



IBM Developer  
SKILLS NETWORK

# Winning Space Race with Data Science

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# Outline

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

# Executive Summary

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## ➤ Summary of methodologies

- Data Collection & Wrangling
- Data Exploration with Visualization
- Interactive Visualization/Dashboard with Folium and Plotly
- Machine Learning (predictive analysis on classification)

## ➤ Summary of all results

- Best launch site: KSC LC-39A
- Best ML model: Decision Tree
- Correlation between launch success rate and number of launches

# Introduction

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## ➤ Project Background and context

- Falcon 9 first stage landing prediction
- Competition from other providers
- Cost of launch and success rate of launch will determine SpaceX's pricing and bidding

## ➤ Problems that need answers

- What are the factors that related to success launches?
- What are the sites that perform best with launches?
- What are the factors that determine successful landings?



Section 1

# Methodology

# Methodology

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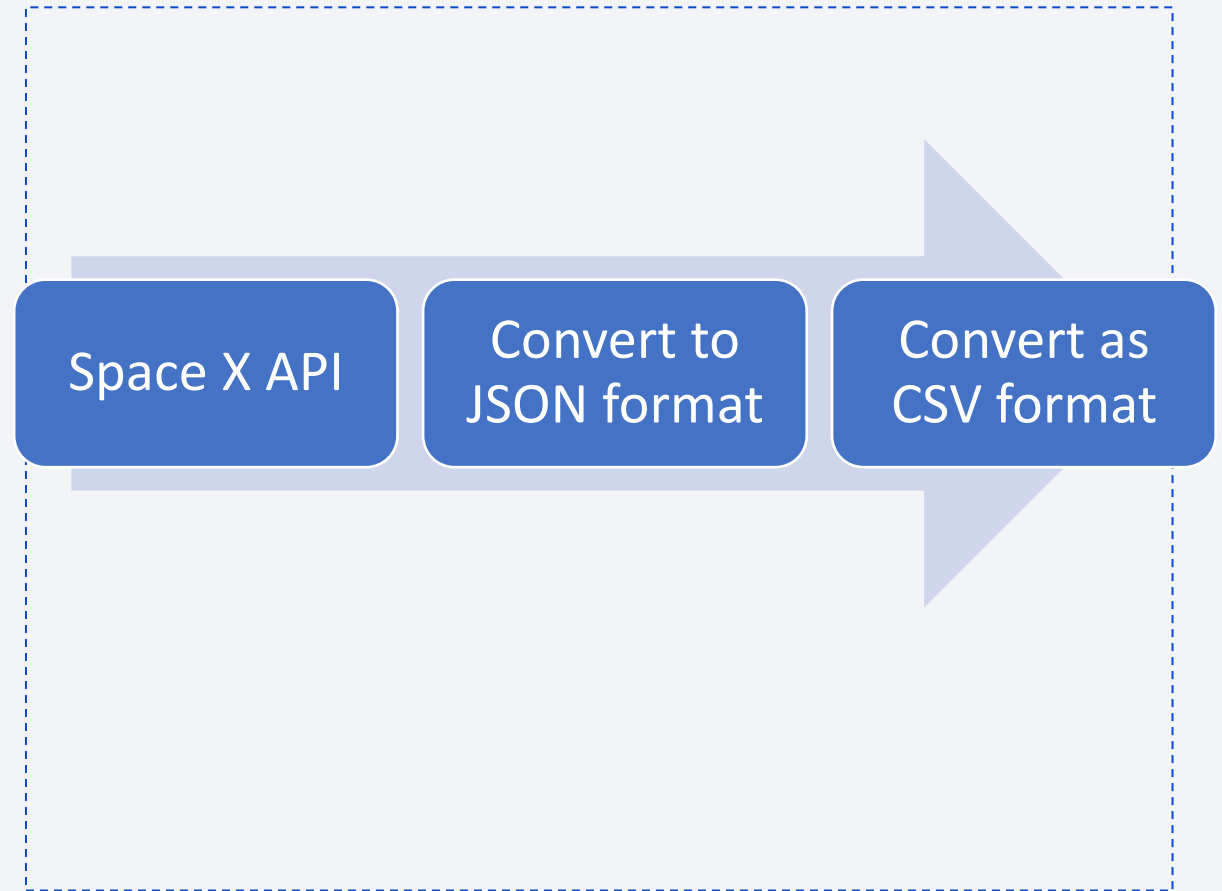
## Executive Summary

- Data collection methodology:
  - Direct: SpaceX API
  - Indirect: Wikipedia Web scrapping
- Perform data wrangling
  - Data normalization, grouping
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification model
  - Run through 4 machine learning models to find best one

# Data Collection – SpaceX API

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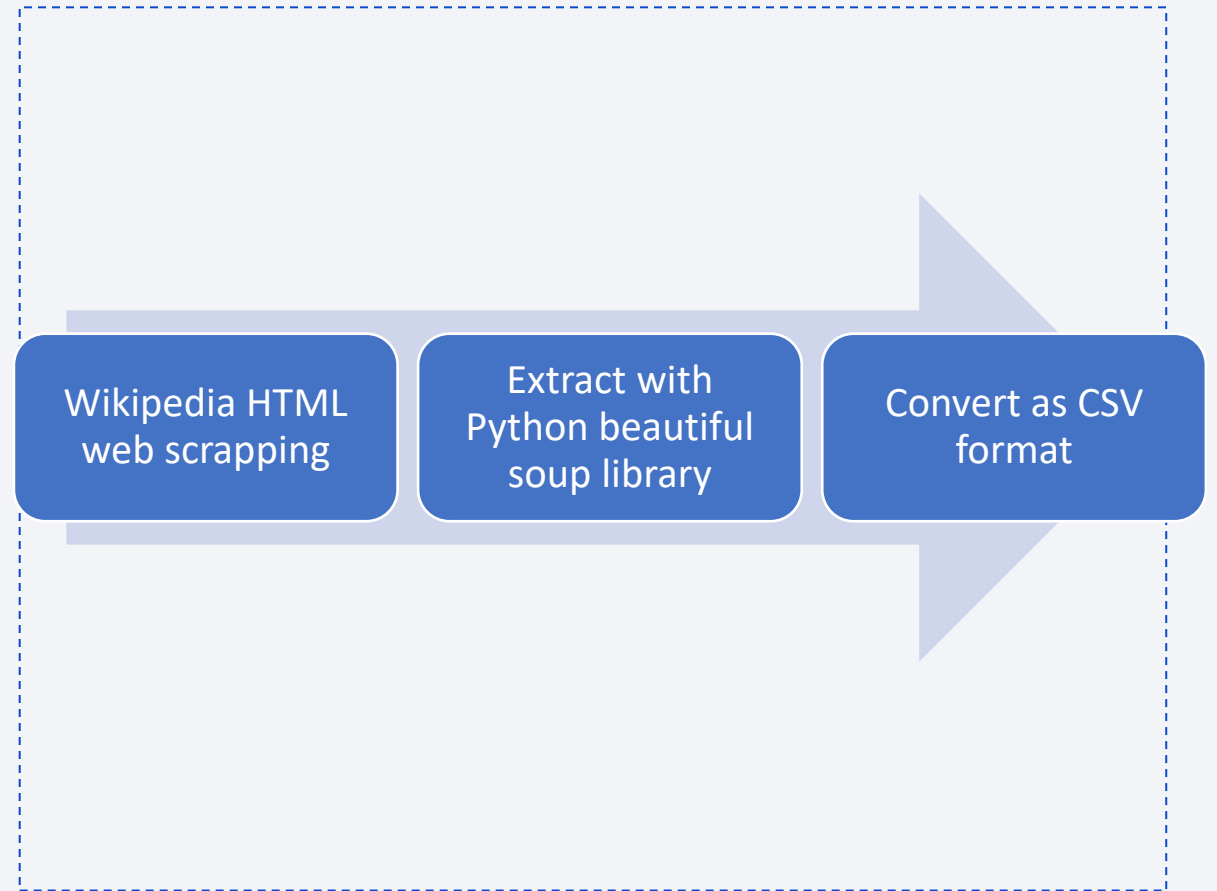
- Data collection with SpaceX REST calls using key phrases and flowcharts
- [GitHub URL of the completed SpaceX API calls](#)



# Data Collection - Scraping

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- Web scraping process using key phrases and flowcharts
- [GitHub URL of the completed web scraping notebook](#)





# Data Wrangling

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- Data wrangling process using key phrases and flowcharts
- [GitHub URL](#)



