



IBM Developer
SKILLS NETWORK

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 and wrangling using SpaceX API and Wikipedia Web Scrapping
 - Exploratory Data analysis
 - Interactive visualization (map with Folium and dashboard with Plotly Dash)
 - Predictive analysis (classification models)
- Summary of all results
 - Exploratory data analysis results - [Section 2](#)
 - Interactive analytics demo in screenshots - [Section 3](#)
 - Predictive analysis results - [Section 4](#)

Introduction

- **Context**

As a new provider in Space Transportation, we would like to beat SpaceX.

But how can we win the competition with the cheapest provider in the market (on average 3 times cheaper than others)?

SpaceX saved that much money due to reuse of the first stage of the rocket.

- **Goal**

Predict if the Falcon 9 first stage will land successfully and understand what are key factors of success

Section 1

Methodology

Methodology

Executive Summary

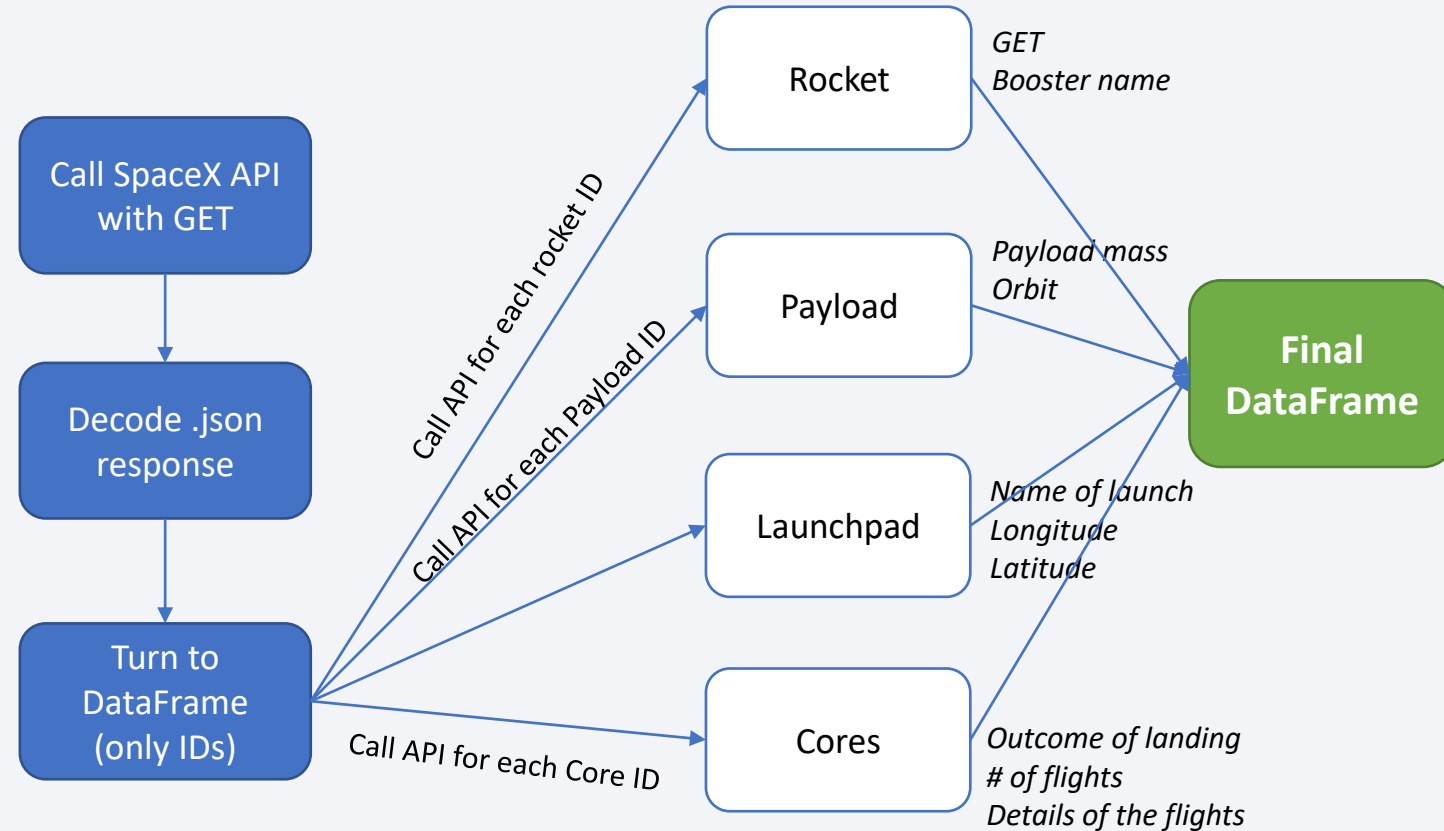
- Data collection methodology:
 - SpaceX API
 - [Wikipedia page](#) Web Scraping
- Perform data wrangling
 - NAs eliminated, Data types check, Dependent variable (Y) created
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - How to build, tune, evaluate classification models

Data Collection

- Data for this project were collected from 2 sources:
 - Official [SpaceX REST API](#). Please see [GitHub page](#) for more details
 - [Wikipedia page](#)

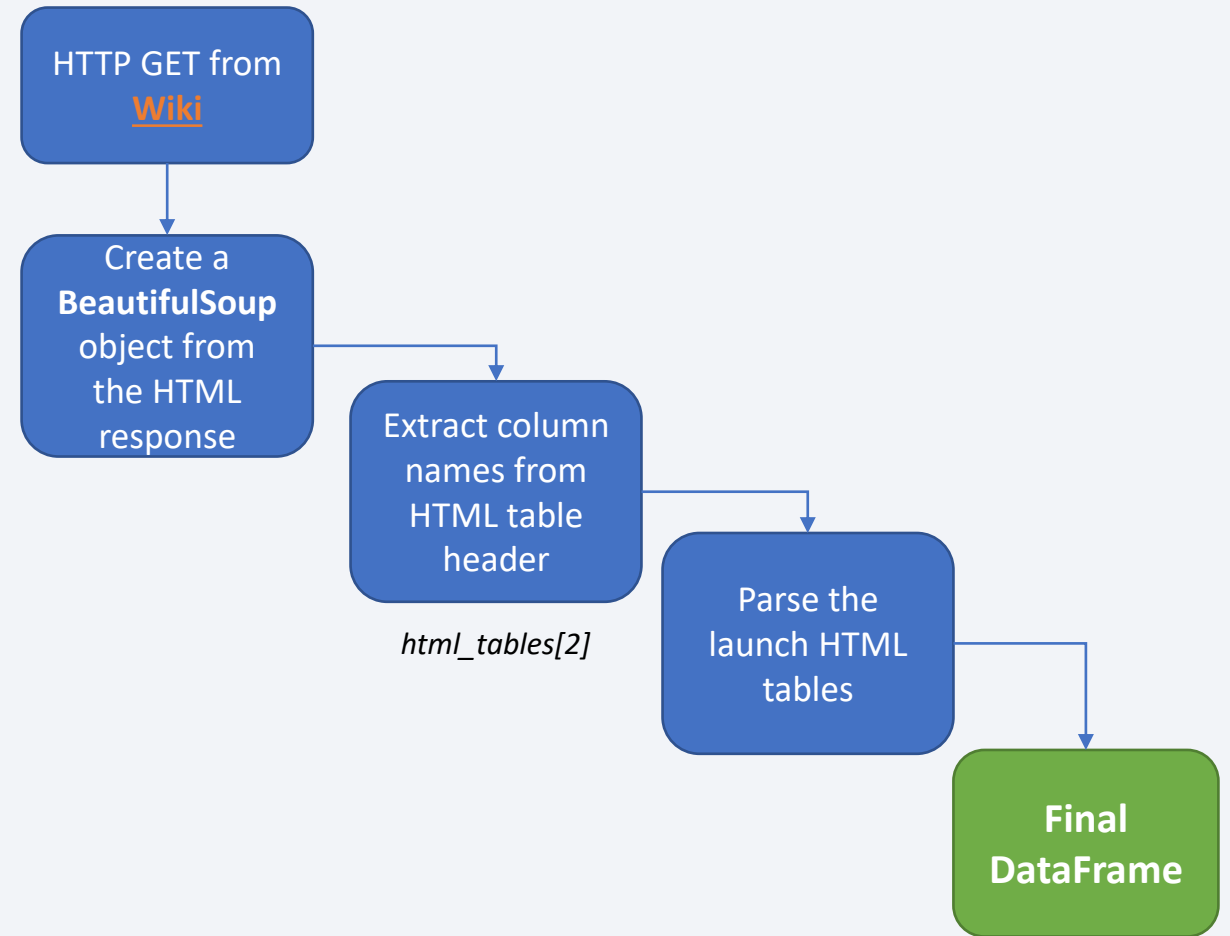
Data Collection – SpaceX API

- To collect data from SpaceX API we first collect the roster of all launches, but only with IDs in columns
- Next we restore all missing descriptive columns by calling API for each ID (example, `requests.get("https://api.spacexdata.com/v4/rockets/rocket_id).json()`)
- [GitHub URL](#)



