

Winning Space Race with Data Science

EastPersiaLtd 15th Jan 2023



Outline

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

Executive Summary

• Summary of methodologies: By using requests, sklearn, pandas, folium, matplotlib packages, I could perform all process of data collecting, wrangling, visualising, analysing, and predicting.

- Summary of all results:
- 1. Mission setting and booster performance effect on success. Especially: Mission orbit.
- 2. Space mission launch site is far from city, infrastructure in common.
- 3. SpaceX's success rate, include retrieving boosters, is going increased steadily.
- 4. Decision Tree was the best model for classification.

Introduction

 To practice real world data for proving abilities as a data scientist, I researched SpaceX's launch dataset

- Questions:
- 1. Do boosters' performance or mission setting effect on success?
- 2. What is Geographical conditions of Launch site?
- 3. SpaceX doing well since they started to participate space missions?
- 4. Which model is the best for classify cases?



Methodology

Executive Summary

- Data collection methodology:
 - Originally planned to use API, but due to sustain constant dataset, used pre-defined.
- Perform data wrangling
 - Data was cleant by using SQL, and DataFrame controlling.
- Perform exploratory data analysis (EDA) using visualization and SQL
- Perform interactive visual analytics using Folium and Plotly Dash
- Perform predictive analysis using classification models
 - With sklearn, used KNN, DecisionTree, SVM, Logistic Regression

Data Collection

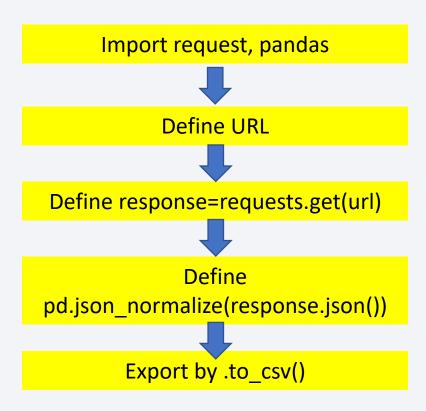
- Original Plan
- By using requests function to <u>Space X website</u>, get information in JSON form, and transform to pd.DataFrame.

- Actual
- Due to keep constant data file by IBM's rule, used pre-defined file and transformed it into pd.DataFrame.

Data Collection - SpaceX API

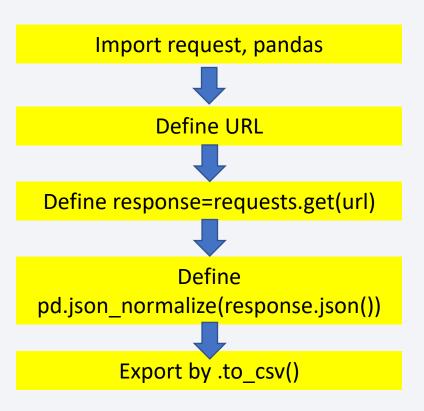
• SpaceX API was easy-to-use, but it included encoded strings.

Request to SpaceX API returned 200 code and information.



Data Collection - Scraping

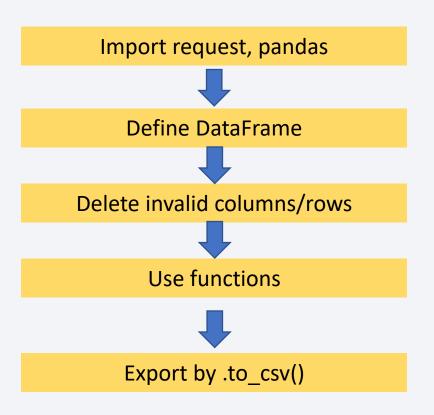
• Due to course policy of IBM, Dataset I used is offered by Coursera. Process was same to using SpaceX API.



Data Wrangling

• By assistant of course ipynb file, Data wrangling was successful.

• To do this process quick and faster, used 3 functions, DataFrame editing functions, some of lambda.



EDA with Data Visualization

 These charts are showing correlations between X and Y axis. In EDA stage, this process is required to seek hidden relations.

• Several type of chart used: Scatter, Line, Bar.



