

# SpaceX- Falcon 9 series- Data Analysis

IBM Data Science Final Project

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# Executive Summary

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In a highly competitive business environment:

- SpaceX has revolutionized the launch of satellites with a concept of reusable launcher/booster: Falcon9/Falcon 9 Heavy
- The main advantage of this concept is the significant reduction in cost per kg.
- Reliability problems remain compared to classic launch vehicles like Soyuz or Ariane-5.
- For maintaining the “low cost” competitive advantage, compared with classic launchers, Falcon9 mission success is defined as the successful recovery or landing of the booster.
- Falcon9 booster successful recovery depends on features such as:
  - orbit
  - payload mass
  - booster versions
  - Launching sites...
- Based on these features, the best Machine Learning supervised classification model developed in this report, predicted booster recovery outcome with an accuracy close to 94%.

## Introduction - Business understanding

1. **\*\*High Capital Investment\*\***: Rocket launches demand massive upfront costs for development and infrastructure.  
Reusable rockets like SpaceX's Falcon 9 (~\$60-90M/launch) drastically cut per-launch expenses.
2. **\*\*Revenue Streams\*\***: Income stems from satellite launches, crewed missions, and space tourism.  
Contracts like NASA's Artemis and internal projects like Starlink ensure steady cash flow.
3. **\*\*Market Dynamics\*\***: The \$14.7B launch market grows with demand for satellites and constellations.  
SpaceX's low pricing (~\$6,000/kg) disrupts traditional and state-backed competitors.
4. **\*\*Cost Drivers and Risks\*\***: Fuel, labor, and regulations drive costs; failures risk \$100M+ losses.  
Reusability reduces expenses but adds maintenance and geopolitical supply chain risks.
5. **\*\*Economic Impact and Scalability\*\***: Launches fuel a \$447B space economy via satellite services.  
Scalability depends on reusability, standardization, and high launch frequency.

# Methodology

# Methodology

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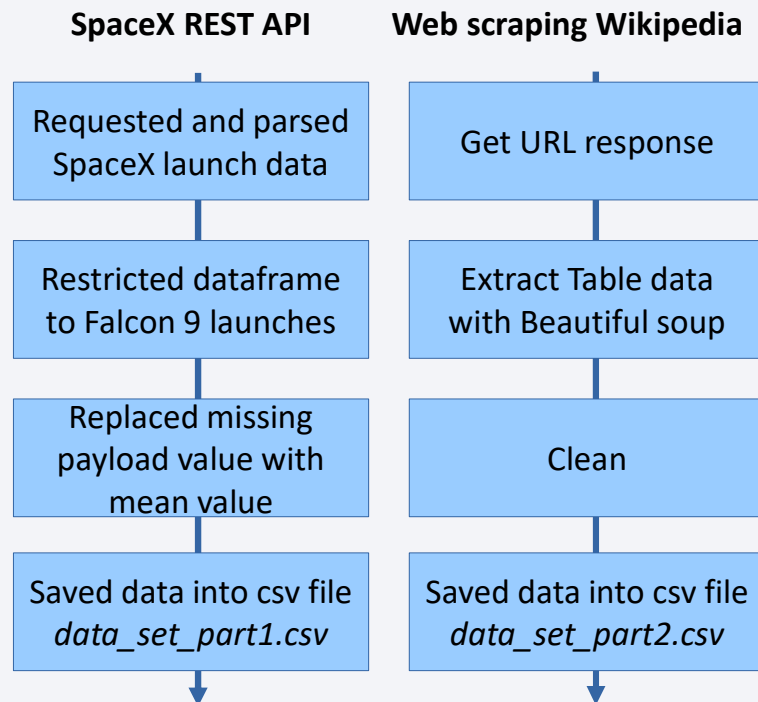
## Methodology steps:

1. **Data collection from open Data base and Wikipedia (Falcon9, Ariane-5).**
2. **Data wrangling.**
3. **Exploratory data analysis using SQL query and visualization of correlation between parameters.**
4. **Visual analytics: launch sites with Folium, success rates with Plotly Dash.**
5. **Classification Models development and validations. Selection of best predictive model.**

# Data Collection

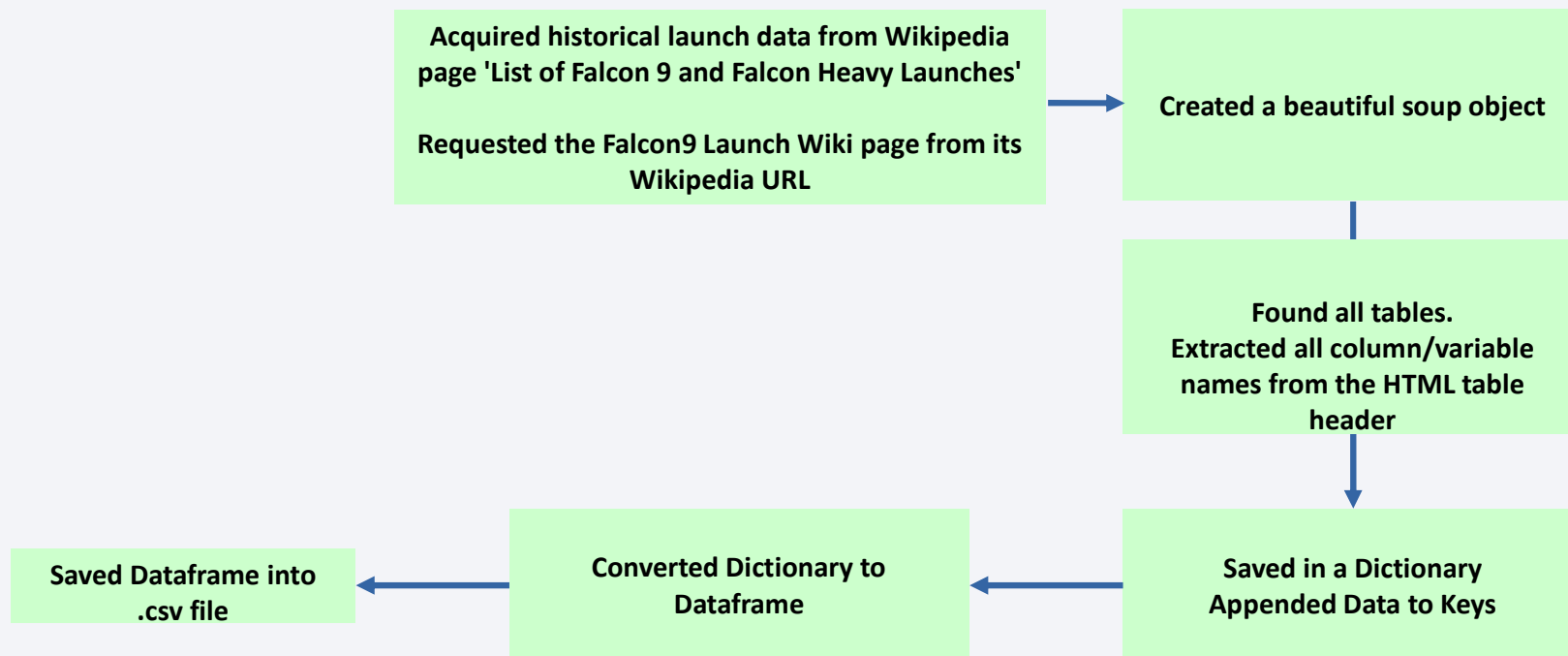
Data was collected from:

- open source SpaceX REST API
- webscraping Falcon9 launch data in Wikipedia
- webscraping Ariane5 launcher data in Wikipedia *for reference graphs only.* (not prepared for ML)