- Number of launches at each site



- Number and occurrence of each orbit



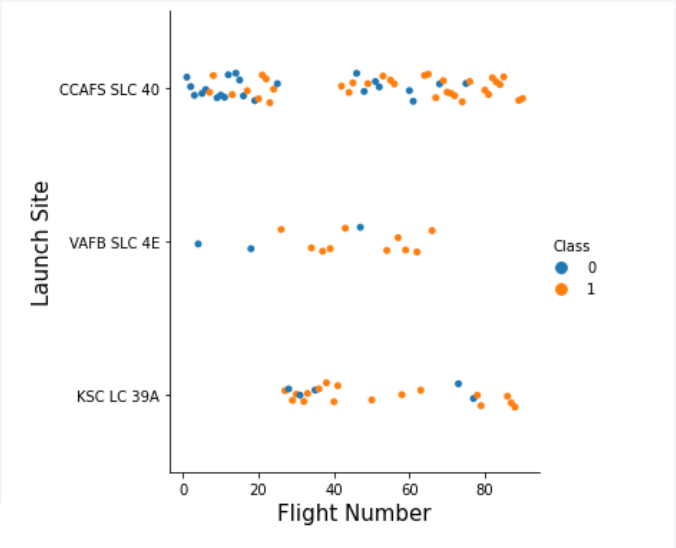
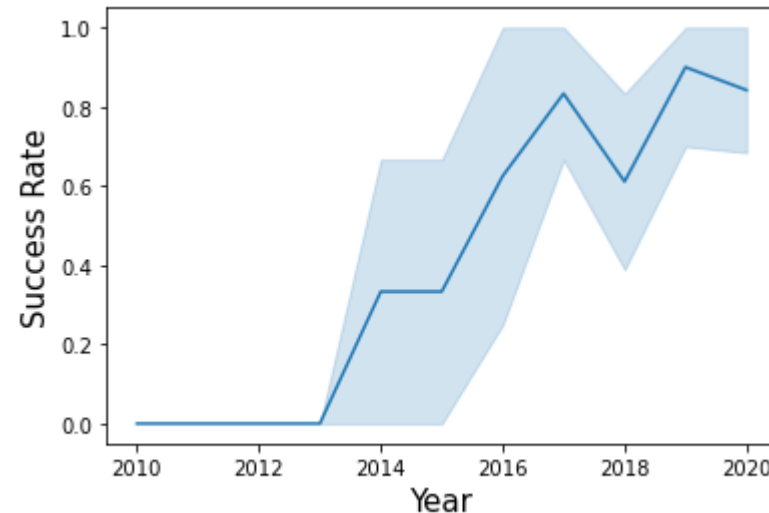
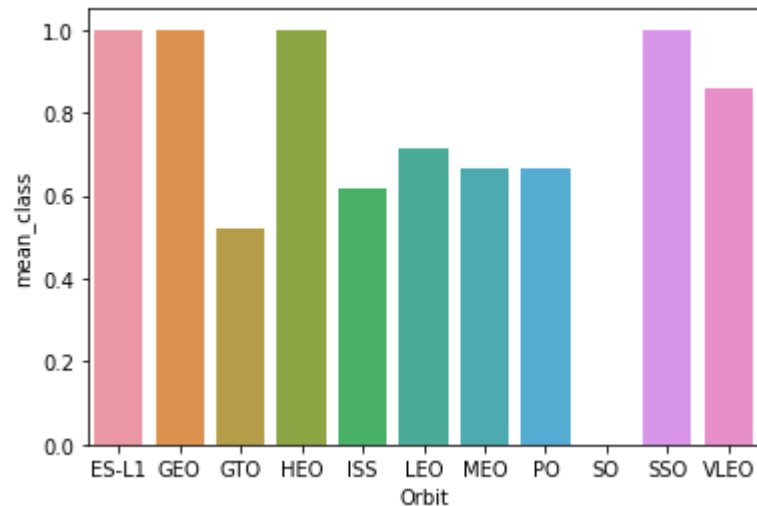
- Number and occurrence of mission outcome per orbit type



- Create a landing outcome label from outcome column

# EDA with Data Visualization

- Scatter plot: easy to tell dependencies between variables
- Bar chart: on categorical values
- Line chart: clear on trends with time
- [GitHub URL](#)



# EDA with SQL

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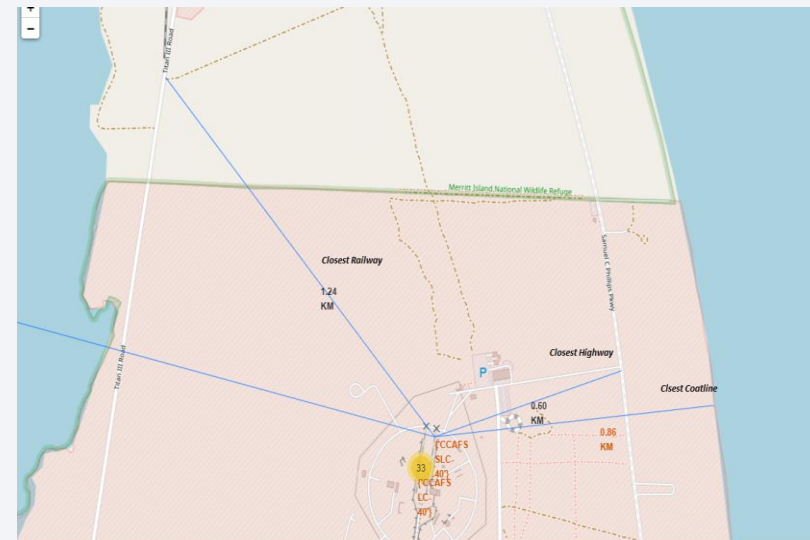
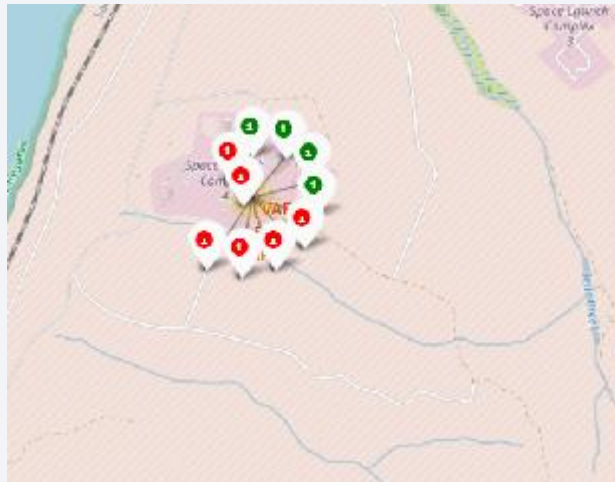
## SQL queries performed

- Names of the unique launch sites in the space mission
- Display 5 records where launch sites begin with the string 'CCA'
- Total payload mass carried by boosters launched by NASA (CRS)
- Display average payload mass carried by booster version F9 v1.1
- List the date when the first successful landing outcome in ground pad was achieved
- Names of the boosters which have success in drone ship and have payload mass greater than 4000 but less than 6000
- The total number of successful and failure mission outcomes
- The names of the booster\_versions which have carried the maximum payload mass. Use a subquery
- [GitHub URL](#)

# Build an Interactive Map with Folium

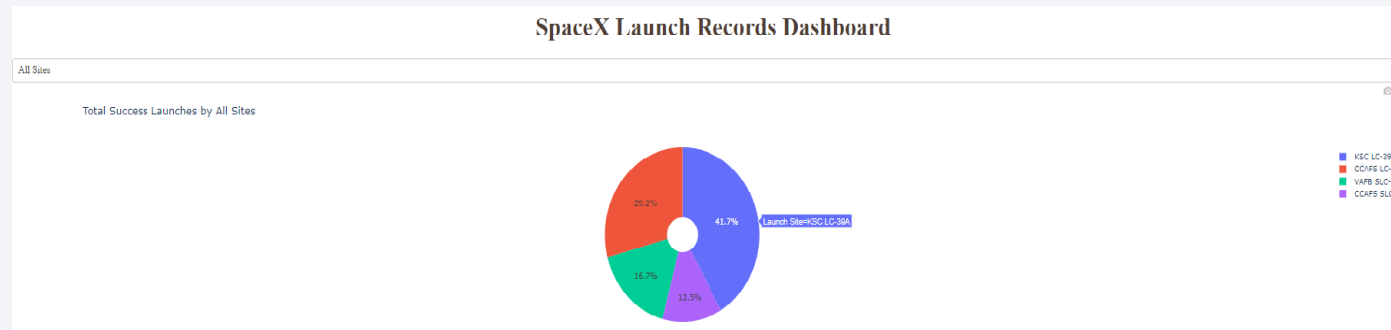
---

- Map objects such as markers, circles, lines are added to a folium map by Longitude and Latitude
- Markers with **green** and **red** color identified **success** and **failures** launches
- Lines are to display the distance of a launch site to nearest railway, highway, coastline and cities.
- [GitHub URL](#)



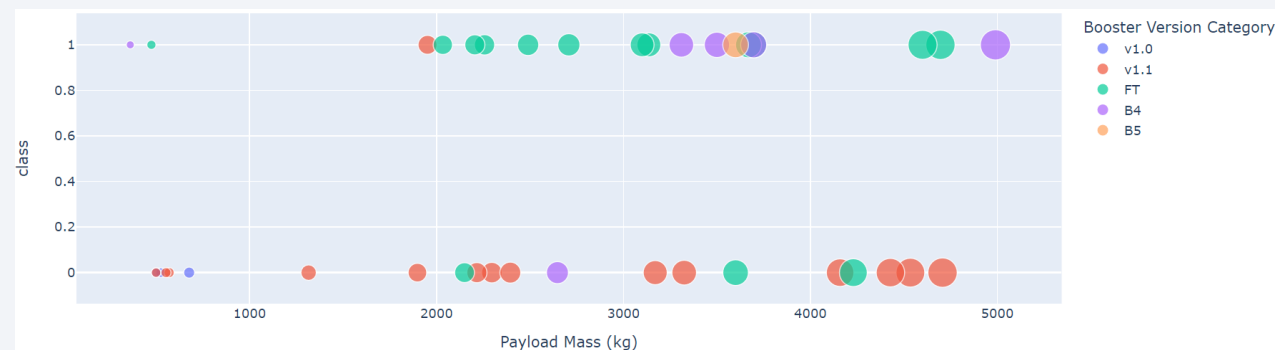
# Build a Dashboard with Plotly Dash

- Interactive pie charts to display success rate on all sites and a selector to narrow down to individual sites