Data Collection - Scrapping

- Web Scrapping HTML table from [Wiki page](#) to get data about Falcon 9 and Falcon Heavy launches
- Parse HTML tables and convert it to the DataFrame
- [GitHub URL](#)



Data Wrangling

- After getting raw data it should be preprocessed for exploratory data analysis and modeling in future.
- Key procedures applied:
 - NULLs check
 - Data types check
 - # records per launch site, orbit and landing outcome
 - Y variable/Class created to determine Success/Failure of launch (dependent variable)
- [GitHub URL](#)

EDA with Data Visualization

- After Data Wrangling stage we're ready to play with data.
- There are different questions we wanted to answer with this step:
 - Relationship between Payload Mass and FlightNumber (by launch sites) to see if anything changes with more tests and technology development
 - Relationship between FlightNumber and Launch site
 - Relationship between PayloadMass and Launch site to understand if there are difference success rate between sites with different Payload Mass
 - Success rate by Orbit type to see how far into space we can launch the rocket successfully
 - Relationship between PayloadMass and Orbit type to see how far into space we can launch the rocket with a heavy mass successfully
 - Yearly trend of launch success to see the progress in time

- [GitHub URL](#)

EDA with SQL

To check the data structure and aggregated statistics we used the following SQL queries:

- Display the names of the unique launch sites in the space mission
- Display 5 records where launch sites begin with the string 'KSC'
- Display the total payload mass carried by boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1
- List the date where the first successful landing outcome in drone ship was achieved
- List the names of the boosters which have success in ground pad and have payload mass greater than 4000 but less than 6000
- List the total number of successful and failure mission outcomes
- List the names of the booster versions which have carried the maximum payload mass
- List the records which will display the month names, successful landing outcomes in ground pad ,booster versions, launch site for the months in year 2017
- Rank the count of successful landing outcomes between the date 2010-06-04 and 2017-03-20 in descending order.

[GitHub URL](#)

Build an Interactive Map with Folium

- The launch success rate may depend on many factors such as payload mass, orbit type, and so on. It may also depend on the location and proximities of a launch site, i.e., the initial position of rocket trajectories. Finding an optimal location for building a launch site certainly involves many factors which we are going to find by analyzing the existing launch site locations.
- Folium will help us create a Map where we mark:
 - All launch sites (Circle and Marker with name to easily identify these locations)
 - Success/failed launches for each site as Cluster Marker to visually assess success rate
 - Distances between a launch site to its proximities (closest city, highway, railway, coast) as line with marker (km) to understand strategic positions of these sites
- [GitHub URL](#)

Build a Dashboard with Plotly Dash

- Building an interactive dashboard and charts is a very convenient and easy way to present data to potential stakeholders.
- With Plotly Dash we showed:
 - Pie chart of success launches by sites together with success rate for each site to get a quick answer about most/least successful site
 - Scatter plot of launch outcome by Payload Mass and Booster version to visualize if launch outcome depends on there 2 variables

[GitHub URL](#)

Predictive Analysis (Classification)

The following steps were performed to do a high-quality predictive analysis:

1. Create a column for the Launch Outcome (Y): Success (1) / Failure (0)
2. Standardize the data using **StandardScaler()**
3. Split into training data and test data
4. Find best Hyperparameter for SVM, Classification Trees and Logistic Regression using **GridSearchCV** using **train** data
5. Find the method performs best using **test** data and calculating **Score** and **Confusion Matrix**

[GitHub URL](#)

Results

- Exploratory data analysis results - [Section 2](#)
- Interactive analytics demo in screenshots - [Section 3](#)
- Predictive analysis results - [Section 4](#)

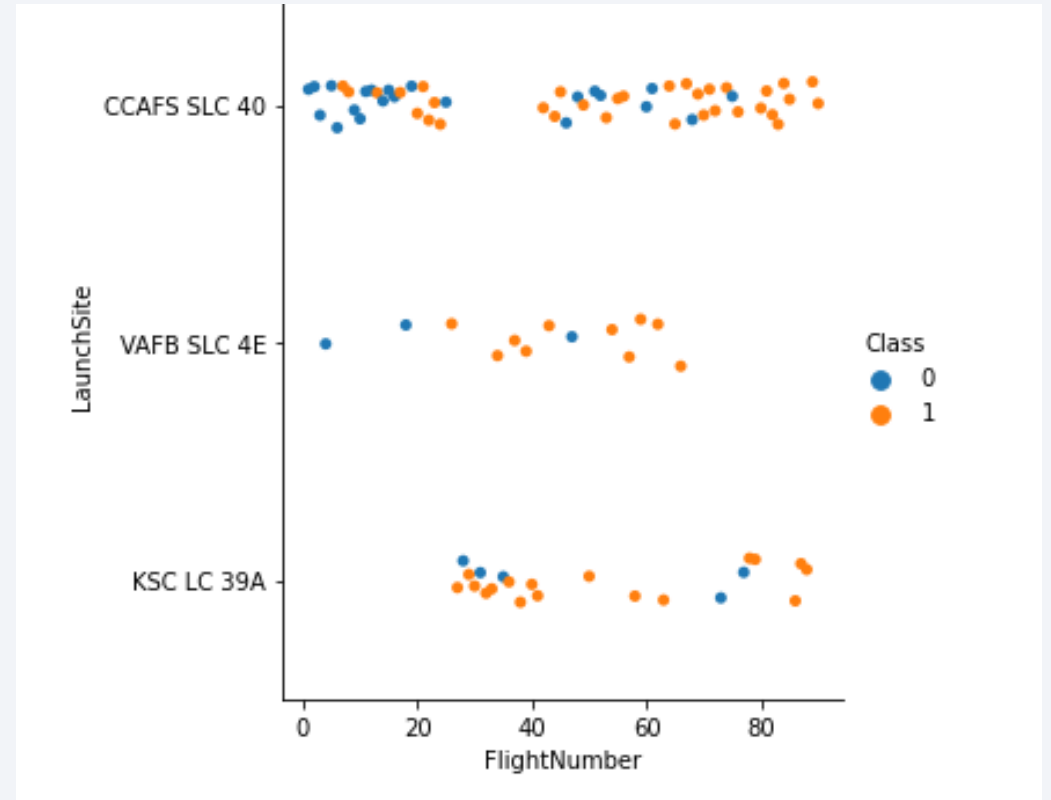
The background of the slide is an abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks in shades of red and cyan. A faint, light blue grid pattern is also visible, particularly in the lower half of the image. The overall effect is dynamic and technological.

Section 2

Insights drawn from EDA

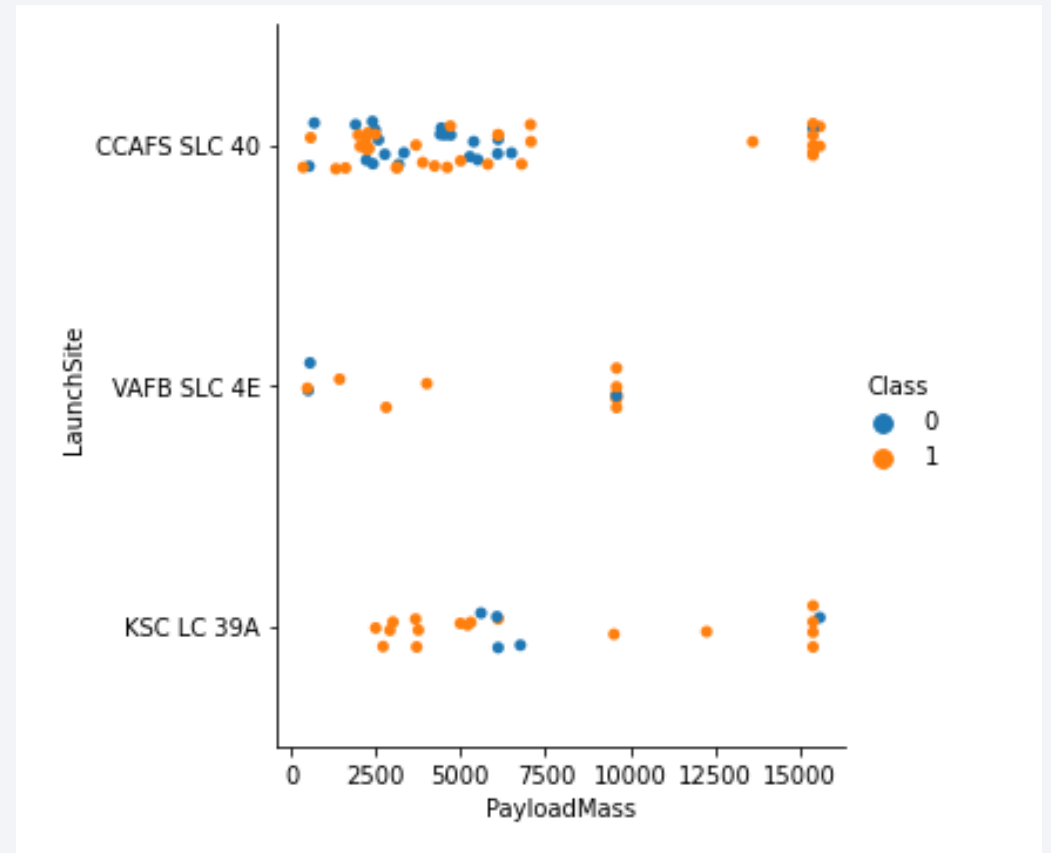
Flight Number vs. Launch Site

- First 20 attempts made mainly from Cape Canaveral were mostly not successful (try and error)
- But after 21-25th launch we can observe quite high success rate in all sites