# Data Collection - Scraping

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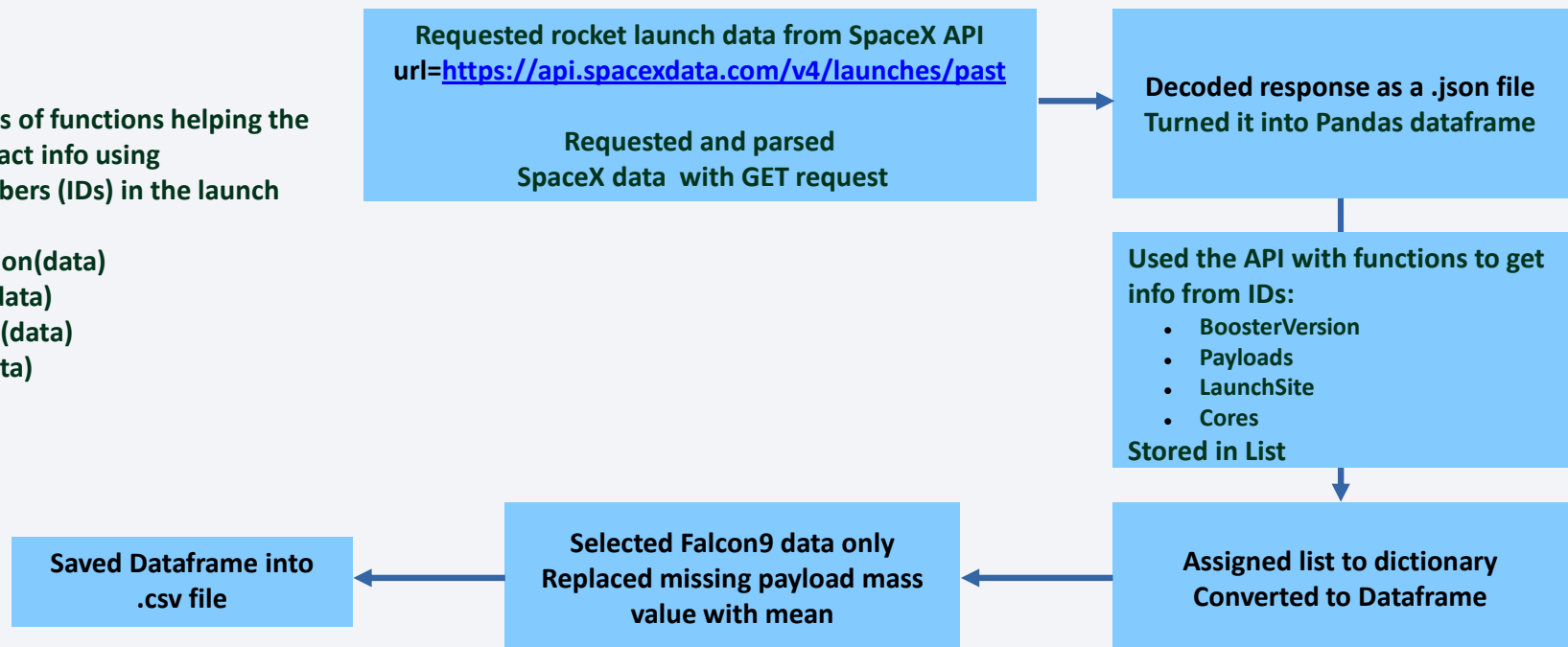


# Data Collection – SpaceX API

## Preliminary Stage

We defined a series of functions helping the use the API to extract info using identification numbers (IDs) in the launch data.

- `getBoosterVersion(data)`
- `getLaunchSite(data)`
- `getPayloadData(data)`
- `getCoreData(data)`



# Data Wrangling

Dataframe  
From SpaceX API

Identified missing values

Replaced missing "PayloadMass"  
value with mean value

SpaceX dataset  
Further Data wrangling

Identified:

- null values for each feature
- numerical and categorical features

Calculated:

- Number of launches on each site
- number and occurrence of each orbit
- number and occurrence of missions outcome per orbit type

Created a set of 1 stage booster landing outcomes

- 0 True ASDS: successful landing on a drone ship
- 1 None None: failure to land
- 2 True RTLS: successful landing to a ground pad
- 3 False ASDS: failed landing on a drone ship
- 4 True Ocean: successful landing, specific region of the ocean
- 5 False Ocean: failed landing, specific region of the ocean
- 6 None ASDS: failure to land
- 7 False RTLS: failed landing to a ground pad

Created a Training label: 'Class'

**Class = 0: booster landing failure**

**Class = 1: booster landing success**

	FlightNumber	Date	BoosterVersion	PayloadMass	Orbit	LaunchSite	Outcome	Class
49	50	2018-05-11	Falcon 9	3750.00	GTO	KSC LC 39A	True ASDS	1
47	48	2018-04-02	Falcon 9	2760.00	ISS	CCAFS SLC 40	None None	0
50	51	2018-06-04	Falcon 9	5383.85	GTO	CCAFS SLC 40	None None	0
44	45	2018-01-31	Falcon 9	4230.00	GTO	CCAFS SLC 40	True Ocean	1
11	12	2015-01-10	Falcon 9	2395.00	ISS	CCAFS SLC 40	False ASDS	0

# EDA with SQL

For the reviewer, for emulating SQL queries:

- I created a MySQL local server
- I created a database: 'spacex\_database'
- I saved SpaceX csv file in Table "spacex\_v11" (see picture)
- I established a connection "conn" with "spacex\_database".

```
conn=pymysql.connect(host='localhost',  
                    port=int(3306),  
                    user='root',  
                    passwd="#####",  
                    db='spacex_database' )
```

We ran SQL queries with "pd.read\_sql\_queries":

```
df=pd.read_sql_query("select * from spacex_v11 ",conn)  
df.head(5)
```

1. to download data in a Jupyter Notebook (dataframe)
2. to display information about:
  - Launch sites
  - Payload mass
  - Booster versions
  - Mission outcomes
  - Booster landings

MySQL local server  
Building "spacex\_v11" table.

The screenshot shows the MySQL Workbench interface. On the left, the 'SCHEMAS' pane shows the 'spacex\_database' with a table 'spacex\_v11'. The main editor shows the SQL command to create the table 'spacex\_v11' with columns: id (INT), Date (DATE), Time (UTC) (TIME(1)), Booster\_Version (VARCHAR(45)), Launch\_Site (VARCHAR(45)), Payload (VARCHAR(100)), PAYLOAD\_MASS\_KG (INT), Orbit (VARCHAR(45)), Customer (VARCHAR(100)), Mission\_Outcome (VARCHAR(45)), and Landing\_Outcome (VARCHAR(45)). Below the editor, the 'Result Grid' shows the first 9 rows of data. The 'Output' pane at the bottom shows the execution of the query, indicating that 101 row(s) were returned.

id	Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG	Orbit	Customer
1	2010-04-06	18:45:00.0	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX
2	2010-08-12	15:43:00.0	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of...	0	LEO (ISS)	NASA (COT)
3	2012-05-22	07:44:00.0	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COT)
4	2012-10-08	00:35:00.0	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)
5	2013-03-01	15:10:00.0	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)
6	2013-09-29	16:00:00.0	F9 v1.1 B1003	VAFB SLC-4E	CASSIOPE	500	Polar LEO	MDA
7	2013-12-03	22:41:00.0	F9 v1.1	CCAFS LC-40	SES-8	3170	GTO	SES
8	2014-01-06	22:06:00.0	F9 v1.1	CCAFS LC-40	Thaicom 6	3325	GTO	Thaicom
9	2014-04-18	19:25:00.0	F9 v1.1	CCAFS LC-40	SpaceX CRS-3	2296	LEO (ISS)	NASA (CRS)

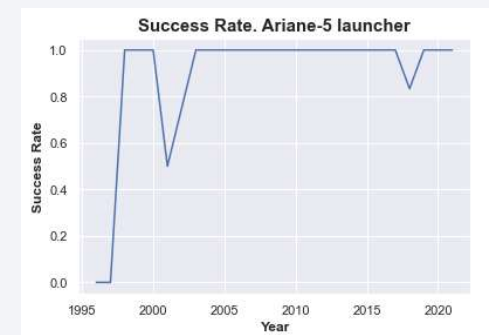
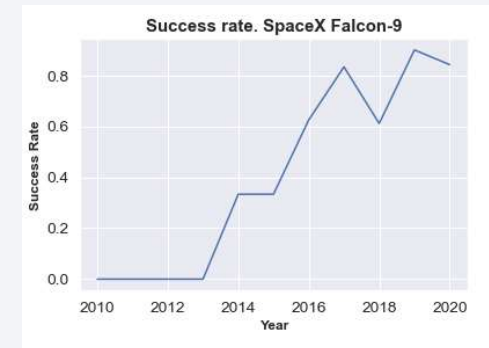
# EDA with Data Visualization

Exploring data in the Falcon 9 dataframe searching for factors and relations influencing launching success rate (booster recovery).

- Payload mass
- Orbit type
- Launch site

Graphs and scatter charts with Matplotlib – Seaborn and Analysis. Results with Scatter charts are labeled: class 0-1 (failure/success).

- Payload mass v. Flight Number
- Launch Site v. Flight number
- Launch Site v. Payload mass
- Orbit v. Flight number
- Orbit v. Payload mass
- Histogram: success rate for each orbit
- Falcon 9 & Ariane-5 launch success yearly trend.