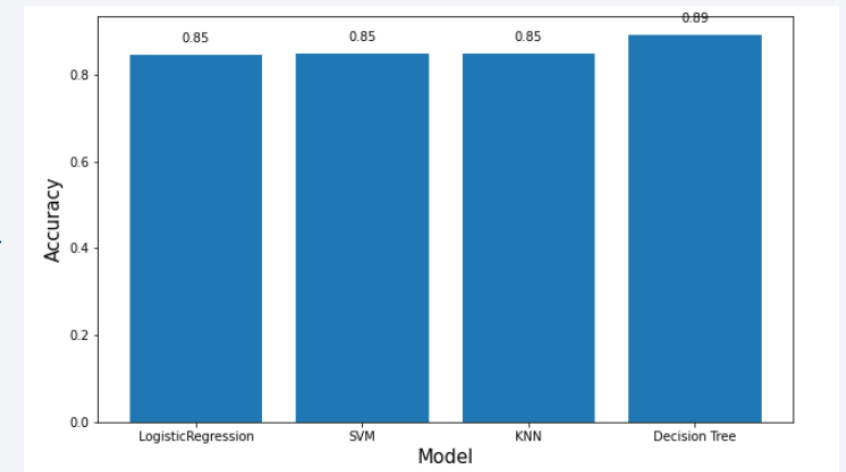
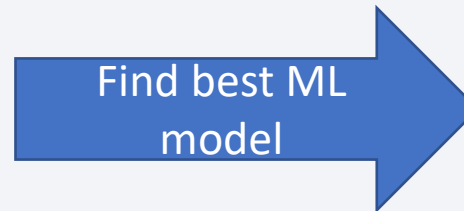
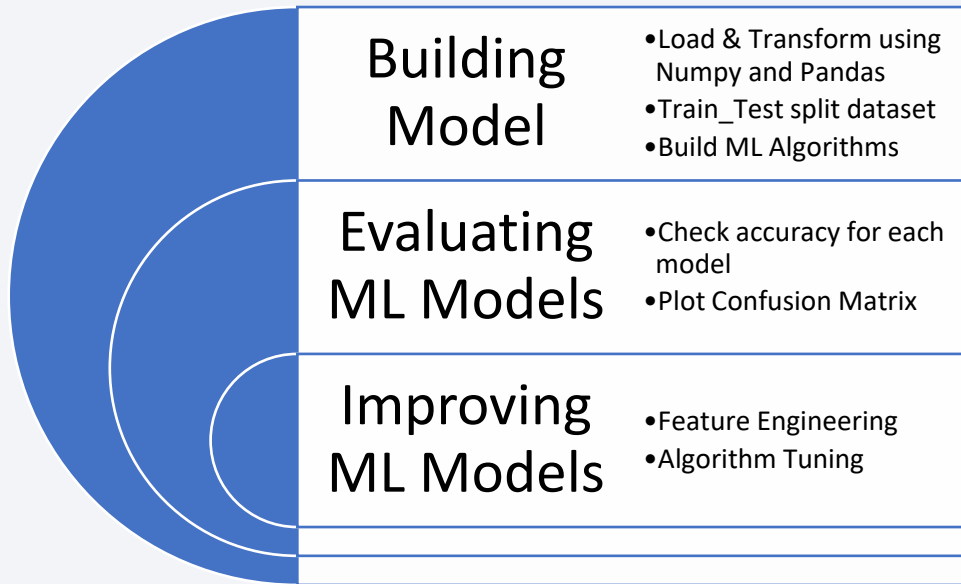


- Scatter plot to display success and failures on all payload range with adjustable range selector

- [GitHub URL](#)



# Predictive Analysis (Classification)



- [GitHub URL](#)



# Results

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



The background of the slide is a complex, abstract composition. It features a dark blue base color. Overlaid on this are numerous diagonal streaks and lines in shades of red and cyan. These lines vary in thickness and opacity, creating a sense of depth and movement. A faint, light blue grid pattern is also visible, particularly in the lower-left quadrant. The overall effect is a high-tech, digital aesthetic.