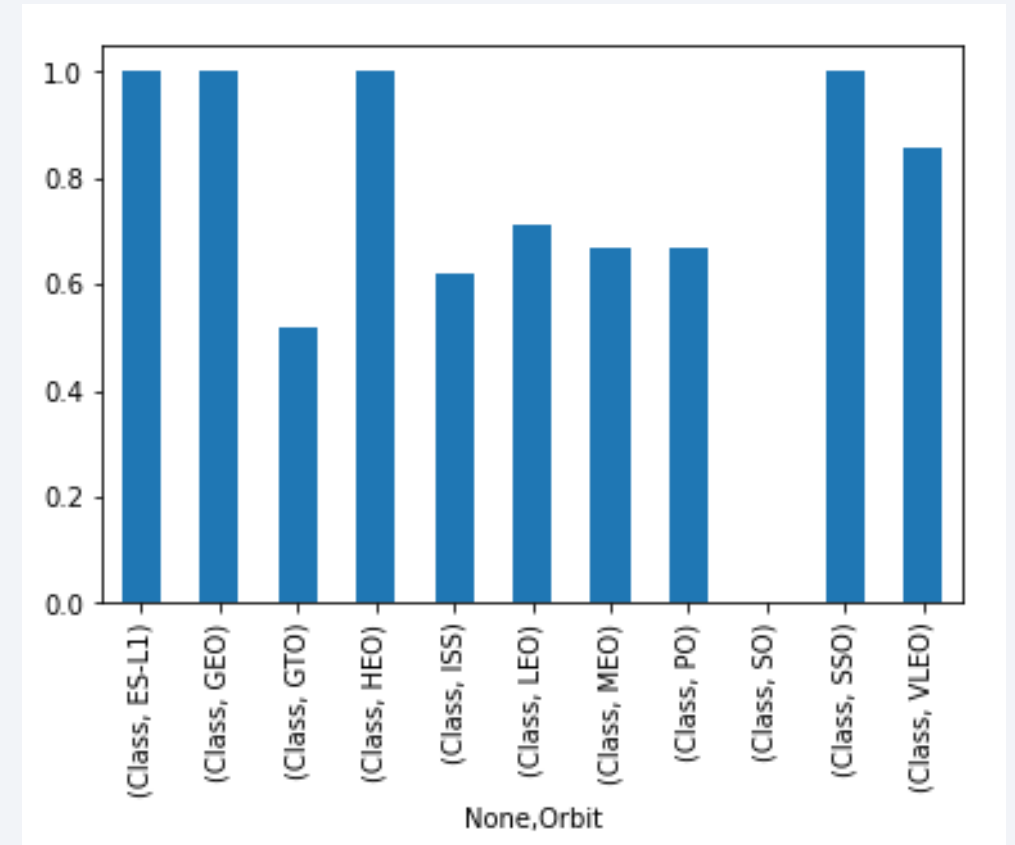
Payload vs. Launch Site

- CCAFS site doesn't show good stable result in payload range 0-7,500kg. However, it succeeds with high loads > 12,500 kg
- Meanwhile, KSC LC showed 100% success rate for lower payloads (2,500-5,000 kg) as well as high ones (>10,000 kg)



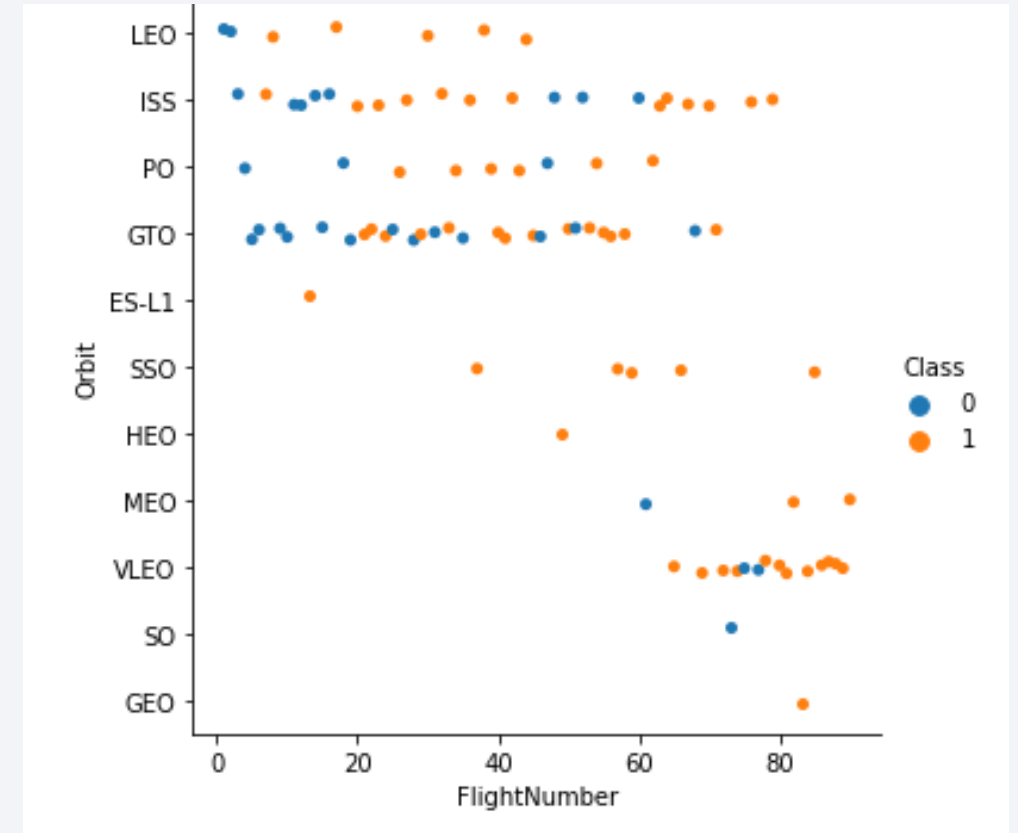
Success Rate vs. Orbit Type

- Highest success rate is observed for orbits:
 - ES-L1
 - GEO (a circular geosynchronous orbit 35,786 kilometers (22,236 miles) above Earth's equator)
 - HEO (Geocentric orbits above the altitude of geosynchronous orbit (35,786 km or 22,236 miles))
 - SSO (Sun-synchronous orbit, a nearly polar orbit around a planet)
- Lowest success rate is observed for orbits:
 - SO (error in the db – should have been combined with SSO as it's the same)
 - GTO
 - ISS (International Space Station)
 - MEO (most commonly at 20,200 kilometers (12,600 mi) or 20,650 kilometers (12,830 mi))
 - PO (poles orbit)
 - LEO (Low Earth orbit is an Earth-centred orbit with an altitude of 2,000 km or lower)