# Build an Interactive Map with Folium

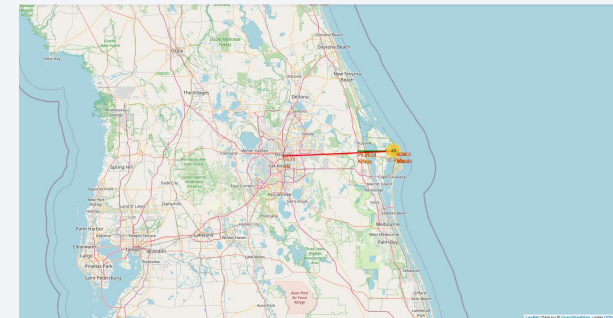
Launch success rate may depend on the location and proximity of a launch site. **Folium Interactive Map was used for visualizing and analyzing SpaceX Launch Sites.**

- Used Interactive mapping library called Folium
- Identified all SpaceX launch sites on a map: Florida, California
- Included longitude and latitude info.
- Identified successful/failed launches for each site on map

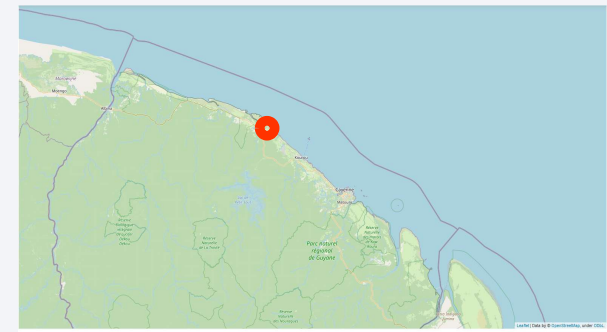
Calculated the distance between a launch site (CCAFS\_SLC40 in Cape Canaveral, FL ) and:

- Closest coastline
- Closest high traffic density railway: Florida East Coast Railway
- Closest high traffic density highway: Interstate I95
- Closest high density urban area: Orlando (FL)

For reference, we added the localization of European Space Agency (ESA) /ArianeEspace Ariane 5 and Soyuz launch pads in Kourou, French Guiana.



CCAFS\_SLC40 in Cape Canaveral FL  
Coordinates: -80.577°, 28.563°



Ariane launch pad - Kourou in French Guiana  
Coordinates: -52.792°, 5.265° (~ Equator)

[https://github.com/basheergit/coursera/blob/basheergit-spacex-data-analysis-project/5\\_Data\\_Visualization\\_for\\_Launch\\_Sites.ipynb](https://github.com/basheergit/coursera/blob/basheergit-spacex-data-analysis-project/5_Data_Visualization_for_Launch_Sites.ipynb)

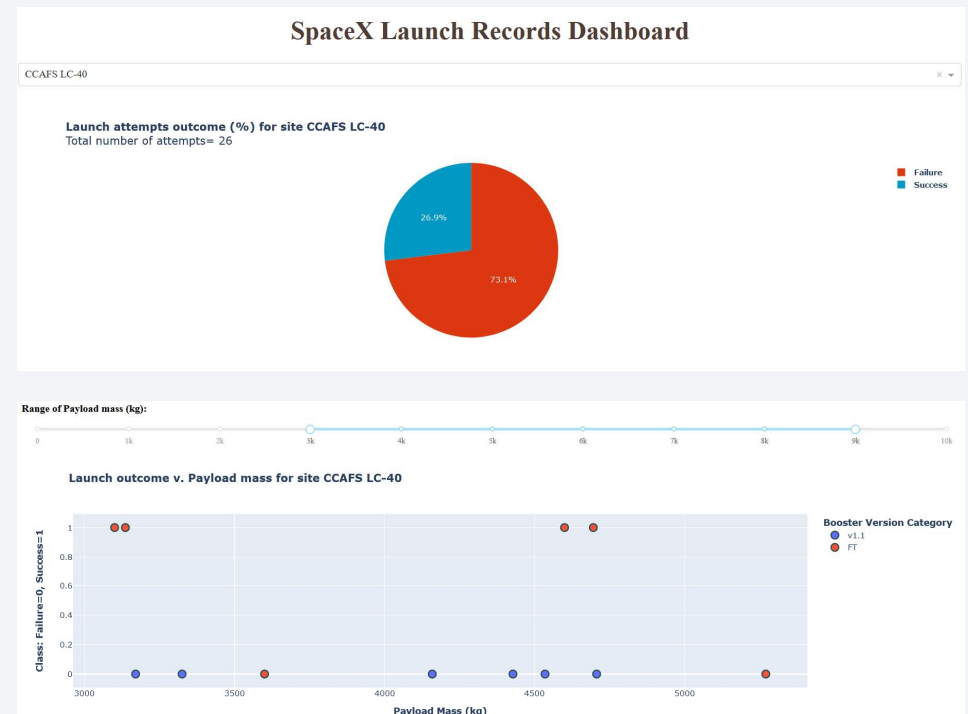
# Build a Dashboard with Plotly Dash

We built an interactive dashboard with Plotly including:

- Dropdown menu for selecting launch sites
- Pie charts displaying success rate.
- Scatter chart displaying launch site, payload mass, success/failure
- Range slider for selecting range of payload mass (kg).

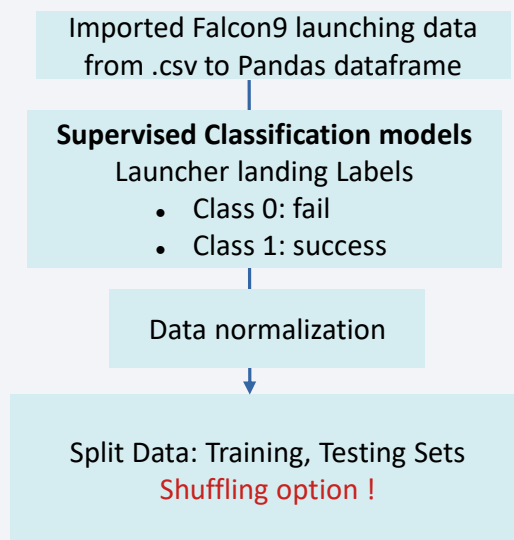
for analyzing SpaceX launch records features:

- site with largest successful launches.
- site with highest launch success rate
- payload range(s) with highest launch success rate
- payload range(s) with lowest launch success rate
- F9 Booster version (v1.0, v1.1, FT, B4, B5, etc.) with highest launch success rate.



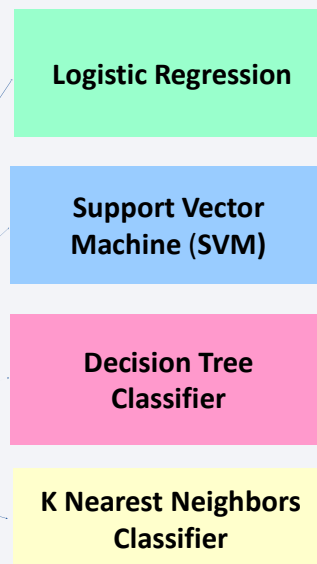
# Predictive Analysis (Classification)

## Model Training

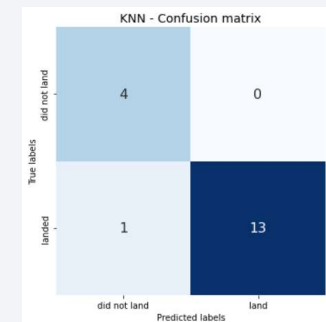


### Note to the reviewer:

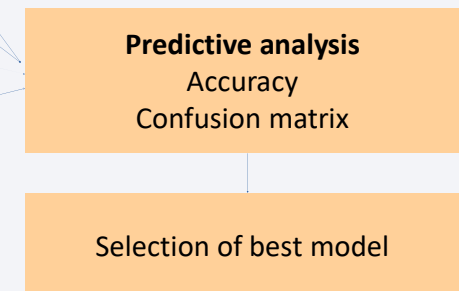
Data was shuffled in the split data process. For making results a bit more interesting  
“train\_test\_split(X, Y, test\_size=0.2, random\_state=3)”



Optimization of models hyper-parameters with Scikit-learn GridSearchCV  
Refined Optimization when possible (LR, SVM) for achieving highest accuracy with training set.



