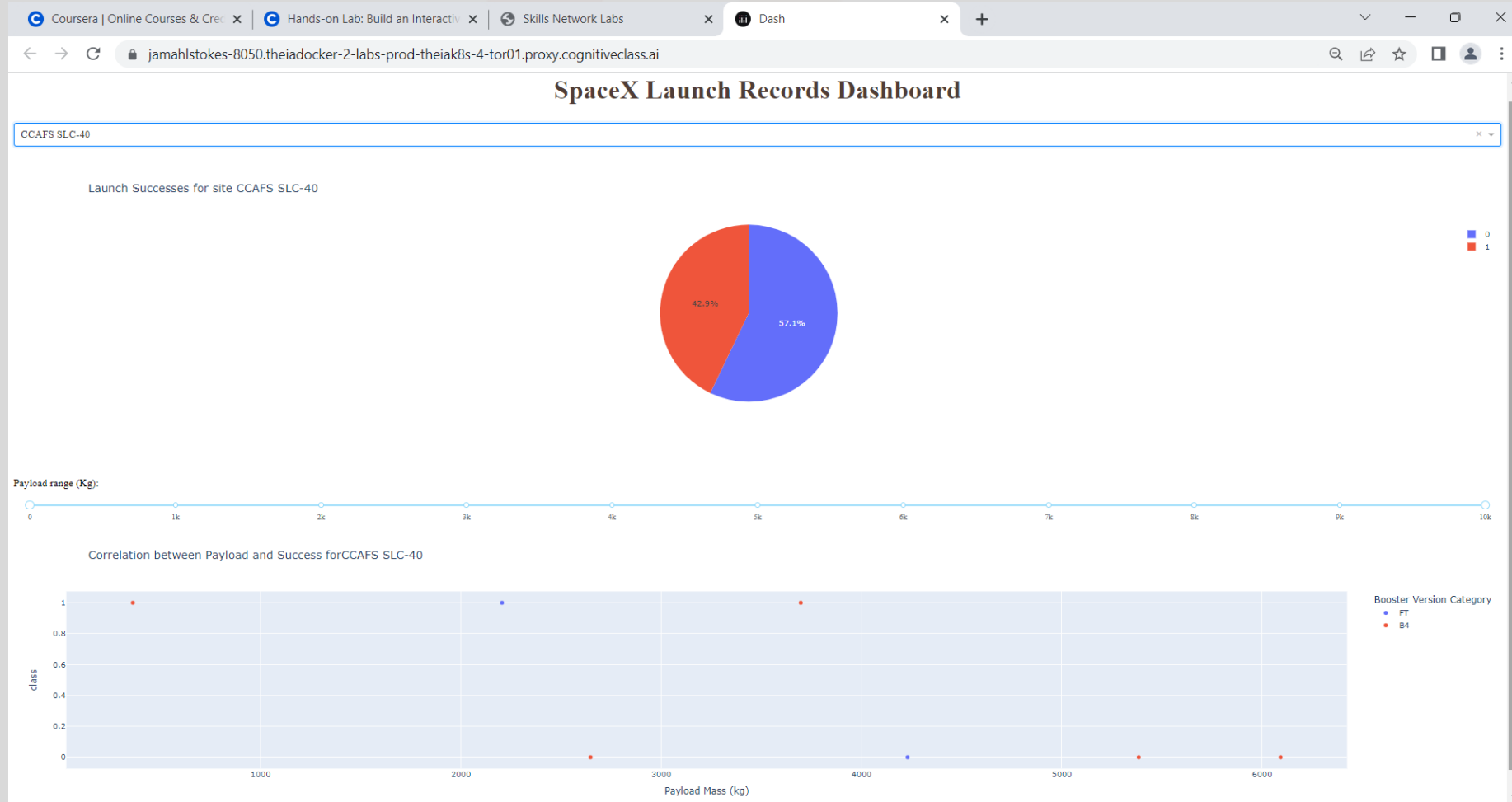
DASHBOARD – Dropdown & Slider



Interactive
Slider that
updates
Scatter Plot

DASHBOARD

Dashboard
filters to
specific site
upon
selection



PREDICTIVE ANALYSIS

- Various classification models were evaluated on training data and had accuracy scores as follows.

Logistic Regression	.846
Support Vector Machine	.848
Decision Tree Classification	.888
K-Nearest Neighbors	.848

PREDICTIVE ANALYSIS

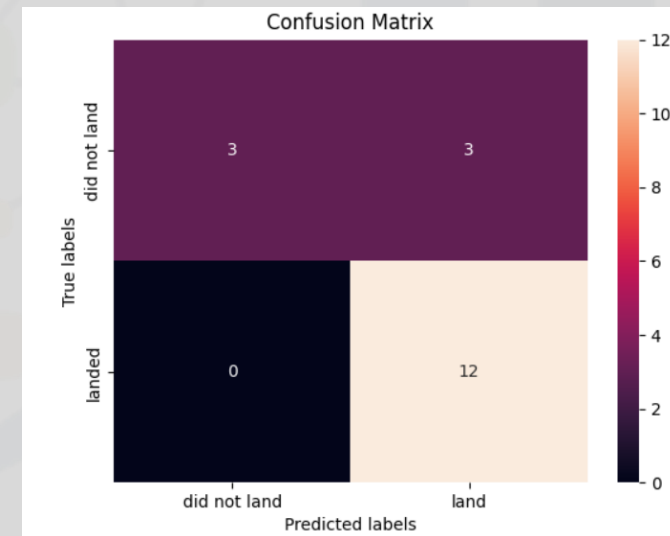
- When these models were evaluated on test data their accuracy scored as follows.



Logistic Regression	.833
Support Vector Machine	.833
Decision Tree Classification	.666
K-Nearest Neighbors	.833

PREDICTIVE ANALYSIS

- SVM, and KNN show equal performance here, superior to that of decision tree and logistic regression.



CONCLUSION

Overall Findings

- Geography may influence selection of launch sites
- Sites varied in the profiles of payloads they launched
- Several booster versions carry the maximum payload
- Predictive models accurately classified failures, with lower scores classifying successes

CONCLUSION



- Re-capture of stage 1 of rockets saves significant costs.
- Factors such as mission orbit, launch site, launch date, and payload impact success rates.
- Predictive models, informed by past launches, can be developed to help forecast future launch successes.

APPENDIX



SUPPLEMENTAL FILE LOCATIONS

Supporting files for all analyses presented have been uploaded to my personal Github account <<https://github.com/jamahlstokes/IBM-DataScience-Public>>.

Enjoy!