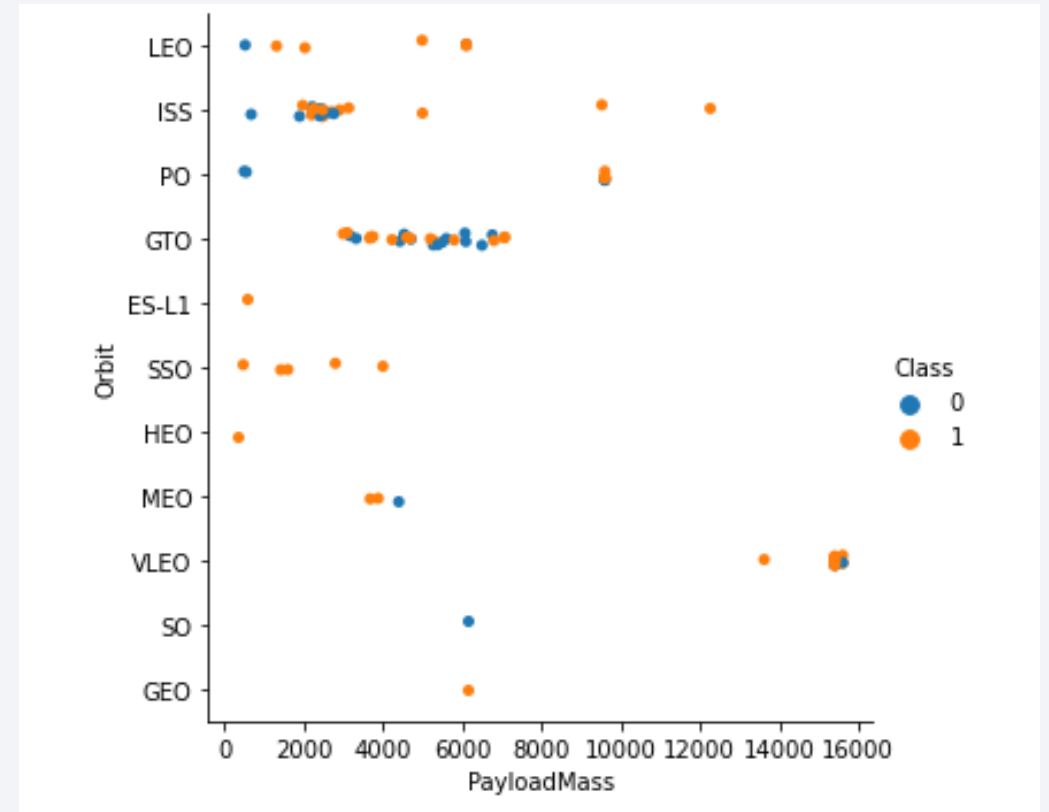
Flight Number vs. Orbit Type

- First attempts on low orbit types were not successful
- However, with more tries and errors almost all orbits launches became **successful**
- **Exception:**
 - GTO has many launches and still around 50% of success rate



Payload vs. Orbit Type

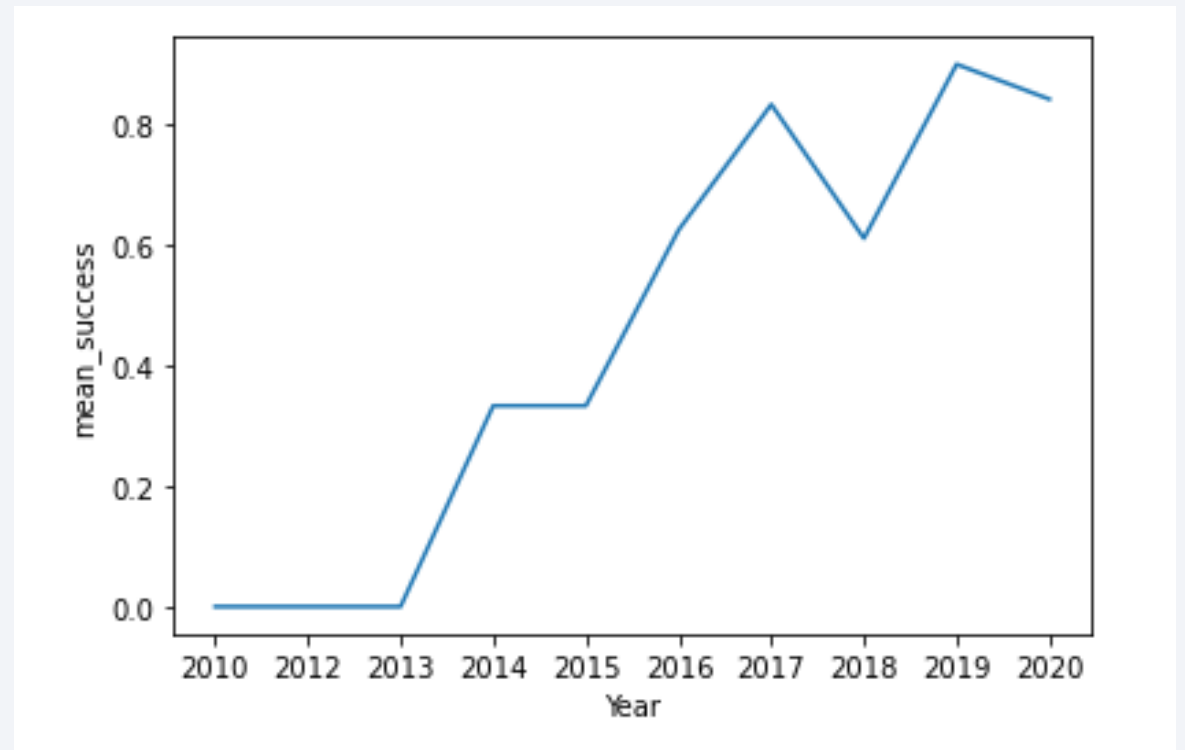
- Very heavy payloads were launched only to low orbits (and successfully):
 - VLEO (very low earth orbit)
 - ISS (international Space Station)
 - PO (poles orbit)
- Payloads <5,000 kg were effectively launch in 100% of cases to SSO, HEO and ES-L1



Launch Success Yearly Trend

With development of new booster versions and space technologies, the success rate has been gradually increasing and has achieved

> 80% in 2019



All Launch Site Names

- There are 4 launch sites:
 - 3 on East Coast (CCAFS x2 and KSC)
 - 1 on West Coast of US (VAFB SLC-4E)

launch_site

CCAFS LC-40

CCAFS SLC-40

KSC LC-39A

VAFB SLC-4E

Launch Site Names Begin with 'KSC'

- Below we can see first 5 records of the launch site KSC LC-39A, the site with the highest success rate.

DATE	time_utc_	booster_version	launch_site	payload	payload_mass_kg_	orbit	customer	mission_outcome	landing_outcome
2017-02-19	14:39:00	F9 FT B1031.1	KSC LC-39A	SpaceX CRS-10	2490	LEO (ISS)	NASA (CRS)	Success	Success (ground pad)
2017-03-16	06:00:00	F9 FT B1030	KSC LC-39A	EchoStar 23	5600	GTO	EchoStar	Success	No attempt
2017-03-30	22:27:00	F9 FT B1021.2	KSC LC-39A	SES-10	5300	GTO	SES	Success	Success (drone ship)
2017-05-01	11:15:00	F9 FT B1032.1	KSC LC-39A	NROL-76	5300	LEO	NRO	Success	Success (ground pad)
2017-05-15	23:21:00	F9 FT B1034	KSC LC-39A	Inmarsat-5 F4	6070	GTO	Inmarsat	Success	No attempt

Total Payload Mass

- In total rockets carried over 45,000 kg from NASA

total_payload

45596

Average Payload Mass by F9 v1.1

- The average payload mass carried by booster version F9 v1.1 equals to 2,534 KG

avg_payload

2534

First Successful Ground Landing Date

- The first successful landing outcome on ground pad happened on 8 Apr 2016

2016-04-08

Successful Drone Ship Landing with Payload between 4000 and 6000

- The following 3 boosters have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

booster_version

F9 B4 B1040.1

F9 B4 B1043.1

F9 FT B1032.1

Total Number of Successful and Failure Mission Outcomes

- Almost all mission were successful

mission_outcome	2
Failure (in flight)	1
Success	99
Success (payload status unclear)	1

Boosters Carried Maximum Payload

- The following booster versions carried the maximum payload mass
- All have version F9 B5...

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

2017 Launch Records

- In 2017 most of successful landing on ground pad happened from KSC LC-39A launch site

month_name	landing__outcome	booster_version	launch_site
February	Success (ground pad)	F9 FT B1031.1	KSC LC-39A
May	Success (ground pad)	F9 FT B1032.1	KSC LC-39A
June	Success (ground pad)	F9 FT B1035.1	KSC LC-39A
August	Success (ground pad)	F9 B4 B1039.1	KSC LC-39A
September	Success (ground pad)	F9 B4 B1040.1	KSC LC-39A
December	Success (ground pad)	F9 FT B1035.2	CCAFS SLC-40

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

- There were only 8 successful landing outcomes between the date 2010-06-04 and 2017-03-20
 - 5 on drone ship
 - 3 on ground pad