## Testing set



# Results

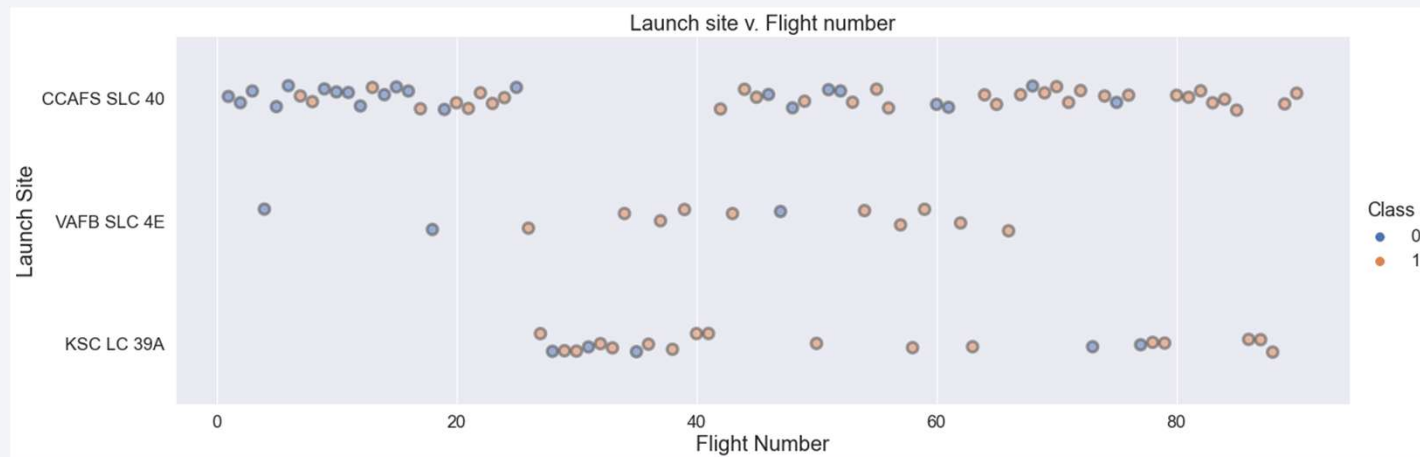
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- Exploratory data analysis results
- Interactive analytics demo in screenshots
- Predictive analysis results



# Insights from EDA

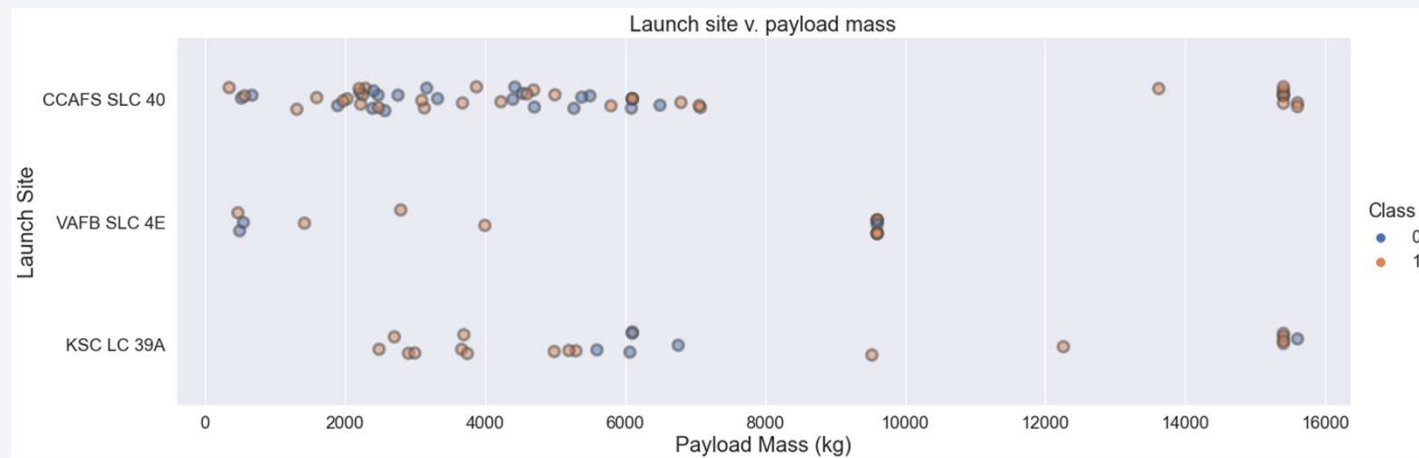
# Launch Site v. Flight Number



The chart displays valuable info about:

- Chronology: flight numbers
  - Number of flights per launch site
  - Success/Failure per launch site
- 
- Cape Canaveral CCAFS-SLC 40 is the most used launch site.
  - CCAFS-SLC 40 concentrates most of failures , particularly **in the early stage of Falcon9 project**.
  - Given CCAFS-SLC 40 southern location, most “risky” GTO and GEO launches may take place there.
  - Additional info needed: orbit, payload mass

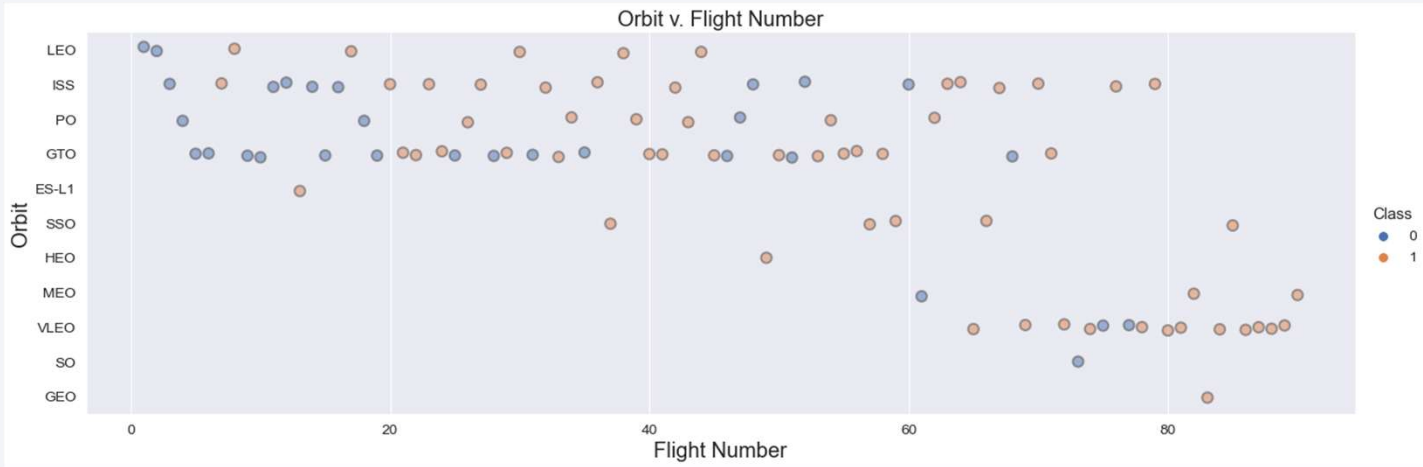
# Launch Site v. Payload



The chart brings additional info:

- Payload mass per launch site
  - Success/Failure per payload mass
- 
- Given Falcon9 specifications, heavy payloads > 10000 kg are sent to low/medium orbits LEO/MEO only.
  - It looks like the percentage of failures is lower for heavy payload. Which would indicate that low orbits are less risky to the success of the mission (recovery of booster).
  - Light payloads are not necessarily all sent to GTO/GEO.
  - More information is needed for extracting some correlation: success rate v. payload/orbit

# Orbit Type v. Flight Number



The chart brings additional info:

- Number of flights per Orbit.
- Success rate per orbit
- The number of flights for: GEO, SO, HEO, ESL-1, MEO is not significant for concluding about success rate.
- PO, SSO, ISS, VLEO are low orbits
- GTO is a transfer orbit to GEO.

It looks like GTO are higher risk missions, low orbits are lower risk.

We confirm with the following histogram.

# Success Rate vs. Orbit Type

## Remarks:

- GTO is a transfer orbit to GEO. Low thrust engines of the payload (satellite) complete the orbiting phase.
- We ignore results: GEO, SO, HEO, ESL-1, MEO. The number of flights is not significant.

**GTO sees the lowest success rate as suggested in previous slide.  
SSO (polar low orbit) the highest one.**

## Success rate may strongly depend on both:

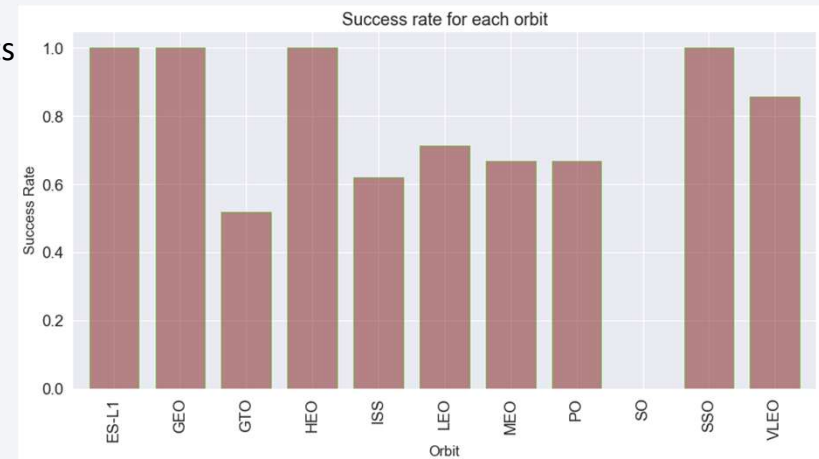
- payload mass
- orbit.