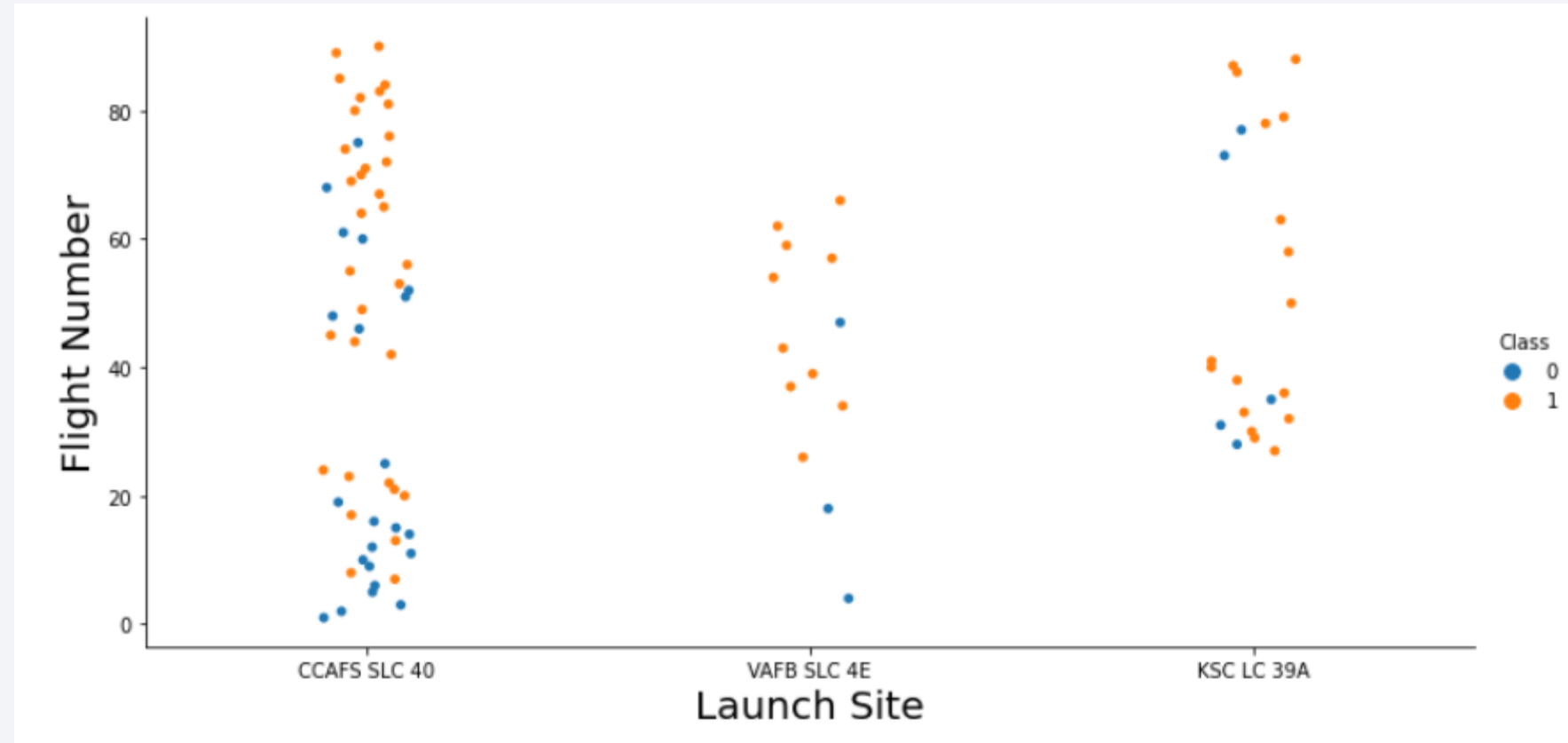
Section 2

# Insights drawn from EDA



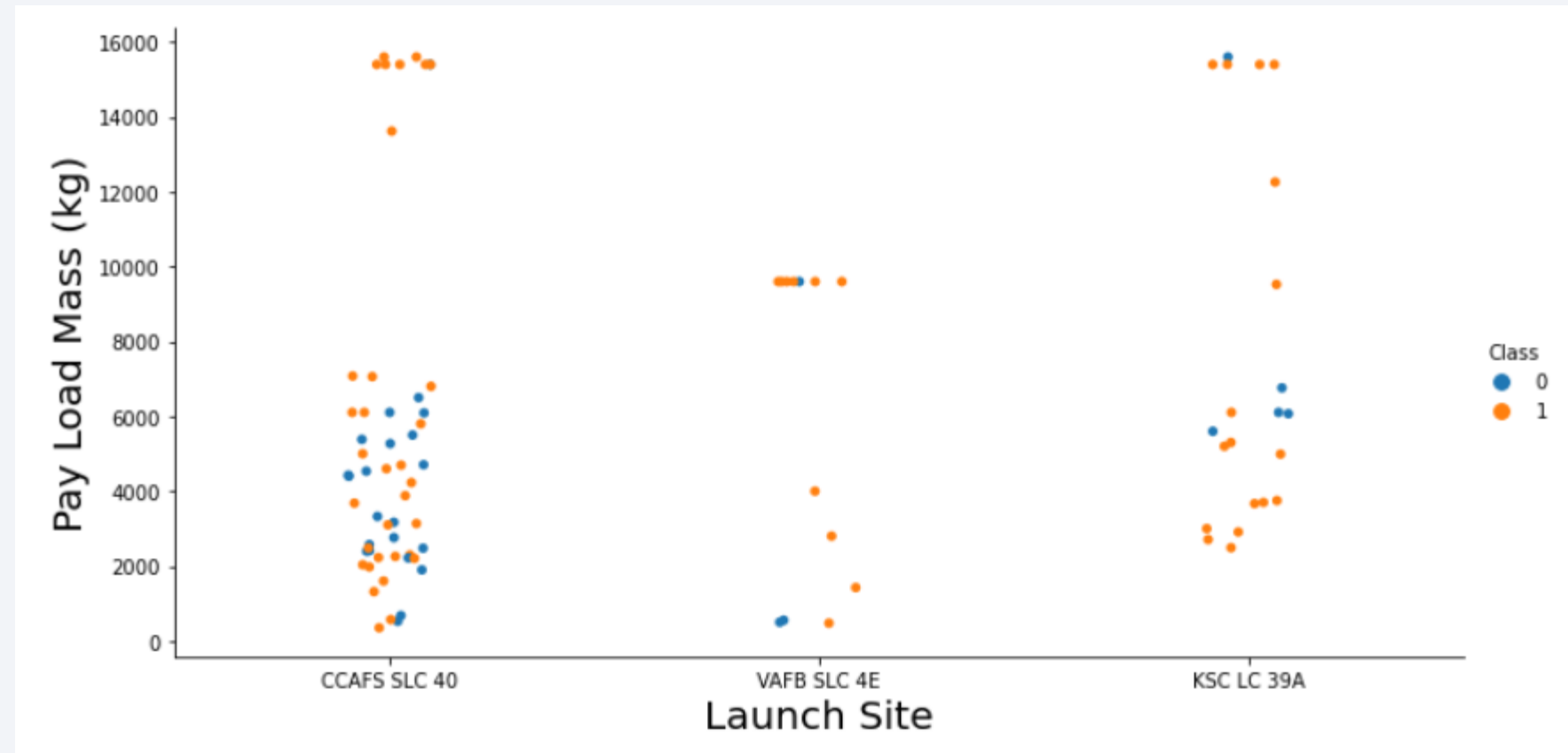
# Flight Number vs. Launch Site

- Class=0 Launch fail
- Class=1 Launch success
- Site with more launches tend to have more successful launches



# Payload vs. Launch Site

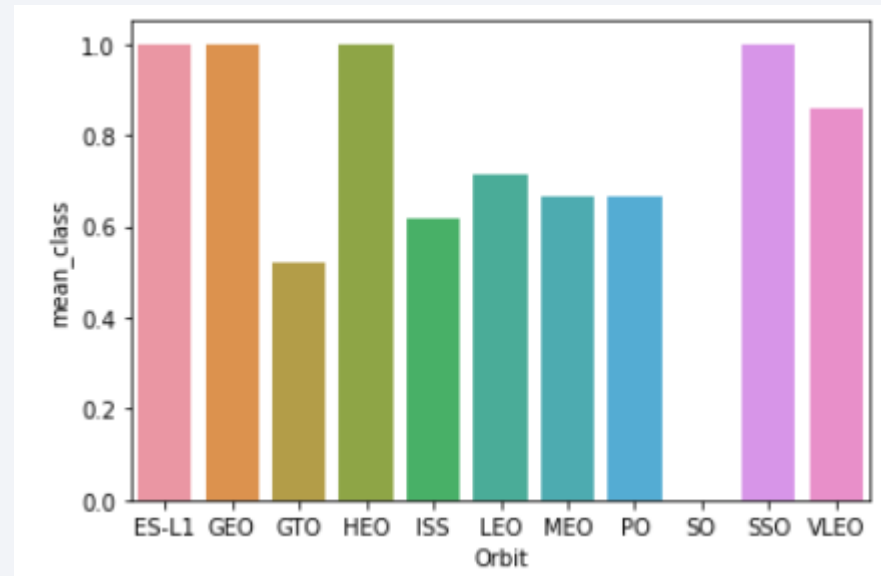
- Class=0 Launch fail
- Class=1 Launch success
- Site CCAFS SLC 40 has more successful rate on heavy pay load launches



# Success Rate vs. Orbit Type

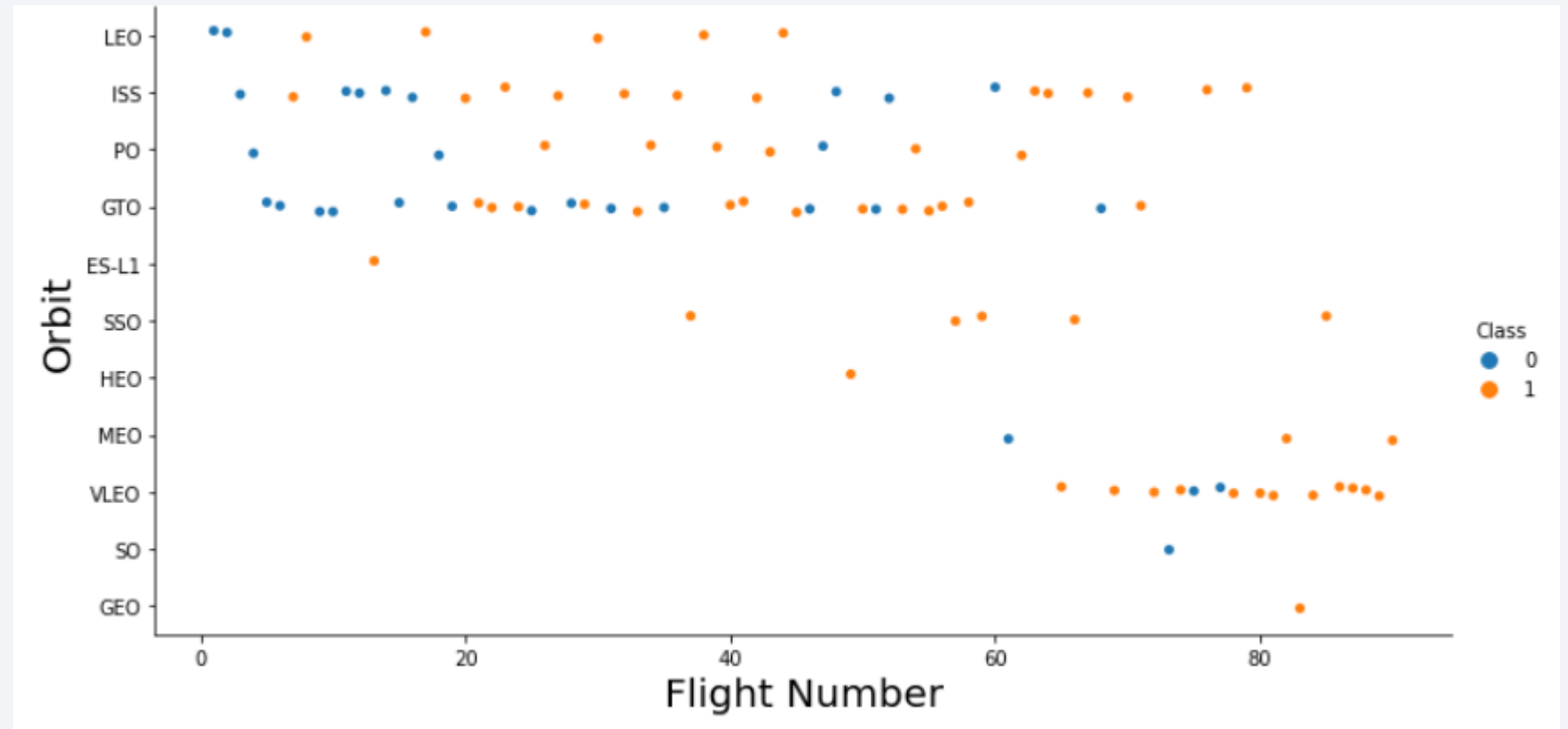
---

- The more mean\_class close to 1 the more success launches with a orbit
- ES-L1, GEO, HEO and SSO have the best success rates