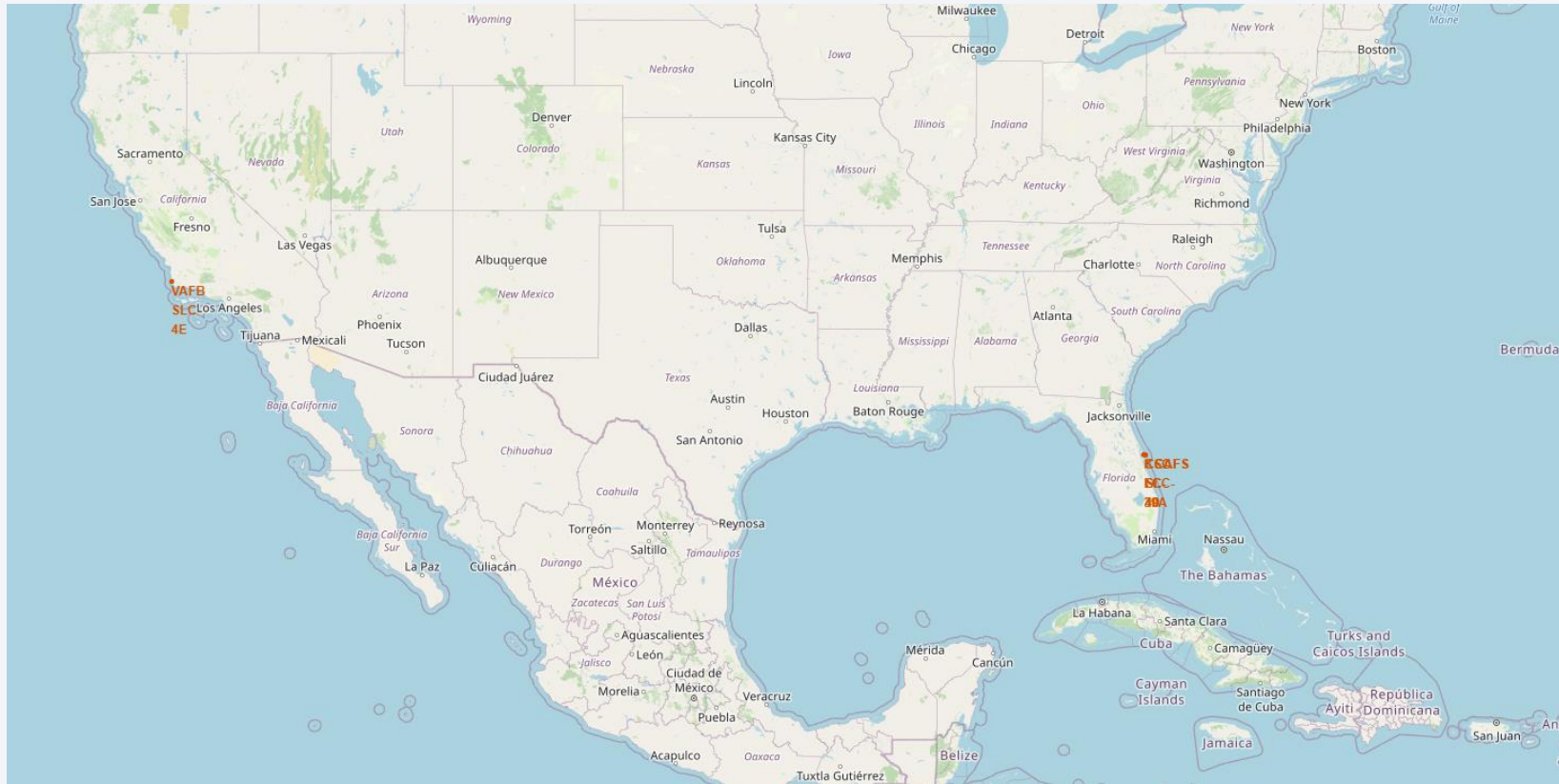
landing_outcome	cn
Success (drone ship)	5
Success (ground pad)	3

A satellite view of Earth from space, showing the curvature of the planet and city lights at night. The background is a deep blue gradient.

Section 3

Launch Sites Proximities Analysis

Launch sites' location



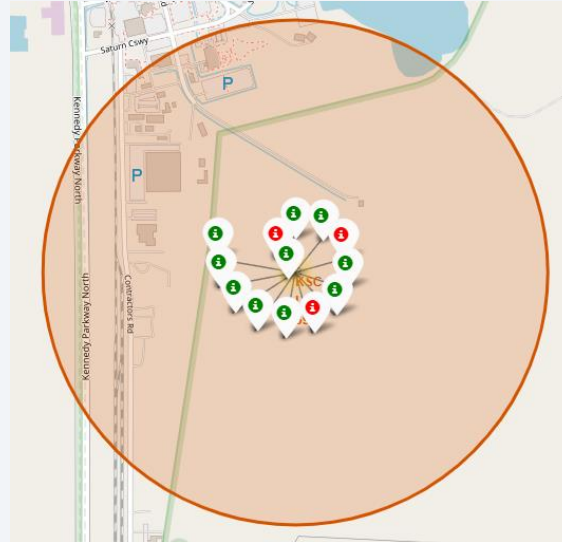
- All launch sites are very close to the coast
- All launch sites are as close to the equator as possible on the US territory

Launch outcomes

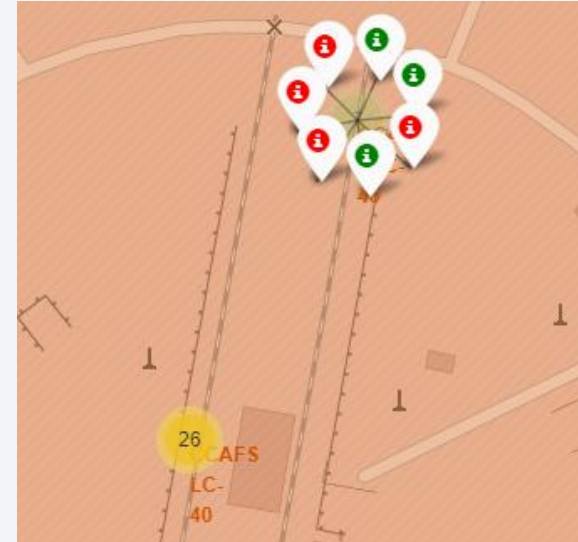
VAFB SLC-4E (West)



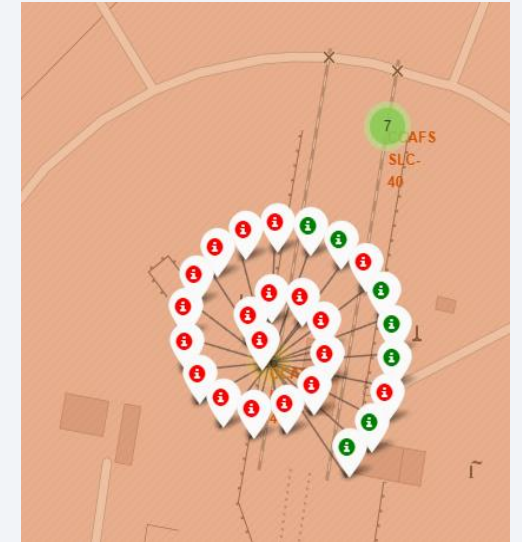
KSC LC-39A (East)



CCAFS SLC-40 (East)



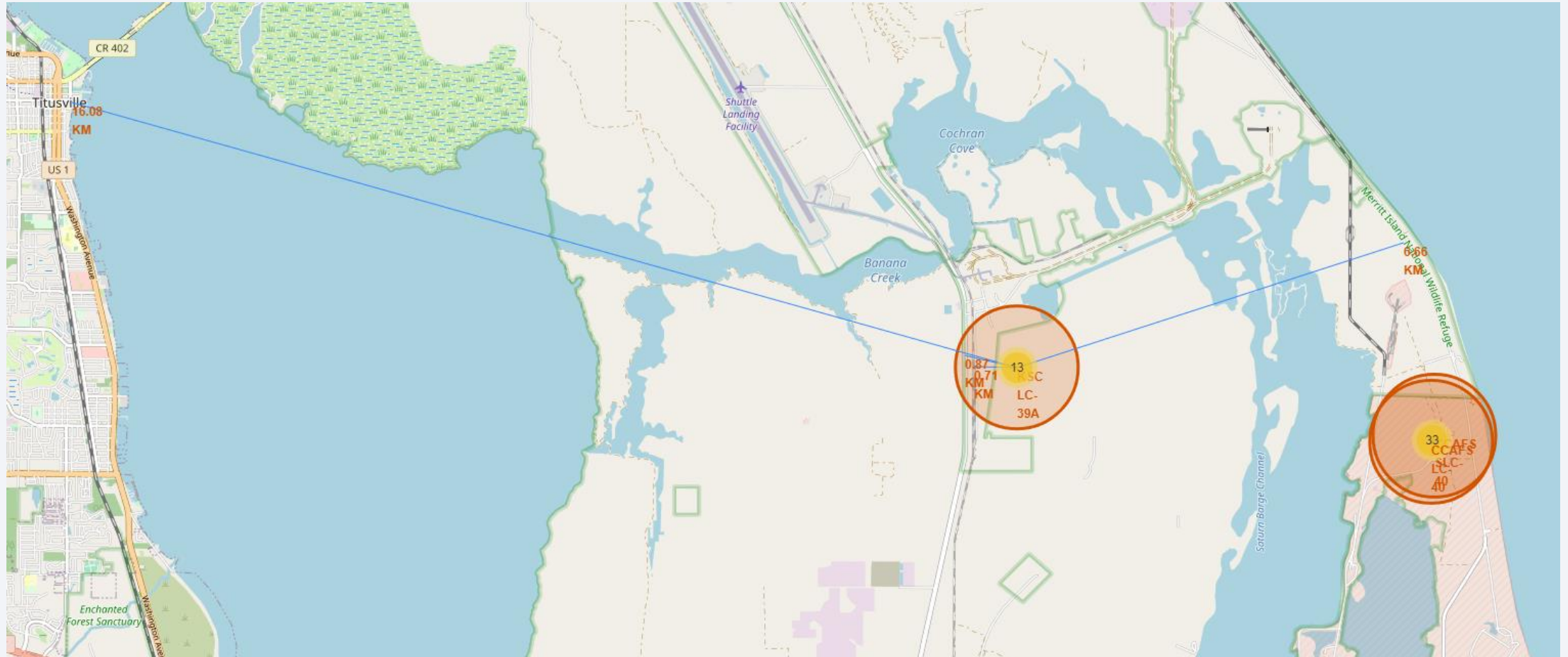
CCAFS LC-40 (East)