**meaning the amount of energy deployed at lift-off**, that may induce **strong noise/vibrations that are known to damage satellites\***.

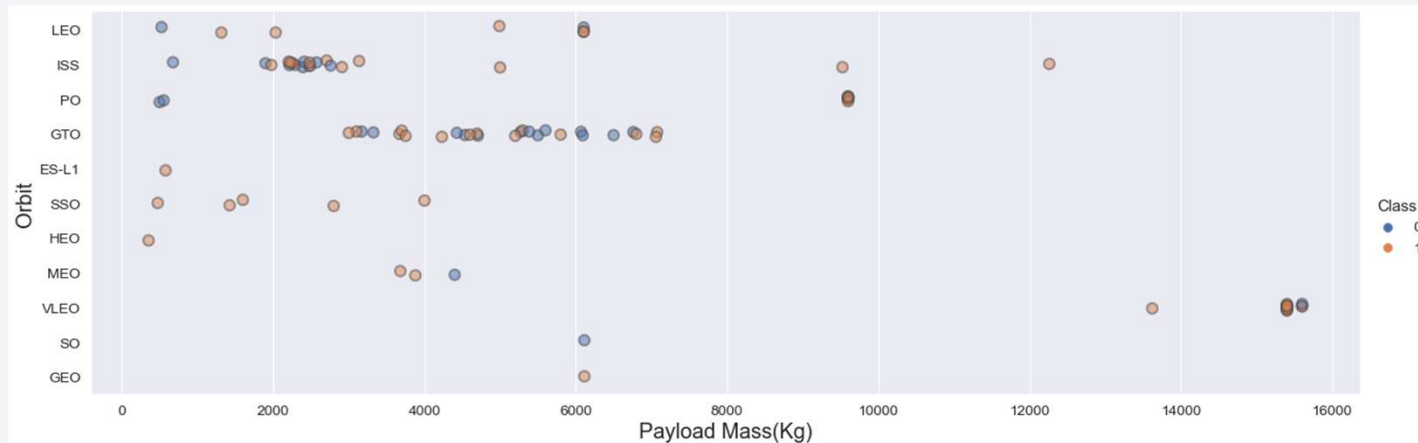
Vibrations could damage some of the booster electronics, inertial guidance systems... and cause booster recovery/landing failure.

We also need additional info about payload mass/orbit. Fortunately it is available.

\* <https://adsabs.harvard.edu/full/1996ESASP.386..237F>



# Orbit Type v. Payload



The chart brings final info about “Orbit v. Payload”. It describes the distribution “success rate v. (payload, orbit)”  
Main trends:

- Maximum success rate with: low orbit except (ISS) and low payload mass
- ISS: based on “[Orbit Type v. Flight Number](#)” 5/8 failures occurred in the early stage of Falcon 9 project. When Falcon 9 reliability was low.
- Between 2000 and 7500 kg, success rate seems to be evenly distributed for GTO.
- Independently of payload mass, GTO is a risky “orbit” affecting missions success rate. Falcon 9 reliability improves over time, but there are still recent failed booster recovery after GTO launches.

## Total Number of Successful and Failure Mission Outcomes

---

Total number of successful and failure mission outcomes (from SQL queries).

Here success is defined based on properly launching/orbiting payload. Success rate is very high: ~99% like Ariane-5.

Nevertheless, Falcon9 maintains a competitive advantage in terms of cost per kg compared with classic launchers like Ariane-5, **only if the reusable booster is recovered**.

Therefore Falcon9 “success rate” in this report is defined after successful booster recovery (landing).

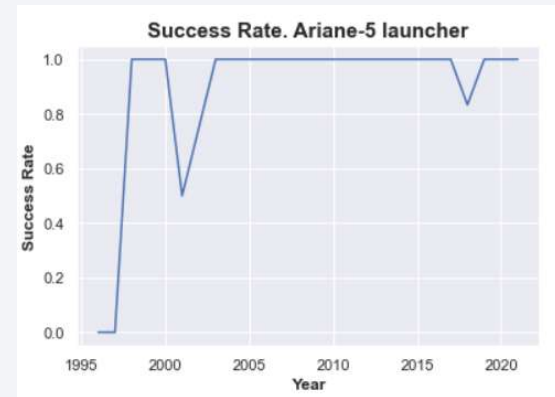
```
# sql query
qsf= """Select (Select Count(Mission_Outcome) from spacex_v11 where Mission_Outcome like '%Success%')
as Successful_Missions,
(Select Count(Mission_Outcome) from spacex_v11
where Mission_Outcome like '%Failure%') as Failed_Missions """

success_failure= pd.read_sql_query(qsf,conn)
print(success_failure)
```

	Successful_Missions	Failed_Missions
0	100	1

# Launch Success Yearly Trend

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Falcon 9 reliability significantly improves over time .

Success rate, **here defined after successful booster recovery for Falcon9**, depends on:

- Payload mass
- Orbit
- + other factors we investigate next
- Independently of payload mass, orbits, **Ariane 5 has a close to 100% success rate** for 82 flights since 2003.
- Falcon9 average booster recovery success rate is 66%.
- Success rate currently sufficient for SpaceX financial viability.



# **Launch Sites and Proximities Analysis**

# Launch Site Names & Records

Before starting launch sites analysis, we list the names of all launch sites and some launch records (from SQL queries).

```
df_unique_launchsites=pd.read_sql_query("Select distinct Launch_Site from spacex_v11 ",conn)
print(df_unique_launchsites)
```

```
Launch_Site
0  CCAFS LC-40
1  VAFB SLC-4E
2  KSC LC-39A
3  CCAFS SLC-40
```

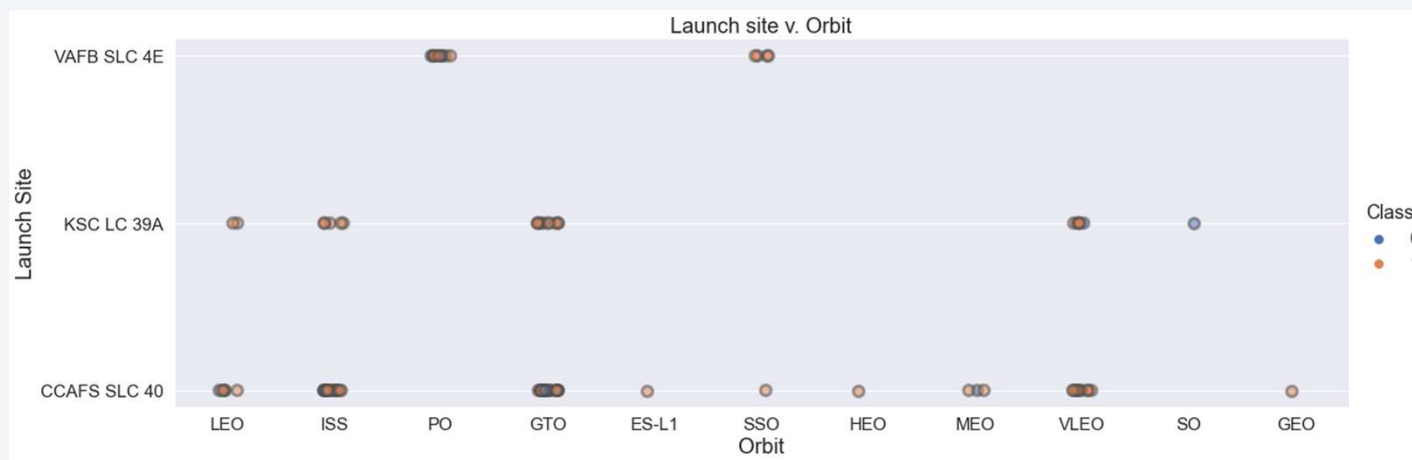
There are 4 distinct launch sites

5 records where launch sites begin with `CCA`

```
df_launchsites_CCA5=pd.read_sql_query("Select * from spacex_v11 where Launch_Site Like 'CCA%' Limit 5",conn)
df_launchsites_CCA5
```

	id	Date	Time (UTC)	Booster_Version	Launch_Site	Payload	PAYLOAD_MASS_KG_	Orbit	Customer	Mission_Outcome	Landing_Outcome
0	1	2010-04-06	0 days 18:45:00	F9 v1.0 B0003	CCAFS LC-40	Dragon Spacecraft Qualification Unit	0	LEO	SpaceX	Success	Failure (parachute)
1	2	2010-08-12	0 days 15:43:00	F9 v1.0 B0004	CCAFS LC-40	Dragon demo flight C1, two CubeSats, barrel of...	0	LEO (ISS)	NASA (COTS) NRO	Success	Failure (parachute)
2	3	2012-05-22	0 days 07:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
3	4	2012-10-08	0 days 00:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1	500	LEO (ISS)	NASA (CRS)	Success	No attempt
4	5	2013-03-01	0 days 15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2	677	LEO (ISS)	NASA (CRS)	Success	No attempt

# Launch site v. Orbit type



We added that scattered chart, which has implications in the analysis of launch sites.