# Flight Number vs. Orbit Type

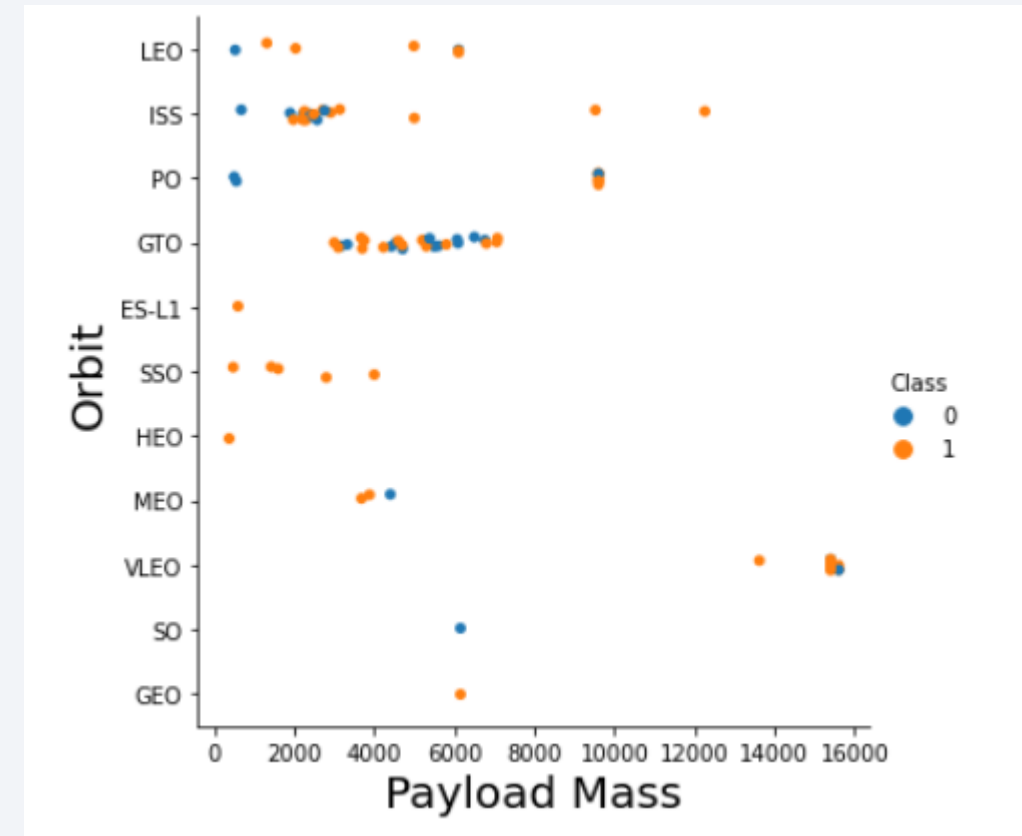
- Class=0 Launch fail
- Class=1 Launch success
- The higher the flight number go, the higher success rates are for launches (60-80)





# Payload vs. Orbit Type

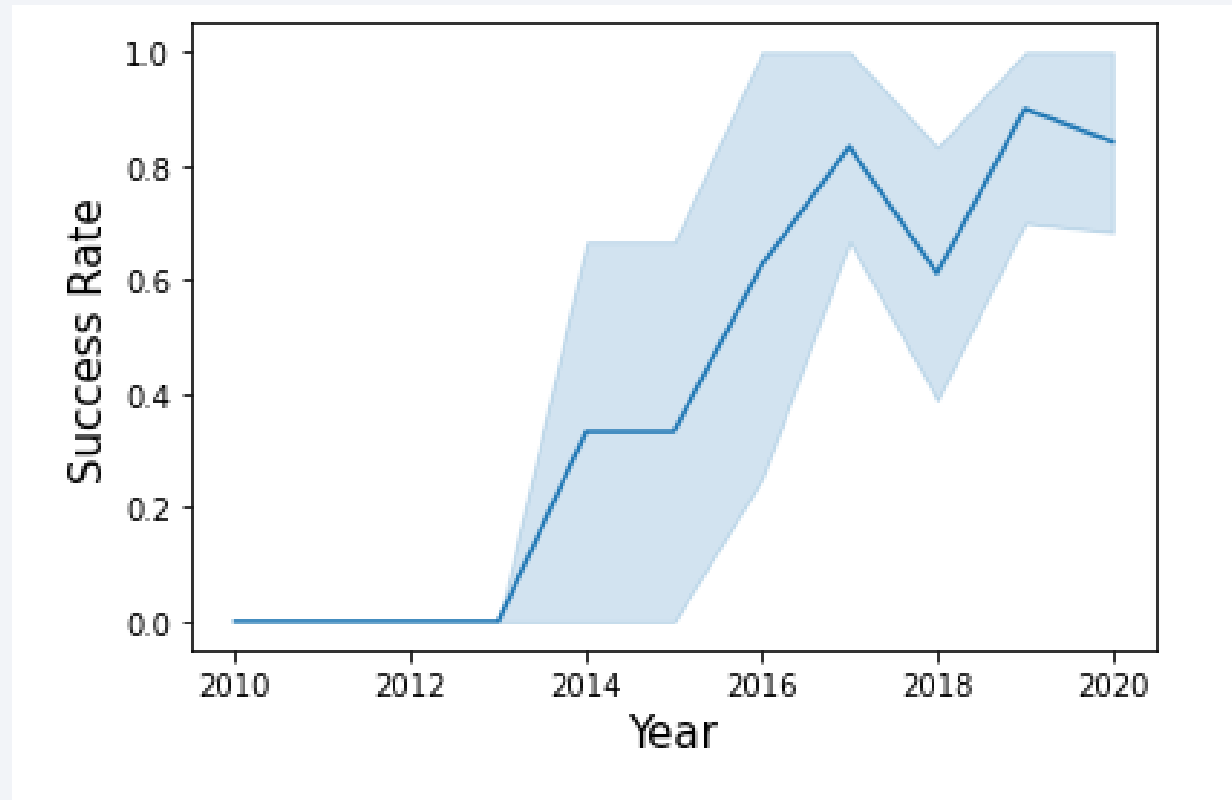
- Class=0 Launch fail
- Class=1 Launch success
- Launch failure mainly happened on under 8000 KG and GTO Orbit



# Launch Success Yearly Trend

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- With time, Space X's launch success rate grow exponentially
- 2018 Space X has a small set back on launch success rate



# All Launch Site Names

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- Below SQL query are showing all unique launch site names
- %sql is using SQL Magic method

```
%sql select distinct launch_site from SPACEXDATASET
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

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

# Launch Site Names Begin with 'CCA'

- Find 5 records where launch sites begin with 'CCA'
- %sql is using SQL Magic method

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

```
%sql select * from SPACEXDATASET where launch_site like 'CCA%' limit 5
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/blddb  
Done.
```

DATE	time__utc_	booster_version	launch_site	payload	payload_mass__kg_	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

# Total Payload Mass

---

- Calculate the total payload carried by boosters from NASA
- Confirm SQL number with Python Pandas calculation

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

```
%sql select sum(payload_mass_kg_) as total_payload from SPACEXDATASET where customer='NASA (CRS)'
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

total_payload
---------------

45596
-------

```
#Pandas solution
```

```
df[df["CUSTOMER"]=="NASA (CRS)"].PAYLOAD_MASS__KG_.sum()
```

```
45596
```

# Average Payload Mass by F9 v1.1

---

- Calculate the average payload mass carried by booster version F9 v1.1
- Confirm SQL result with Python Pandas result

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

```
%sql select avg(payload_mass__kg_) as Average_Payload_Mass from SPACEXDATASET where booster_version='F9 v1.1'
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

average_payload_mass
----------------------

2928
------

*#Pandas solution*

```
df[df["BOOSTER_VERSION"]=="F9 v1.1"].PAYLOAD_MASS__KG_.mean()
```

2928.4



# First Successful Ground Landing Date

---

- Find the dates of the first successful landing outcome on ground pad
- Confirm SQL result with Python Pandas result

*List the date when the first successful landing outcome in ground pad was achieved.*

*Hint: Use min function*

```
%sql select min(DATE) as First_Date from SPACEXDATASET where landing__outcome='Success'
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

first_date
------------

2018-07-22
------------

```
#Pandas solution  
df[df["LANDING__OUTCOME"]=="Success"].DATE.min()
```

```
datetime.date(2018, 7, 22)
```

## Successful Drone Ship Landing with Payload between 4000 and 6000

---

- List the names of boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000

*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 SPACEXDATASET where landing__outcome='Success (drone ship)' and payload_mass__kg_>4000 and payload_mass__kg_<6000
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

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

# Total Number of Successful and Failure Mission Outcomes

---

- Calculate the total number of successful and failure mission outcomes
- Because there are multiple Success and Failure type, this statement shows all outcomes

*List the total number of successful and failure mission outcomes*

```
%sql select mission_outcome, count(*) as result from SPACEXDATASET group by mission_outcome
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

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

# Boosters Carried Maximum Payload

---

- List the names of the booster which have carried the maximum payload mass
- Sub query extract max payload number for each booster\_version and match with main query
- Main query returns unique booster versions

*List the names of the booster\_versions which have carried the maximum payload mass. Use a subquery*

```
%sql select distinct booster_version from SPACEXDATASET x1 where booster_version=(select booster_version from SPACEXDATASET x2 where x1.booster_version=x2.booster_version order by x2.payload_mass__kg_ DESC limit 1)
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