- KSC LC-39A launch site has the highest success rate among all sites
- Interesting that it's the only site which is located further from the ocean coast than other sites (by almost 7 km)

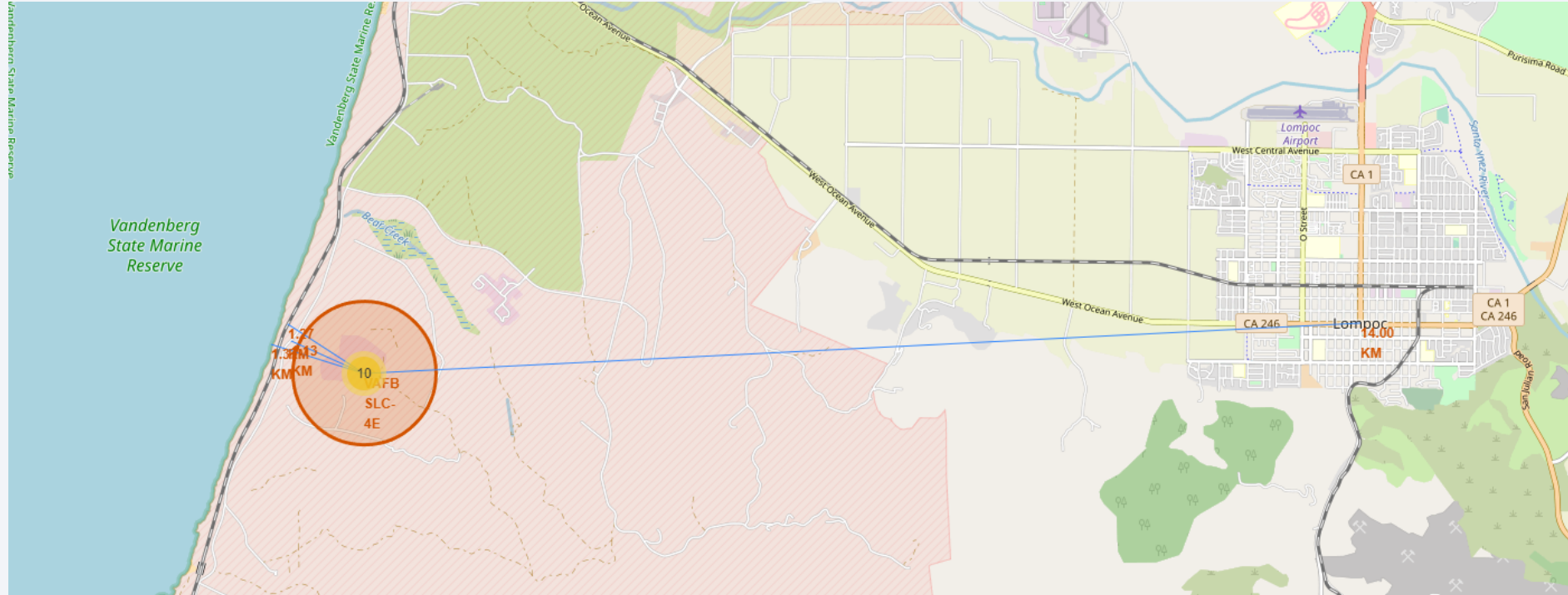
Site location and its proximities

KSC LC-39A (East)



Site location and its proximities

VAFB SLC-4E (West)



- All launch sites are in close distance to the railway and highway (0.5-1.5 km), 3 sites - to the coast as well (except KSC LC-39A as mentioned previously)
- However, they are quite far from the nearest city (>10 km)

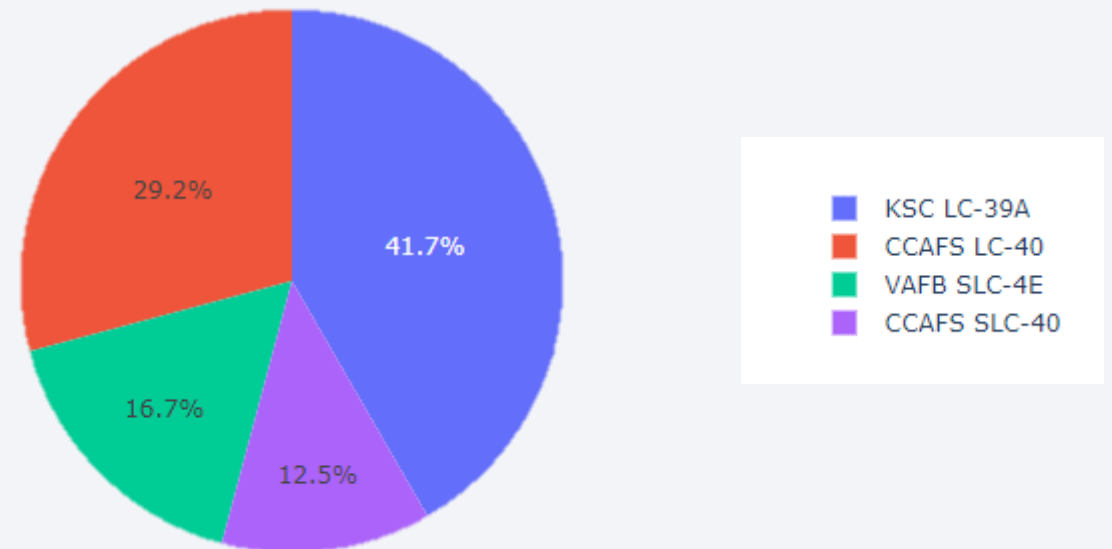


Section 4

Build a Dashboard with Plotly Dash

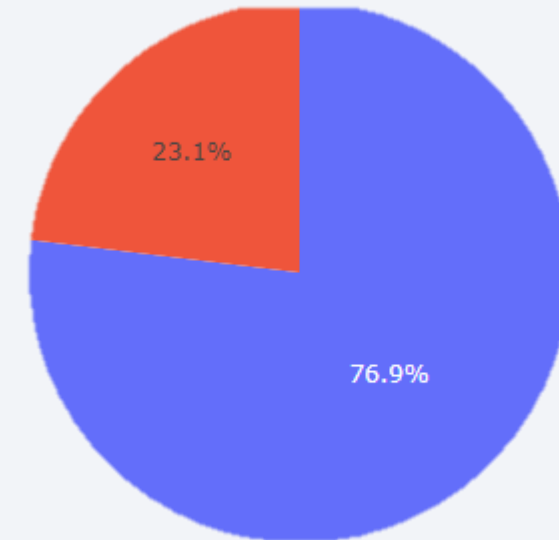
Launch success by sites

- KSC LC-39A has the highest number of success launches (41.7% out of all successes)
- 2 place is for CCAFS LC-40 (29.2%)



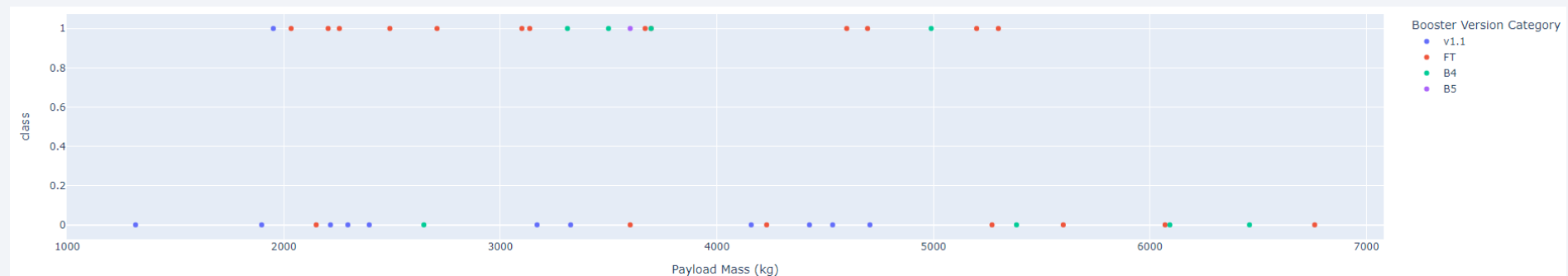
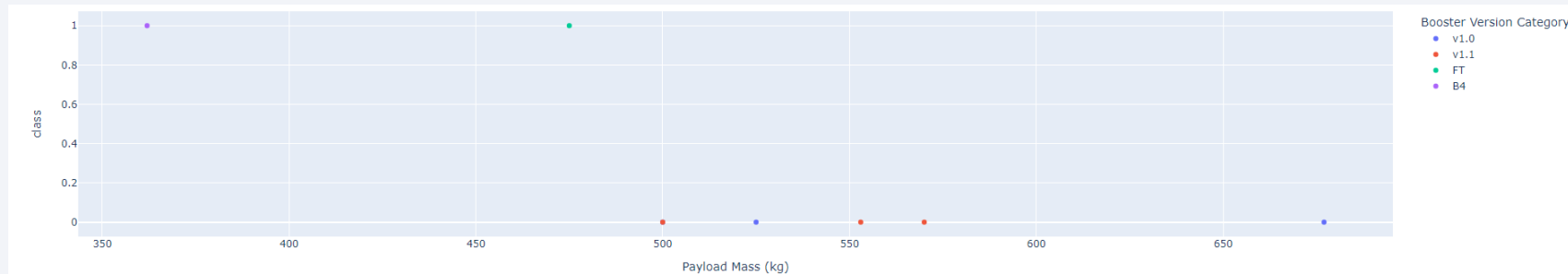
Launch success rate – KSC LC-39A