EDA with SQL

- Four Launch Sites exist
- Total 45.6t of payload has been launched by NASA (CRS)
- F9 v1.1 carried about 2.5t of payload in average
- First successful landing on ground pad was achieved in Dec 2015
- SpaceX achieved 100 Mission success, and 1 Mission failure
- F9 B5 is the strongest booster for payload.
- During 2010-06-04 and 2017-03-20, SpaceX mostly got 'No attempt' at Booster Landing stage.

Build an Interactive Map with Folium

• I created objects on KSC LC-39A, for markers, lines of nearest coastline, city, railway, and highway

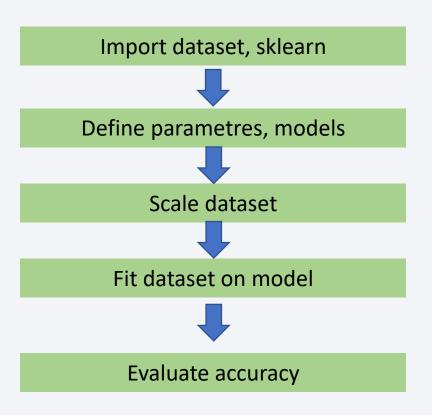
• These are good example to show geographical conditions of launch site.

Build a Dashboard with Plotly Dash

Predictive Analysis (Classification)

• I created KNN, SVM, Decision Tree, Logistic Regression with sklearn, also used StandardScaler

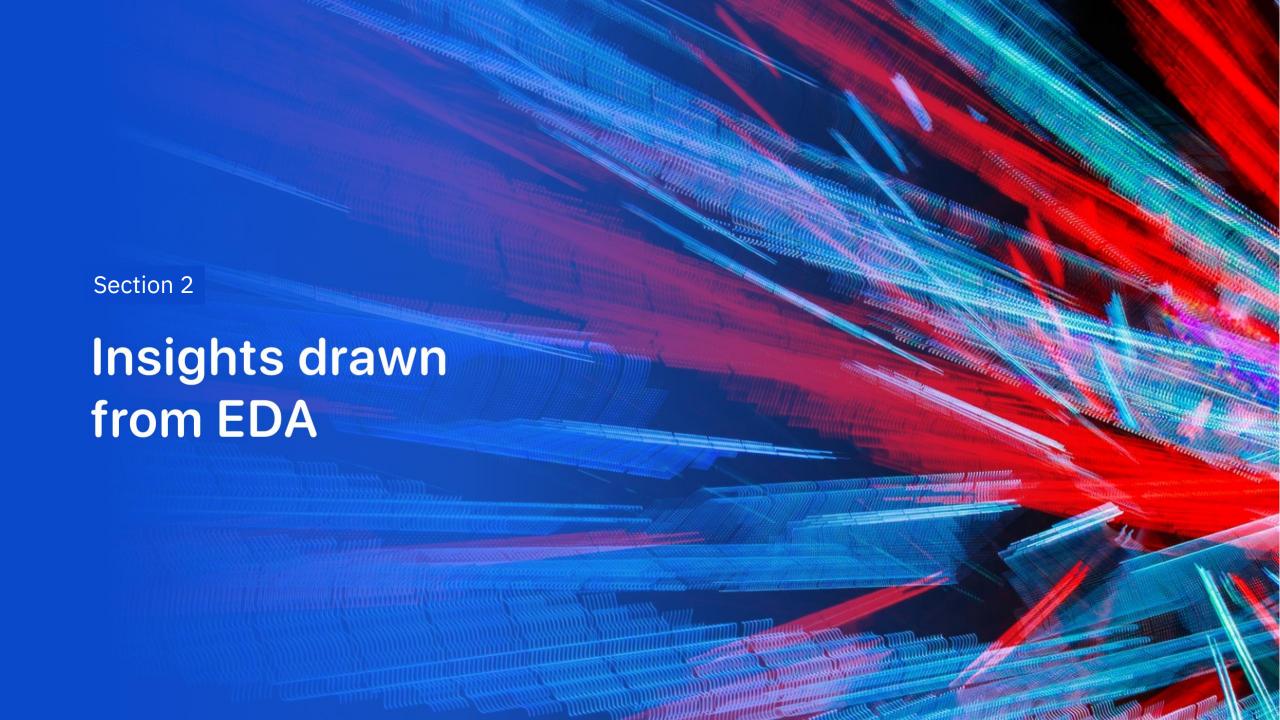
 You need present your model development process using key phrases and flowchart



Results

• During EDA analysis, I could find significant relations between 1. year and payload, success rate, 2. flight number and orbit altitude, 3. flight number and launch site.