# 2015 Launch Records

---

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

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

```
%sql select booster_version,launch_site,landing__outcome from SPACEXDATASET where landing__outcome ='Failure (drone ship)' and year(
DATE)='2015'
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb
Done.
```

booster_version	launch_site	landing__outcome
F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

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

---

- Rank the count of landing outcomes (such as Failure (drone ship) or Success (ground pad)) between the date 2010-06-04 and 2017-03-20, in descending order

```
%sql select landing__outcome,count(*) as rank from SPACEXDATASET where date between '2010-06-04' and '2017-03-20' group by landing__outcome order by rank desc
```

```
* ibm_db_sa://mj190806:***@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb  
Done.
```

landing__outcome	RANK
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



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

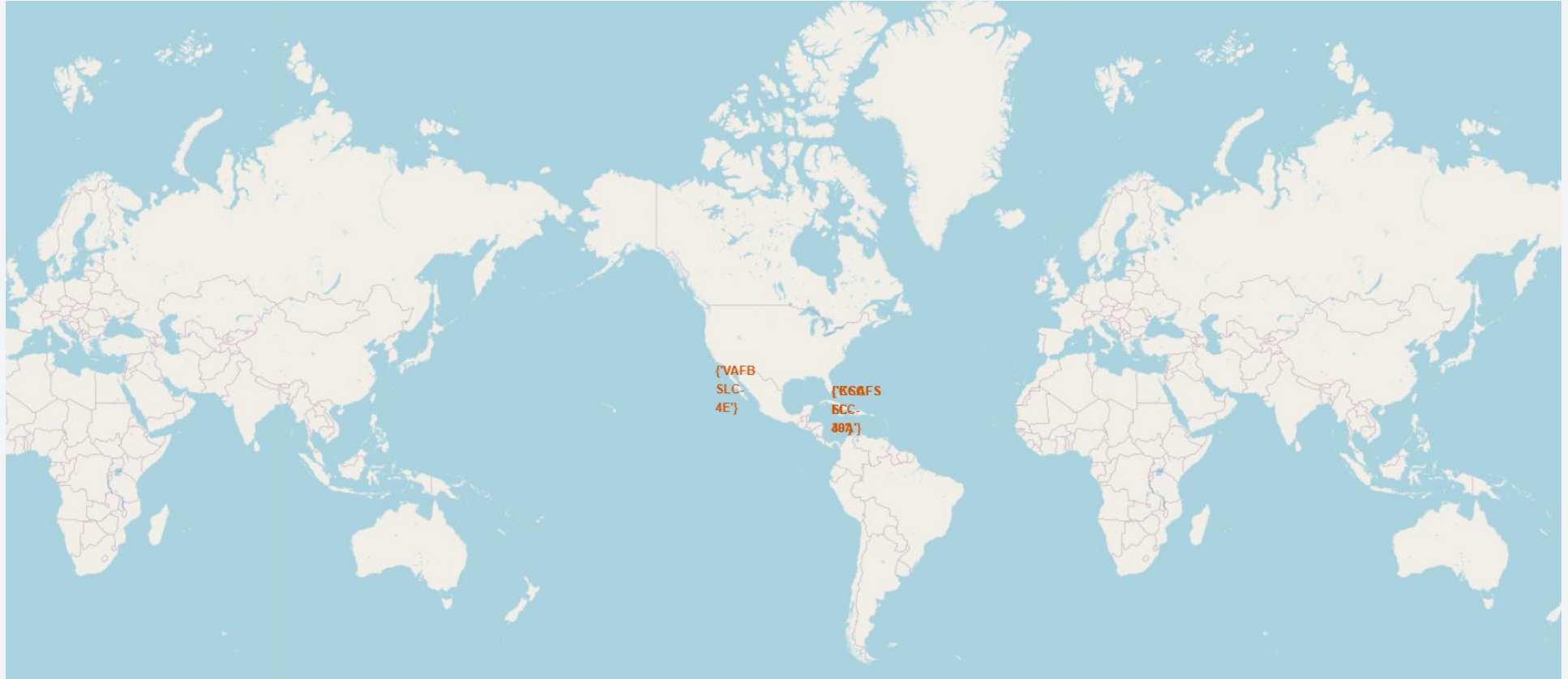
Section 4

# Launch Sites Proximities Analysis

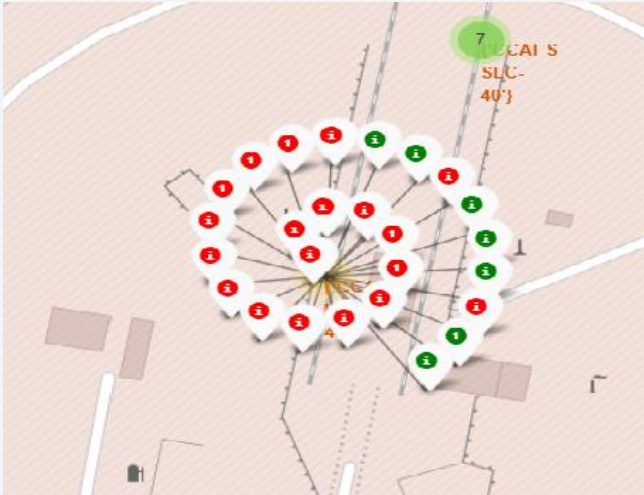
# Launch site on a global map

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- All SpaceX launch sites are in the U.S. continental states (Florida, California)