As mentioned previously, KSC LC-39A has also the highest success rate – **76.9%**



Payload vs Launch Outcome

- Payloads $< 1,000$ kg and $> 6,000$ kg has the lowest launch success rate
- Payloads between $2,000$ kg and $4,000$ kg has the highest launch success rate
- Booster version **v1.1** has the lowest success rate in the whole payload range
- Booster version **FT** has the highest success rate but only for 2,000-5000 kg payload

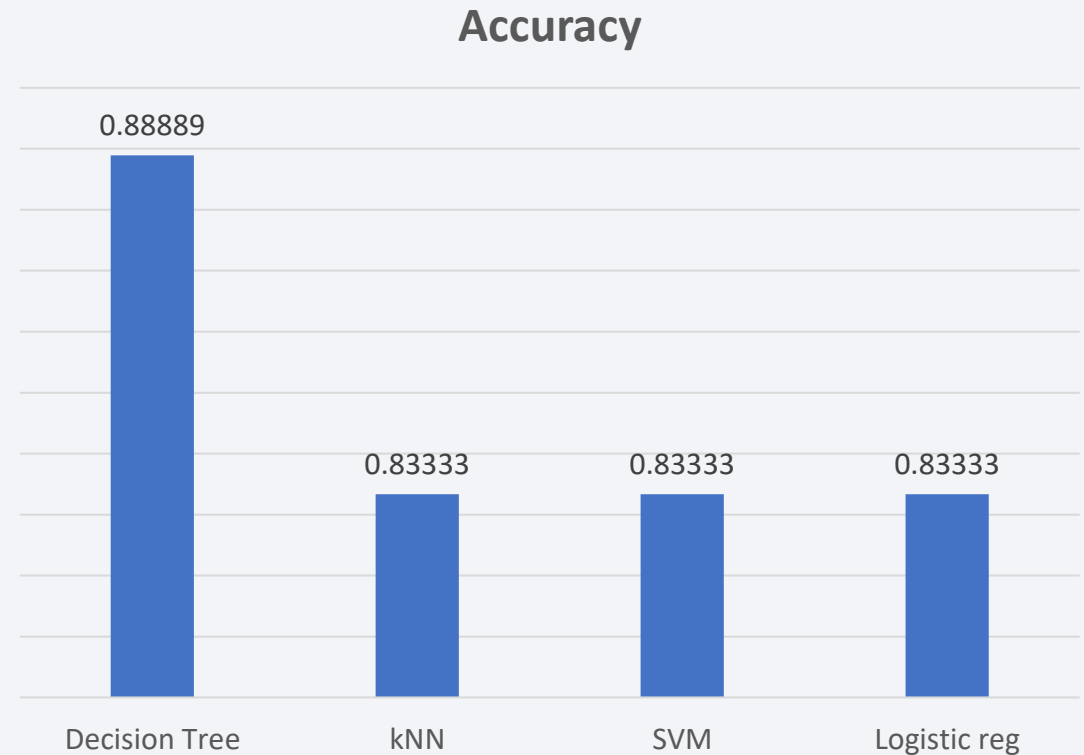


Section 5

Predictive Analysis (Classification)

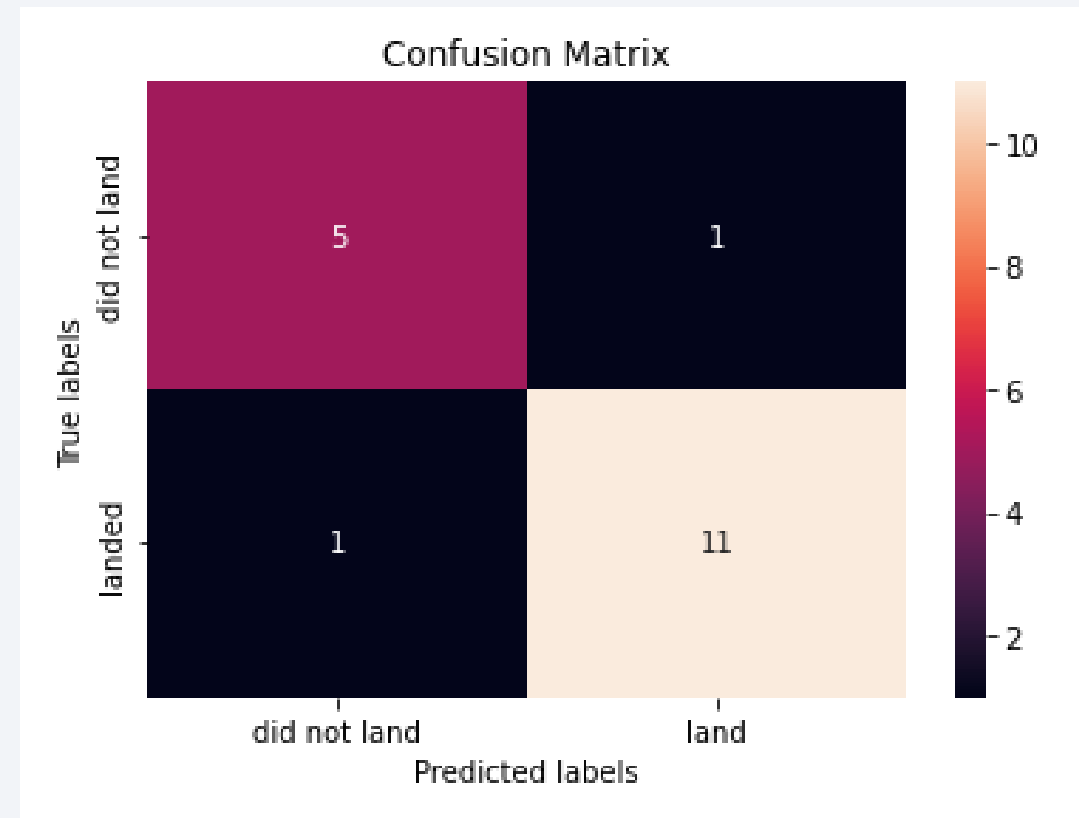
Classification Accuracy

- Decision Tree model showed the highest accuracy amongst other 3 models tested (kNN, SVM and Logistic regression)
- Best parameters chosen for the model are:
 - Criterion: entropy
 - Max Depth: 12
 - Min Samples leaf: 4
 - Min Sample split: 10