- VAFB SLC 4E launch site in California, is used only for PO and SSO orbits, **meaning low altitude polar orbits**. North or south? bound flights over the ocean.
- **All GTO/GEO launches take place in Florida launch sites**. As close as possible to the equator where earth tangential velocity is maximum, acting like a slingshot at lift off, from west to east over the ocean. (see gif). **This helps with orbiting GEO satellites with less energy.**

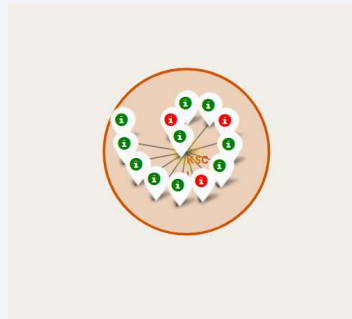
# SpaceX: All launch sites



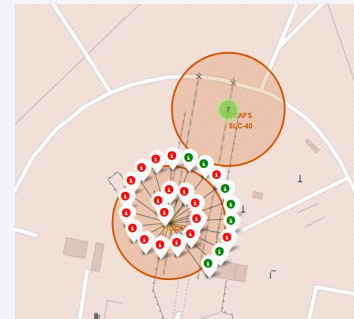
# Falcon 9 Success/Failed launches for each site



Vandenberg Space Launch Complex 4 (CA)  
VAFB SLC-4E



Kennedy Space Center (FL)  
KSC LC 39A



Cape Canaveral (FL)  
CCAFS-LC40



Cape Canaveral (FL)  
CCAFS-SLC40

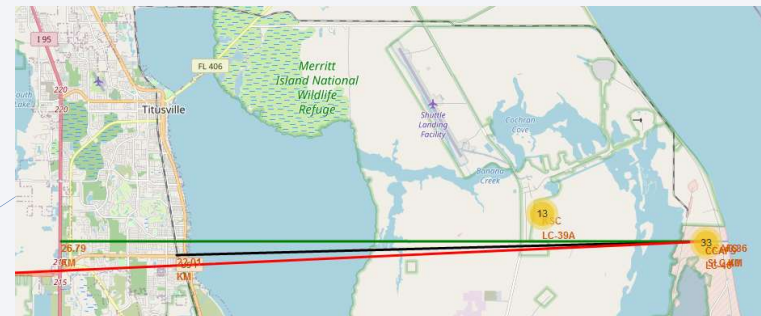
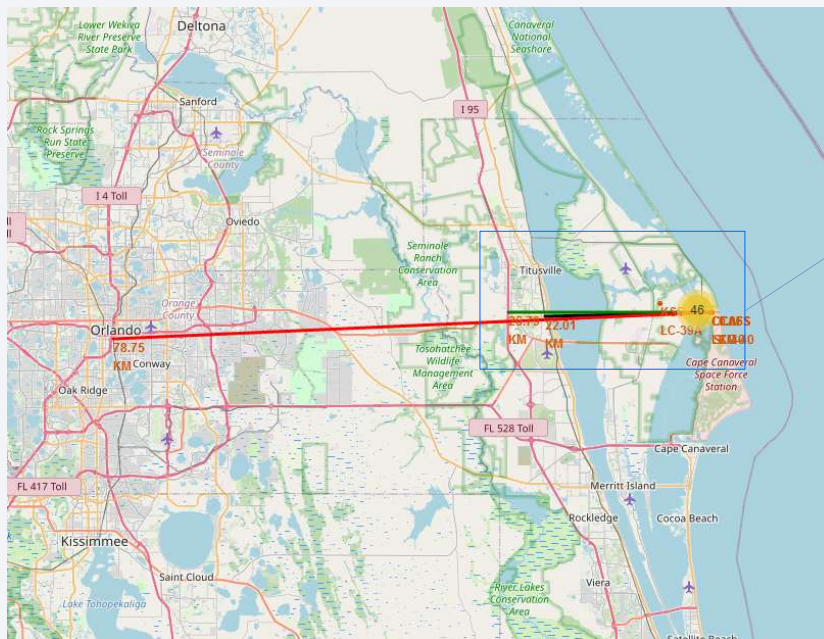
Launch Site	class	
CCAFS LC-40	0	19
	1	7
CCAFS SLC-40	0	4
	1	3
KSC LC-39A	0	3
	1	10
VAFB SLC-4E	0	6
	1	4

**Table: Synthesis of launches outcomes**

Class 0= failure

Class 1= success

# Distances between a launch site to its proximities



Distance from CCAFS\_SLC40 to:

- Closest coast: ~900 m
- Florida East Coast Railway: 22.0 km
- Highway I 95: 26.8 km
- Orlando: 78.75 km

Launch sites are close to coasts. For safety issues if launcher is lost in the early stage of the flight.

Rockets are launched:

- From West to East over the ocean in Florida.
  - North or South bound over the ocean in California. (Polar orbits only)
- Launch sites are relatively far from populated areas for protecting population from serious incidents at lift off: explosion on the launch pad.



# ESA Unique Launch site – Kourou, French Guiana.

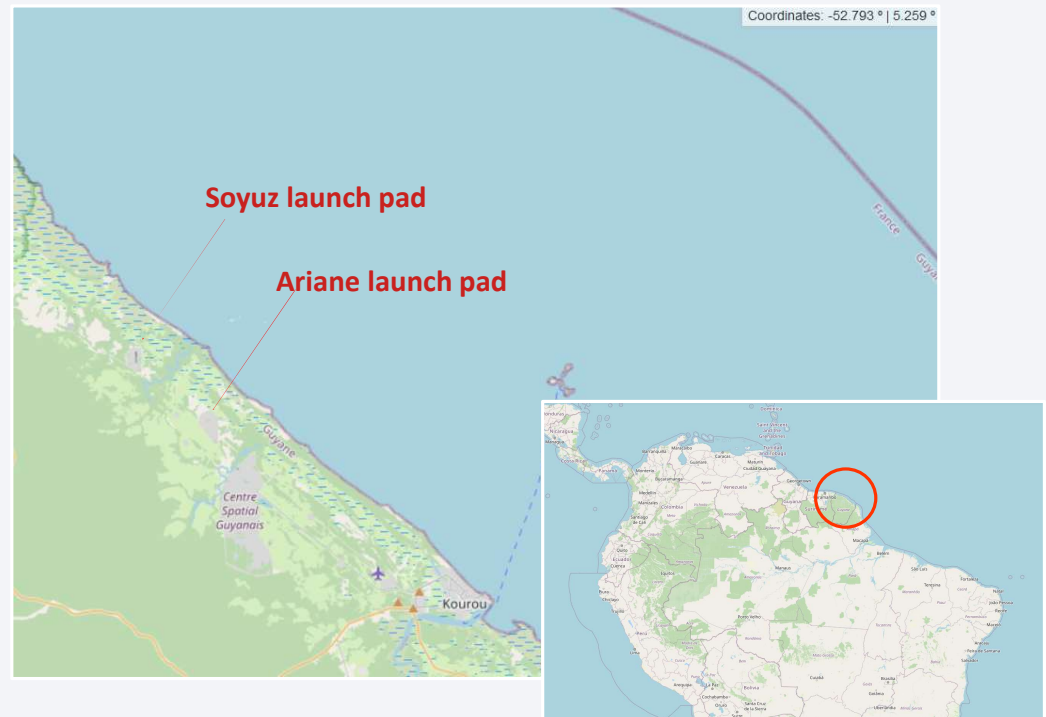
**European Space Agency (ESA)/Arianespace launch sites in Kourou, French Guiana (France).**

Kourou launch sites: Ariane, Soyuz

- are very close to the Equator: it's a strong advantage for GTO/GEO flights
- are in a remote area far from any high density inhabited area and high traffic infrastructures
- Are close to the coast

Strong Advantage over SpaceX, in terms of safety and GTO/GEO flights and energy required at lift-off.