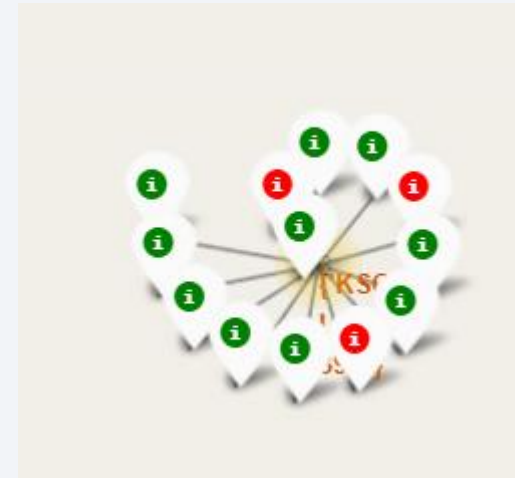
# Map with colored marker



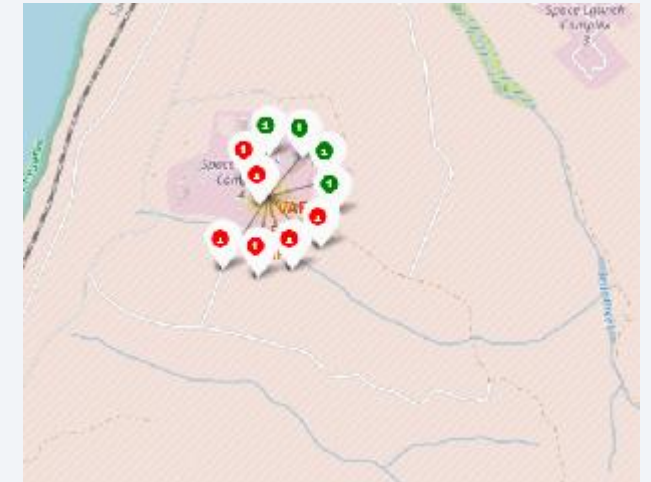
CCAFS LC-40



CCAFS SLC-40



KSC LC-39A



VAFB SLC-4E

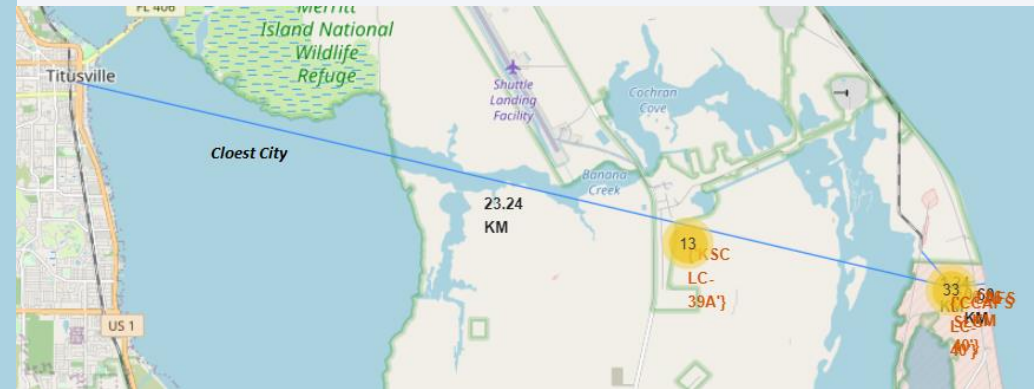
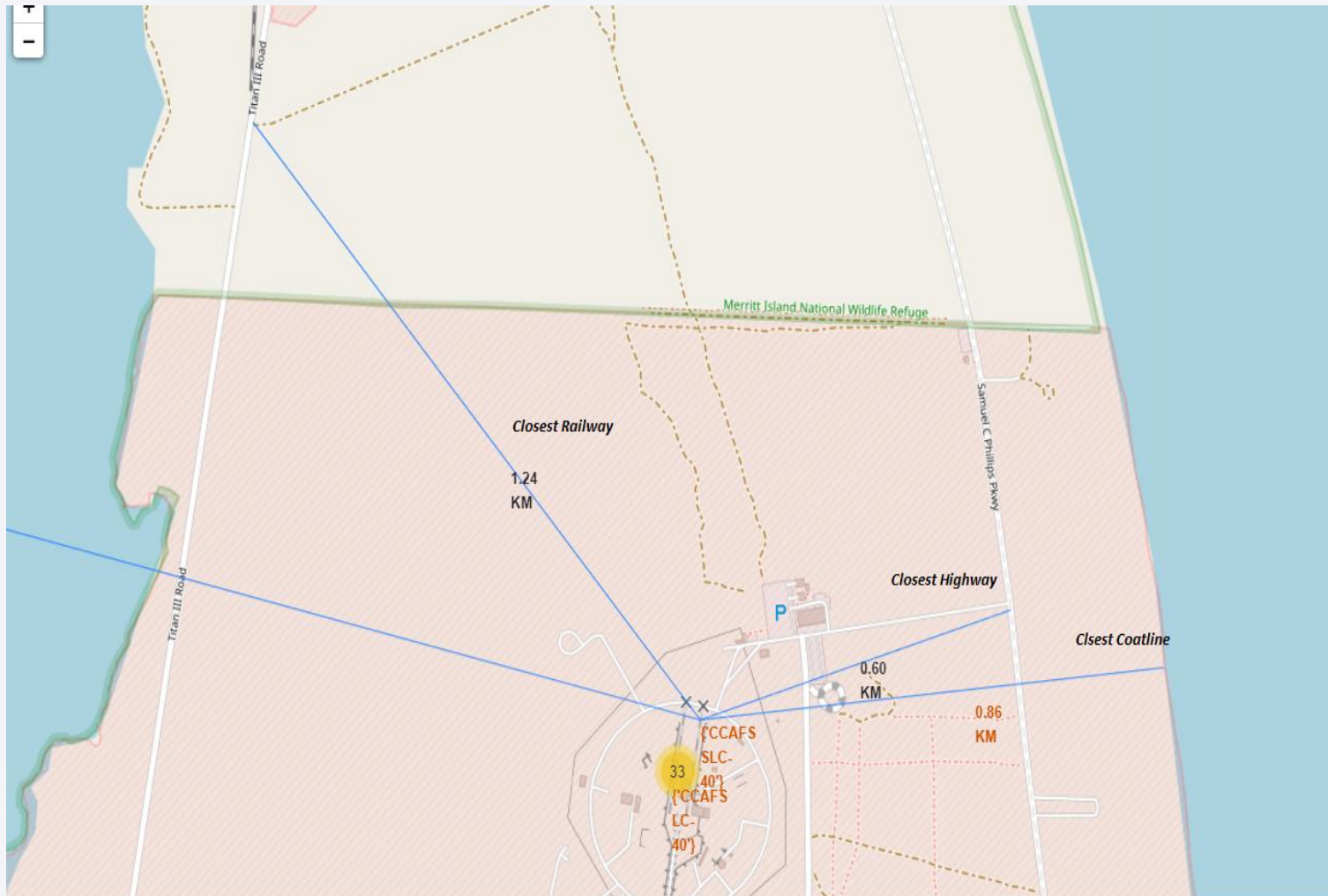
## Florida Launch Sites

## California Launch Site

- Green marker is a successful launch
- Red marker is a failed launch



# Proximities Map



## This launch site

- In close proximity to a railway (1.24km)
- In close proximity to a highway (0.6km)
- In close proximity to a coastline (0.86km)
- **NOT** in close proximity to cities (23.24km)

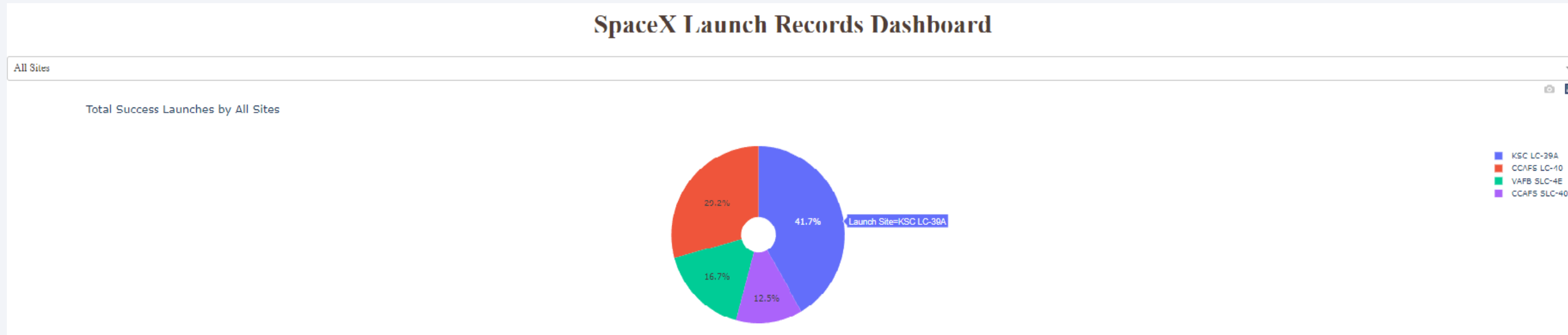


Section 5

# Build a Dashboard with Plotly Dash

# Total Success Launches by All Sites

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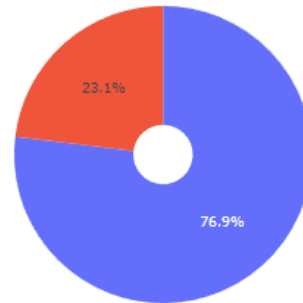


- KSC LC-39A has the most successful launches from all sites

# Launch site with highest launch success ratio

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Total Success Launches for Site → KSC LC-39A

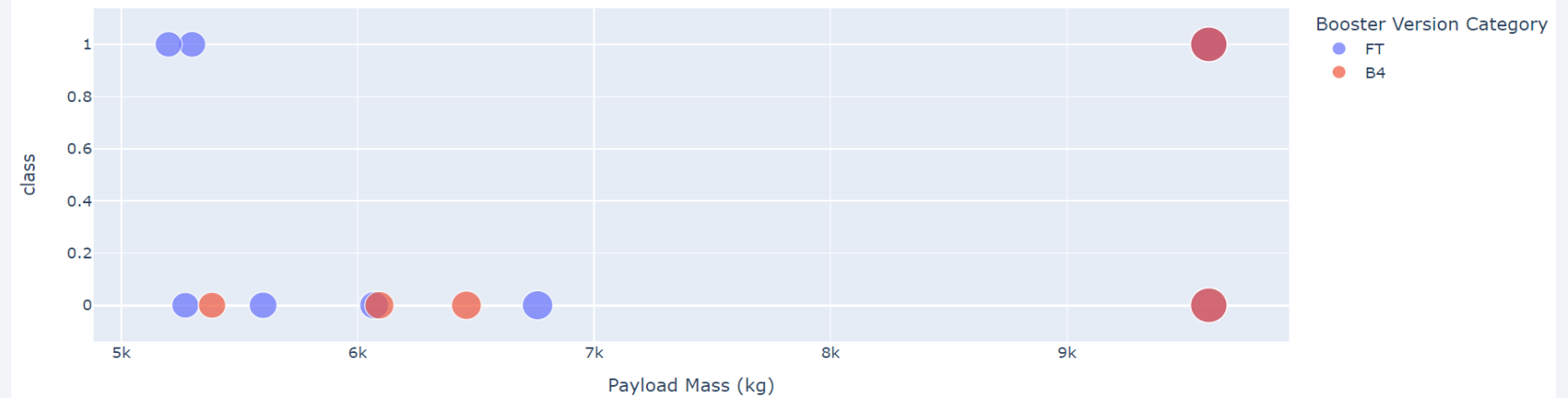


- KSC LC-39A has a success rate of 76.9%, which is the highest launch success rate