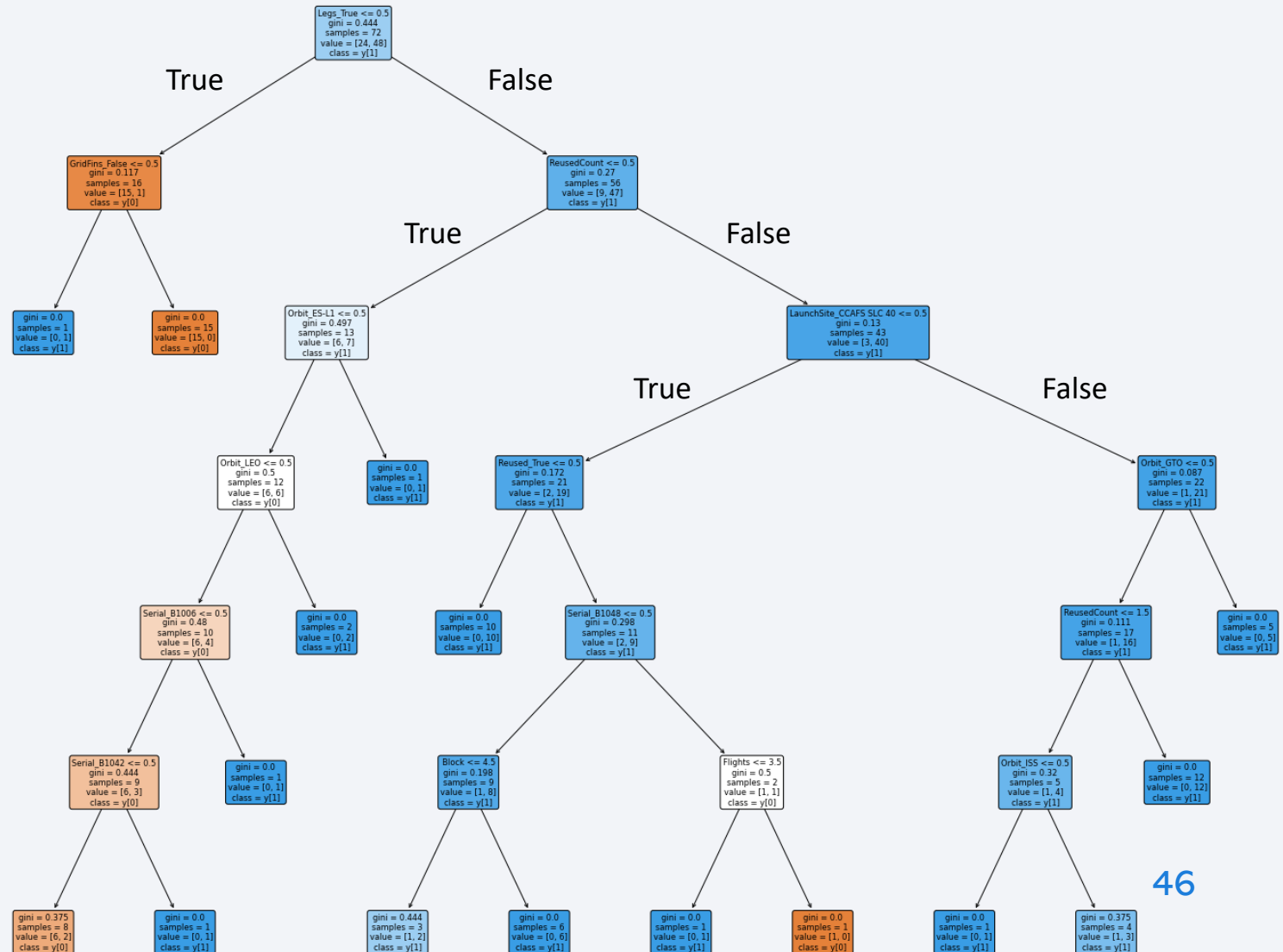
Confusion Matrix – Decision Tree

- On the graph below we can see that the model made only 2 mistakes:
 - 1 case which landed as didn't land
 - 1 case – vice verse.
- 11 out of 12 **landed** and 5 out of 6 **didn't land** were predicted accurately



Decision Tree

- In this illustration the analysis was done on the original, **not Scaled** Data as it'd be hard to interpret the results of the model with relative values
- The highest level of the tree says that if legs and grid fins were not used during landing (≤ 0.5), then the rocket will crash.
- If we go to False branch, we can have a success if:
 - using legs during landing
 - The rocket is not re-used
 - Orbit is either LEO or ES-L1



Conclusions

To successfully launch a new rocket and reuse the first stage with highest probability we need:

- Set up site closer to equator and not too close to the coast (around 6 km)
- Rocket must use legs during landing
- Choose closer orbit such
- Rocket should be re-used no more than 3 times

Appendix 1.1 – SQL used for analysis

Task 1

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

```
In [14]: %sql select distinct launch_site from spacextbl

* ibm_db_sa://bpb32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.

Out[14]: launch_site
CCAFS LC-40
CCAFS SLC-40
KSC LC-39A
VAFB SLC-4E
```

Task 2

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

```
In [15]: %sql select * from spacextbl where launch_site like 'KSC%' limit 5

* ibm_db_sa://bpb32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.

Out[15]:
```

	DATE	time_utc	booster_version	launch_site	payload	payload_mass_kg	orbit	customer	mission_outcome	landing_outcome
	2017-02-19	14:39:00	F9 FT B1031.1	KSC LC-39A	SpaceX CRS-10	2490	LEO (ISS)	NASA (CRS)	Success	Success (ground pad)
	2017-03-16	06:00:00	F9 FT B1030	KSC LC-39A	EchoStar 23	5600	GTO	EchoStar	Success	No attempt
	2017-03-30	22:27:00	F9 FT B1021.2	KSC LC-39A	SES-10	5300	GTO	SES	Success	Success (drone ship)
	2017-05-01	11:15:00	F9 FT B1032.1	KSC LC-39A	NROL-76	5300	LEO	NRO	Success	Success (ground pad)
	2017-05-15	23:21:00	F9 FT B1034	KSC LC-39A	Inmarsat-5 F4	6070	GTO	Inmarsat	Success	No attempt

Appendix 1.2 – SQL used for analysis

Task 3

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

```
In [20]: %sql select sum(payload_mass_kg_) as total_payload from spacextbl where customer='NASA (CRS)'
```

* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.

```
Out[20]: total_payload
```

45596

Task 4

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

```
In [23]: %sql select avg(payload_mass_kg_) as avg_payload from spacextbl where booster_version like 'F9 v1.1%'
```

* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.

```
Out[23]: avg_payload
```

2534

Task 5

List the date where the first succesful landing outcome in drone ship was acheived.

Hint: Use min function

```
In [25]: %sql select min(DATE) from spacextbl where landing_outcome='Success (drone ship)'
```

* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90108kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.

```
Out[25]: 1
```

2016-04-08

Appendix 1.3 – SQL used for analysis

Task 6

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

```
In [27]: %sql select distinct booster_version from spacextbl where landing__outcome='Success (ground pad)' and payload_mass__kg_>4000 and payload_mass__kg_<6000

* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.
```

```
Out[27]: booster_version
F9 B4 B1040.1
F9 B4 B1043.1
F9 FT B1032.1
```

Task 7

List the total number of successful and failure mission outcomes

```
In [29]: %sql select mission_outcome,count(*) from spacextbl group by mission_outcome

* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.
```

```
Out[29]: mission_outcome  2
Failure (in flight)      1
Success                  99
Success (payload status unclear)  1
```

Task 8

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

```
In [32]: %sql select booster_version from spacextbl s, (select max(payload_mass__kg_) as max_payload from spacextbl)m where s.payload_mass__kg_=m.max_payload

* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludb
Done.
```

Appendix 1.4 – SQL used for analysis

Task 9

List the records which will display the month names, succesful landing_outcomes in ground pad ,booster versions, launch_site for the months in year 2017

```
In [46]: %sql select sysfun.monthname(DATE) as month_NAME, landing__outcome,booster_version, launch_site from spacextbl where landing__outcome like 'Success (g  
* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludb  
Done.
```

```
Out[46]:
```

month_name	landing__outcome	booster_version	launch_site
February	Success (ground pad)	F9 FT B1031.1	KSC LC-39A
May	Success (ground pad)	F9 FT B1032.1	KSC LC-39A
June	Success (ground pad)	F9 FT B1035.1	KSC LC-39A
August	Success (ground pad)	F9 B4 B1039.1	KSC LC-39A
September	Success (ground pad)	F9 B4 B1040.1	KSC LC-39A
December	Success (ground pad)	F9 FT B1035.2	CCAFS SLC-40

Task 10

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

```
In [36]: %sql select landing__outcome, count(*) as cn from spacextbl where landing__outcome like 'Success%' and DATE between '2010-06-04' and '2017-03-20' grou  
* ibm_db_sa://bpp32620:***@0c77d6f2-5da9-48a9-81f8-86b520b87518.bs2io90l08kqb1od8lcg.databases.appdomain.cloud:31198/bludb  
Done.
```

```
Out[36]:
```

landing__outcome	cn
Success (drone ship)	5
Success (ground pad)	3

Appendix 2.1 – Data preparation for modeling

1. Standardize X variables:

```
transform = preprocessing.StandardScaler()
```

```
X=transform.fit(X).transform(X)
```

2. Split into Train and Test sets:

```
X_train, X_test, Y_train, Y_test=train_test_split(X,Y,test_size=0.2, random_state=2)
```

Appendix 2.2 – Parameters tuning

GridSearchCV is used for parameters tuning. Example of Logistic Regression:

```
lr=LogisticRegression()
logreg_cv=GridSearchCV(lr,parameters,cv=10)
logreg_cv.fit(X_train, Y_train)

print("tuned hyperparameters :(best parameters) ",logreg_cv.best_params_)
print("accuracy :",logreg_cv.best_score_)
```

```
tuned hyperparameters :(best parameters) {'C': 0.01, 'penalty': 'l2', 'solver': 'lbfgs'}
accuracy : 0.8464285714285713
```

Thank you!