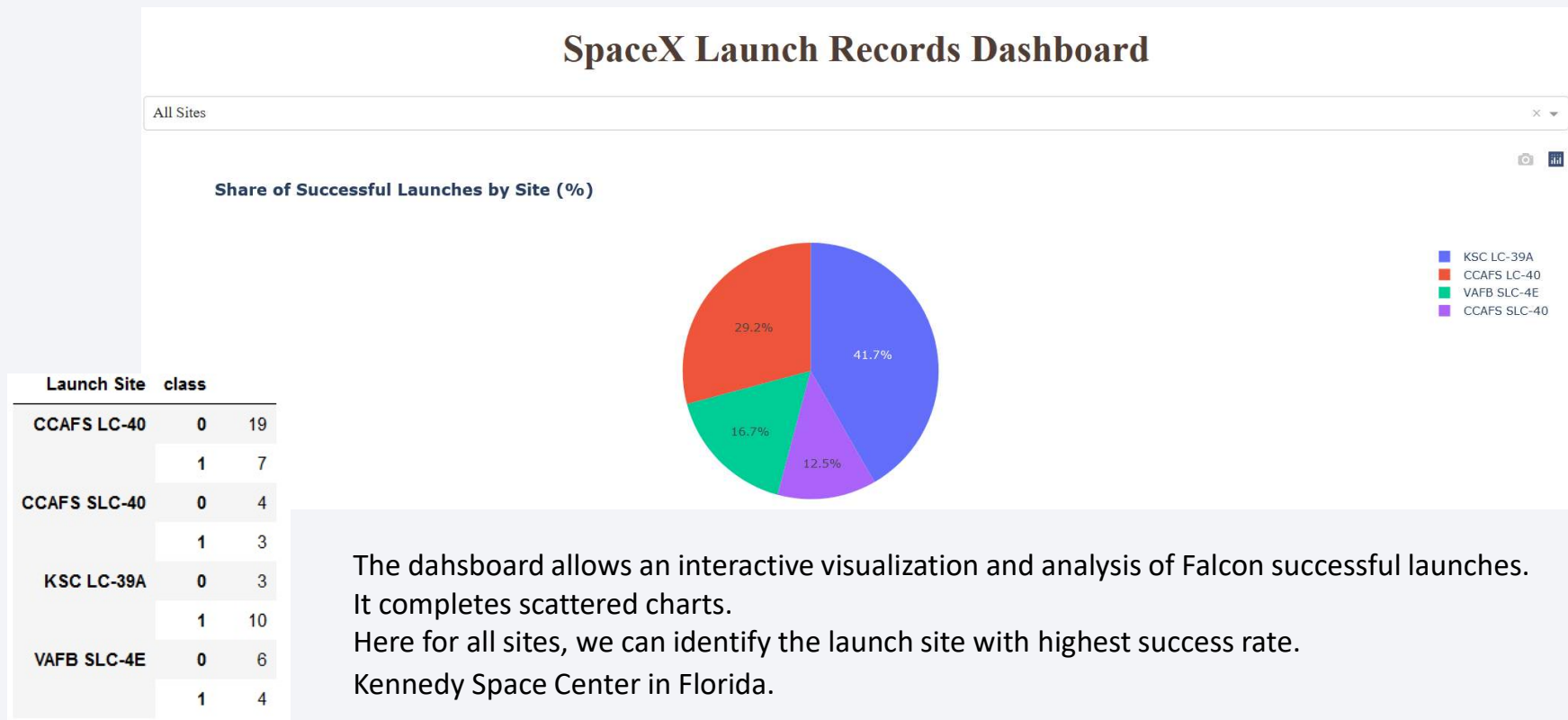
SpaceX could compete and reduce energy at lift-off for GTO-GEO flights by introducing a concept like “Sea Launch”.



# **Build a Dashboard With Plotly Dash**



# SpaceX Falcon 9: Launch success count for all sites

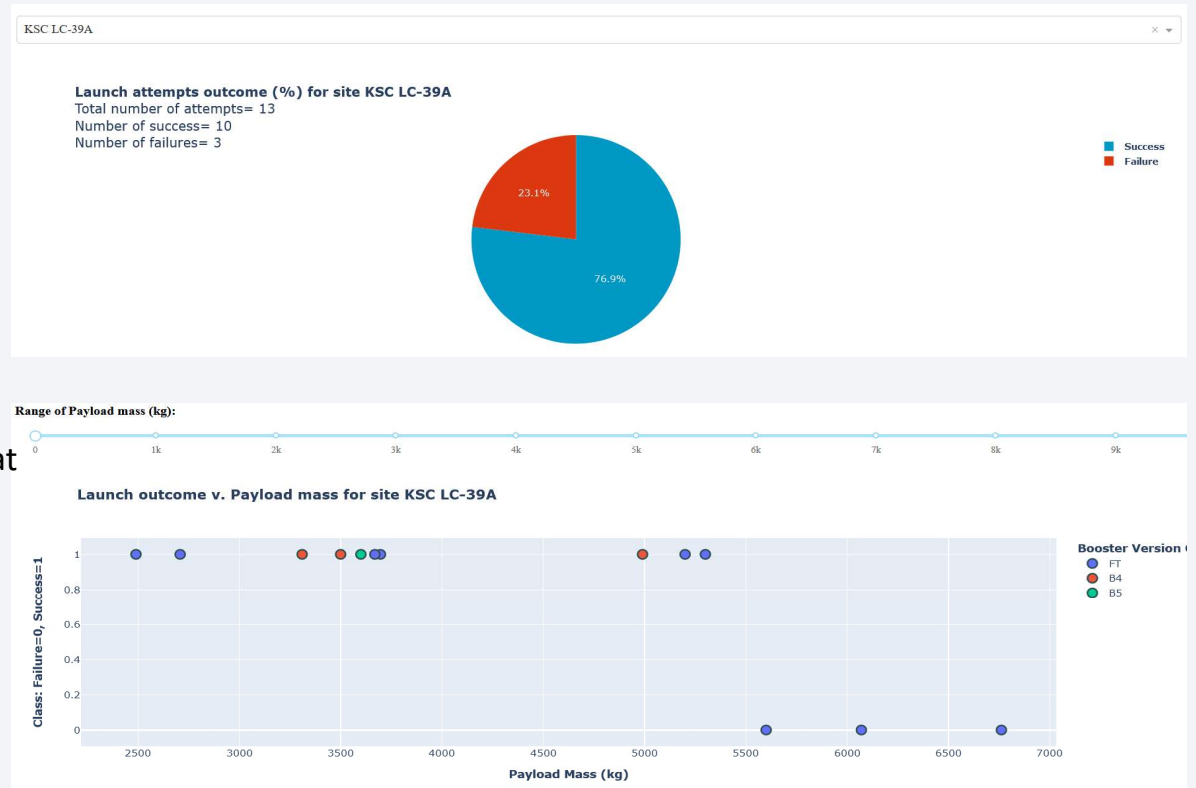


# SpaceX Falcon9 Launch site with highest launch success ratio

## KSC LC-39A

Kennedy Space Center in Florida.  
13 flights, 10 successful missions.

- Heavy payload are “high risk”
- Success does not seem to depend upon boosters versions with low mass payload <5500kg.
- B5 and FT are the most reused launchers. Data is not sufficient, but may indicates that they are as reliable as 1 time launchers.



# Launch outcome v. Payload mass (all sites)

- V1.0 and v1.1 are early launchers with low reliability.  
Landing legs, were pioneered on the Falcon 9 v1.1 version, but that version never landed intact.  
They were phased out in 2015.
- FT: “Full Thrust” is the next generation and has the highest success rate for payload mass under 6 tons. Including with “drone landing” (see details in next slide).
- Many FT flights are done with reused launchers. And show good reliability.
- Heavy payload are “high risk”.



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

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List the names of Falcon 9 boosters which have successfully landed on drone ship and had payload mass greater than 4000 but less than 6000 kg

Recent Full Thrust (FT) boosters exhibit the highest success rate on drone ship landing. Including with GTO flights.