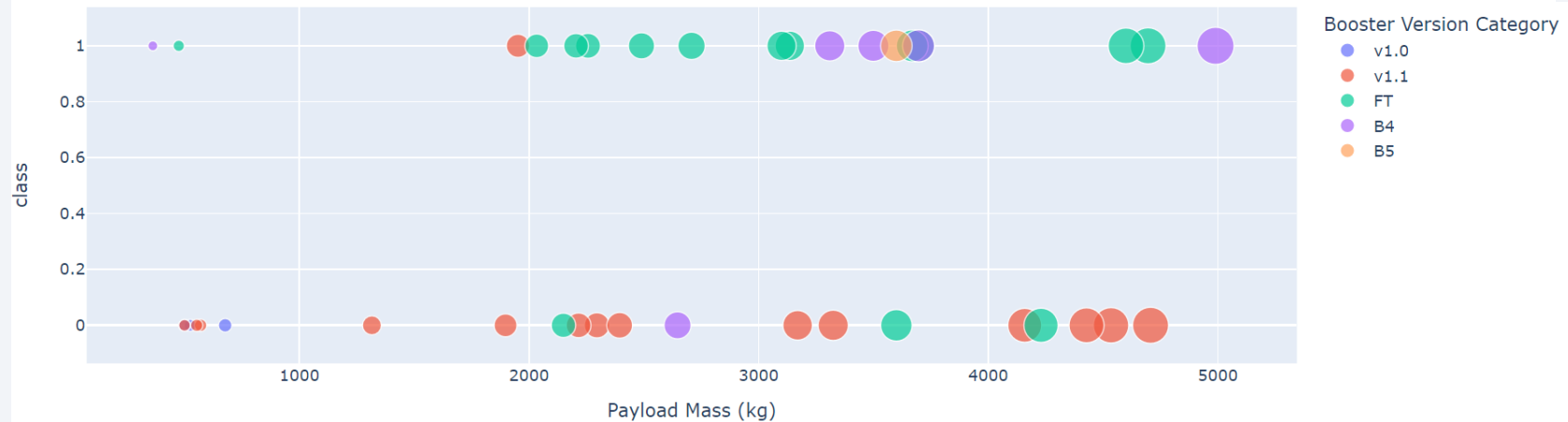


# Payload vs Launch outcome for all sites

High Payload range  
(5000kg-10000kg)  
outcome scatter chart



Low Payload range (0  
kg-5000kg) outcome  
scatter chart



- Success rates for low payload range is higher than high payload range (more dots on top)



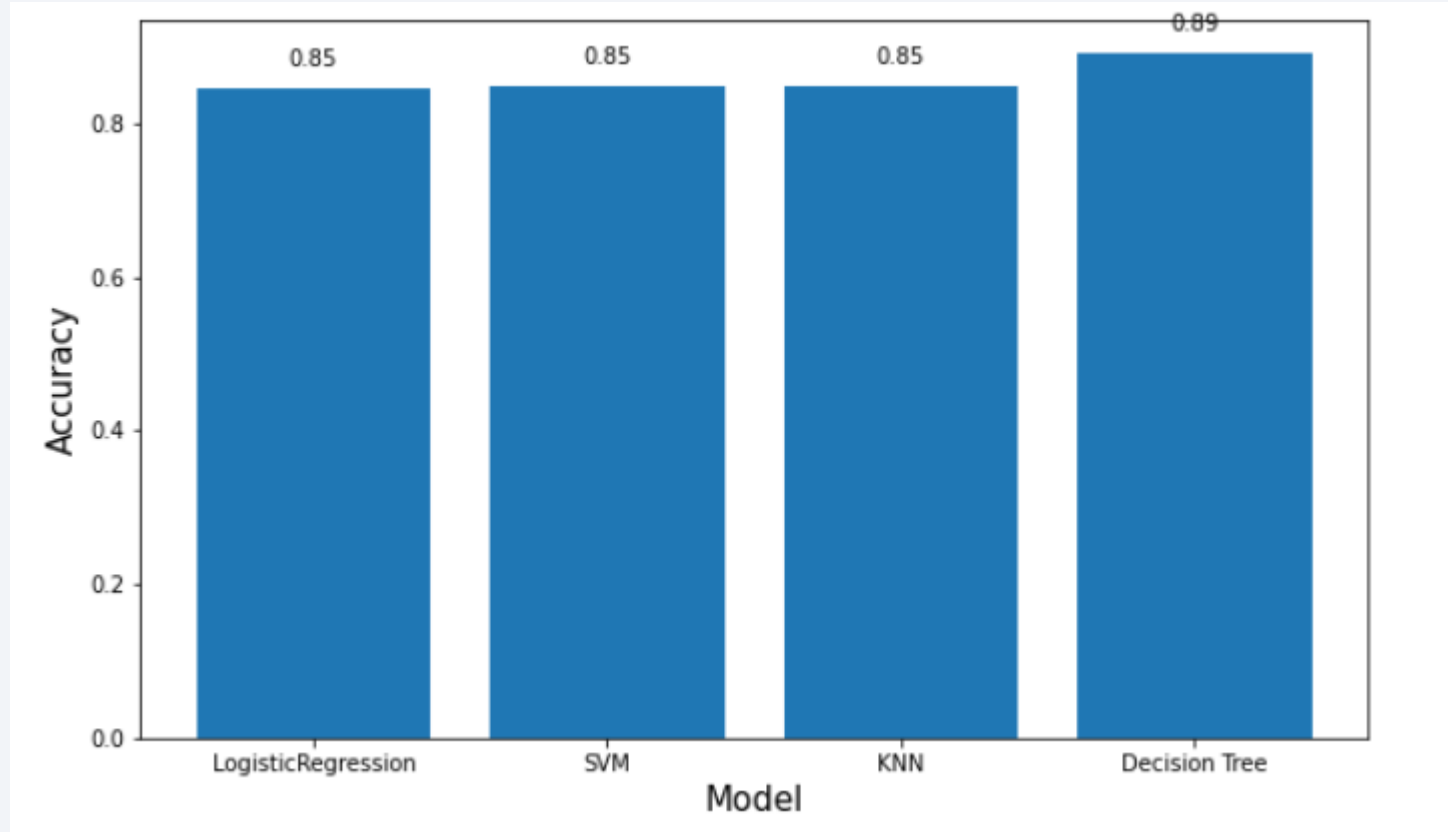
Section 6

# Predictive Analysis (Classification)

# Classification Accuracy

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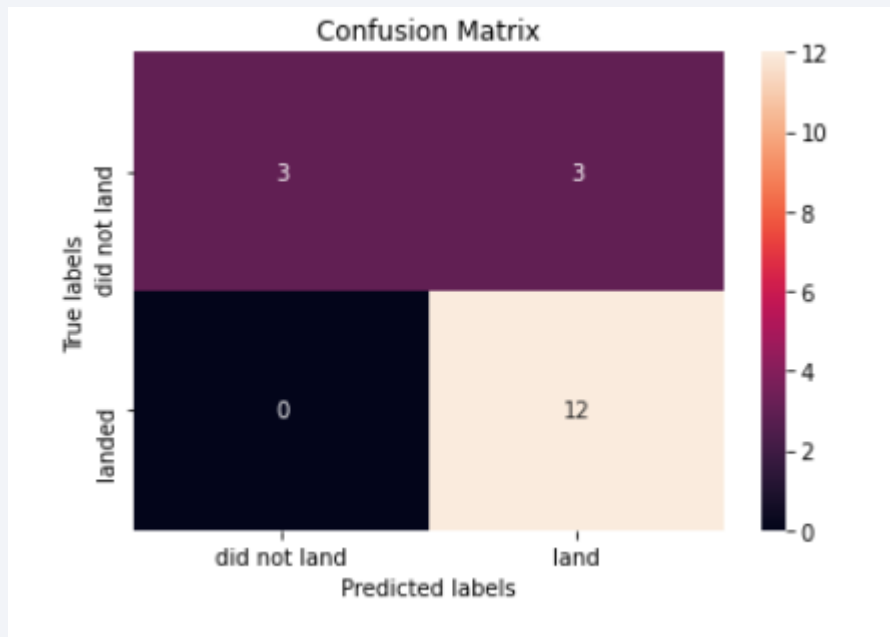
- Best model is Decision Tree with accuracy of 89.11% (83.33% on test data)



# Confusion Matrix

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- Decision Tree model shows strong positive results for predicted values



# Conclusions

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- The Decision Tree model is the best machine learning model for this dataset
- KSC LC-39A had the most success launches and higher success launch % compare to other sites
- Site with more launches have better success rate
- SpaceX success rate for launches grow exponentially with time

# Appendix

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- 2 Python SQL method used:

- SQL Magic:

```
#Method 1 with SQL Magic
```

```
%sql ibm_db_sa://mj190806:Q76g0IIy4Lo3RWZE@98538591-7217-4024-b027-8baa776ffad1.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud:30875/bludb?security=SSL
```

```
%sql select * from SPACEXDATASET
```

- IBM DB:

```
#Method 2 with ibm db
```

```
import pandas as pd
import ibm_db
import ibm_db_dbi
```

```
dsn_hostname="00000000-0000-0000-0000-000000000000.c3n41cmd0nqnrk39u98g.databases.appdomain.cloud"
dsn_uid="mj190806"
dsn_pwd="Q76g0IIy4Lo3RWZE"
```

```
dsn_driver="{IBM DB2 ODBC DRIVER}"
dsn_database="bludb"
dsn_port="30875"
dsn_protocol="TCPIP"
```

```
dsn = (
    "DRIVER={0};"
    "DATABASE={1};"
    "HOSTNAME={2};"
    "PORT={3};"
    "PROTOCOL={4};"
    "UID={5};"
    "PWD={6};"
    "Security=ssl;").format(dsn_driver, dsn_database, dsn_hostname, dsn_port, dsn_protocol, dsn_uid, dsn_pwd)
```

```
try:
    conn = ibm_db.connect(dsn, "", "")
    print ("Connected to database: ", dsn_database, "as user: ", dsn_uid, "on host: ", dsn_hostname)
```

```
except:
    print ("Unable to connect: ", ibm_db.conn_errormsg() )
```

```
pd_conn=ibm_db_dbi.Connection(conn)
```

```
QUERY = "select *from SPACEXDATASET"
df=pd.read_sql(QUERY,pd_conn)
df.head()
```

Thank you!