It bodes well if SpaceX introduces a concept like “Sea Launch” for GTO launches at sea, close to the equator.

```
# sql query
q_boost_succ= """ select Booster_Version from spacex_v11 where Landing_Outcome = 'Success (drone ship)'
                  and PAYLOAD_MASS_KG_ > 4000
                  and PAYLOAD_MASS_KG_ < 6000 """
```

```
Booster_success_landing=pd.read_sql_query(q_boost_succ,conn)
```

```
print(Booster_success_landing)
```

```
Booster_Version
0      F9 FT B1022
1      F9 FT B1026
2  F9 FT B1021.2
3  F9 FT B1031.2
```

# **Predictive Analysis**

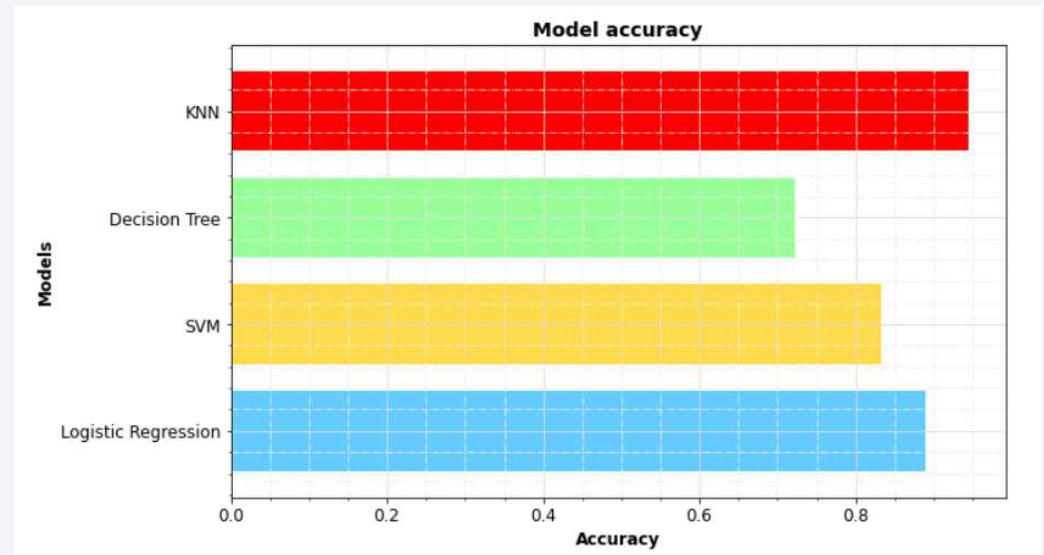
# Classification Accuracy

---

## Classification Accuracy with test set.

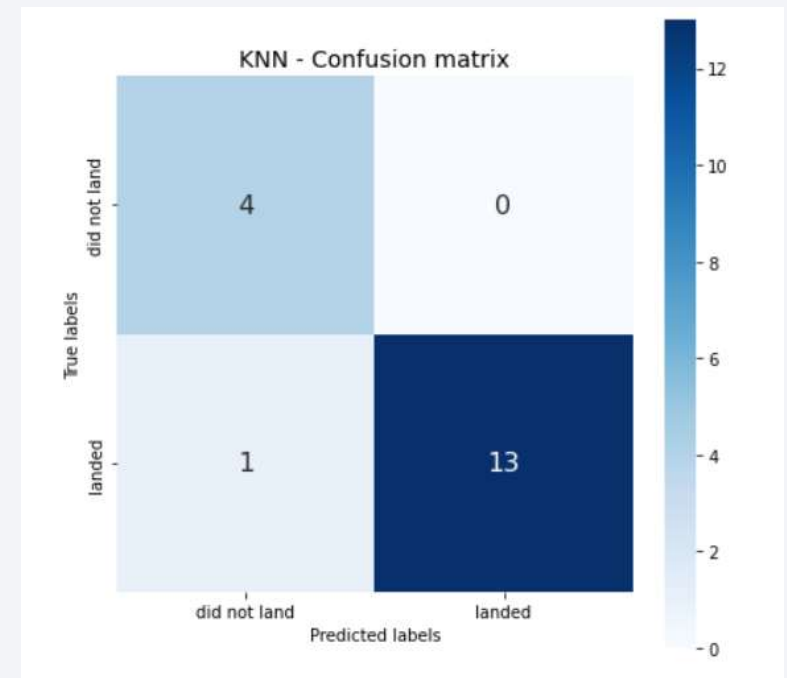
Results with “train test split” random\_state=3

- Optimization of SVM and LR hyper-parameters was refined for increasing accuracy with train set.
- 
- It did not necessarily improved accuracy with test set.
- Test set is too small.
- In our case, KNN exhibits the best accuracy: ~94%



# Confusion Matrix

- k-nearest neighbors algorithm (k-NN) is the best “predictor”
- The model perfectly predicts mission failure
- 1 false negative for successful booster landing (recovery)



## Conclusions

---

1. I have evaluated success through the Falcon 9 booster recovery process, analyzing data from Falcon 9 and Wikipedia to identify key factors: payload, orbit, booster type, and launch sites.
2. Using a supervised classification model, we predicted outcomes with 93% accuracy.
3. While Falcon 9 recovery is riskier than traditional launchers, its success rate—improving over time and at 65%—ensures cost-competitiveness against Ariane 5/6. GEO/GTO missions remain challenging due to energy demands and potential system damage.
4. SpaceX could enhance success by adopting equatorial launches, like “Sea Launch,” leveraging recent drone ship landings.
5. Starship’s potential may further dominate competitors.



# References and Jupyter Notebooks

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## Appendix: SQL queries

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Include any relevant assets like Python code snippets, SQL queries, charts, Notebook outputs, or data sets that you may have created during this project

# Total Payload Mass

Calculate the total payload carried by boosters from NASA (2 methods)

```
# For validation purposes... sum in df_NASA_CRS 'PAYLOAD_MASS_KG' column
df_NASA_CRS=pd.read_sql_query("Select * from spacex_v11 where Customer='NASA (CRS)'",conn)
print(df_NASA_CRS.head(2))
print('----')
print('Total payload mass, customer= NASA (CRS):', df_NASA_CRS['PAYLOAD_MASS_KG'].sum(), ' kg')
```

	id	Date	Time (UTC)	Booster_Version	Launch_Site	Payload \
0	4	2012-10-08	0 days 00:35:00	F9 v1.0 B0006	CCAFS LC-40	SpaceX CRS-1
1	5	2013-03-01	0 days 15:10:00	F9 v1.0 B0007	CCAFS LC-40	SpaceX CRS-2

	PAYLOAD_MASS_KG	Orbit	Customer	Mission_Outcome	Landing_Outcome
0	500	LEO (ISS)	NASA (CRS)	Success	No attempt
1	677	LEO (ISS)	NASA (CRS)	Success	No attempt

----  
Total payload mass, customer= NASA (CRS): 45596 kg

```
# Based on SQL only...
sql_nasa_crs_mass= """ Select sum(PAYLOAD_MASS_KG_) as 'Total payload mass (kg) NASA CRS'
                        from spacex_v11
                        where Customer='NASA (CRS)' """
payload_NASA_CRS=pd.read_sql_query(sql_nasa_crs_mass,conn)
print(payload_NASA_CRS)
```

	Total payload mass (kg) NASA CRS
0	45596.0

# Average Payload Mass by F9 v1.1

---

Calculate the average payload mass carried by booster version F9 v1.1

```
payload_F9v11=pd.read_sql_query("Select avg(PAYLOAD_MASS_KG_) as 'avg mass (kg)' from spacex_v11 where Booster_Version='F9 v1.1'",conn)
print(payload_F9v11)
```

	avg mass (kg)
0	2928.4

# First Successful Ground Landing Date

---

Find the dates of the first successful landing outcome on ground pad

```
min_Date_success_landing=pd.read_sql_query("select min(Date) from spacex_v11 where Landing_Outcome = 'Success (ground pad)'",conn)
print(min_Date_success_landing)
```

```
min(Date)
0 2015-12-22
```

# Boosters Carried Maximum Payload

---

List the names of the boosters which have carried the maximum payload mass

```
# sql query
qboost= """Select distinct Booster_Version, max(PAYLOAD_MASS_KG_) as max_payload_mass
from spacex_v11
group by Booster_Version
order by max_payload_mass desc"""
```

```
boost_max_load= pd.read_sql_query(qboost,conn)
boost_max_load.head(5)
```

	Booster_Version	max_payload_mass
0	F9 B5 B1049.4	15600
1	F9 B5 B1060.2	15600
2	F9 B5 B1048.4	15600
3	F9 B5 B1048.5	15600
4	F9 B5 B1056.4	15600

# 2015 Launch Records

---

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

```
# sql query
q_failed_landing= """ Select Date, Booster_Version, Launch_Site, Landing_Outcome
                        from spacex_v11
                        where Landing_Outcome = 'Failure (drone ship)'
                        and Date like '%2015%' """
```

```
fail_drone= pd.read_sql_query(q_failed_landing,conn)
fail_drone.head(5)
```

	Date	Booster_Version	Launch_Site	Landing_Outcome
0	2015-01-10	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
1	2015-04-14	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 query
q_count_landing= """ Select Landing_Outcome, count(*) as count_landings
                      from spacex_v11
                      where Date between '2010-06-04' and '2017-03-20'
                      group by Landing_Outcome
                      order by count_landings desc """
```

```
count_landing= pd.read_sql_query(q_count_landing,conn)
count_landing.head(10)
```

	Landing_Outcome	count_landings
0	No attempt	10
1	Failure (drone ship)	5
2	Success (drone ship)	5
3	Controlled (ocean)	3
4	Success (ground pad)	3
5	Uncontrolled (ocean)	2
6	Failure (parachute)	1
7	Precluded (drone ship)	1



Thank You