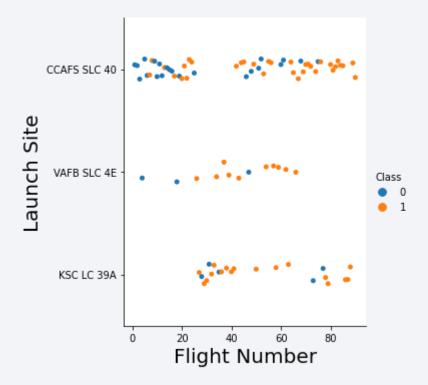
• During Predictive analysis, Decision Tree recorded the highest accuracy, which was 87.5%.



Flight Number vs. Launch Site

• The chart shows that there is significant relations between flight number and launch site

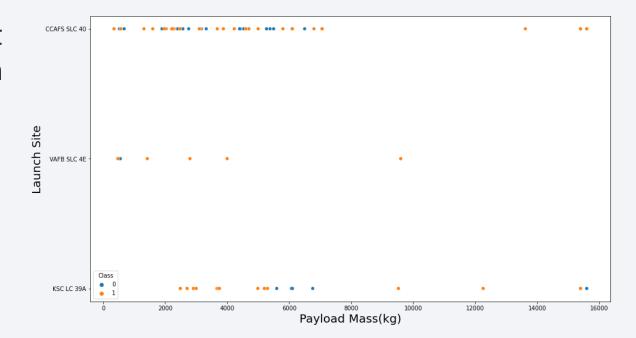
• VAFB SLC 4E is no longer used since KSC LC 39A is in operation.



Payload vs. Launch Site

• In this chart, I couldn't find significant relations between payload and launch site.

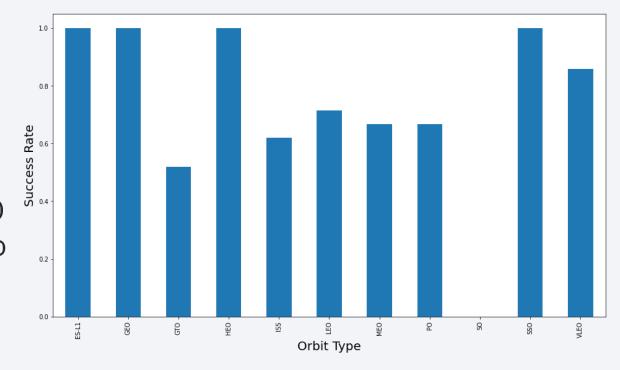
• This chart is not enough to reveal relations; Box-Whisker plot preferred.



Success Rate vs. Orbit Type

• This chart shows that significant relations between success rate and orbit type.

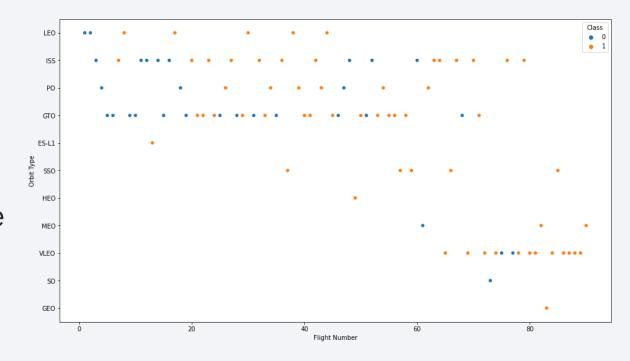
• Stable orbits, such as ES-L1, GEO, SSO recorded high success rate compare to unstable'.



Flight Number vs. Orbit Type

• This chart shows that significant relations between flight number and orbit type.

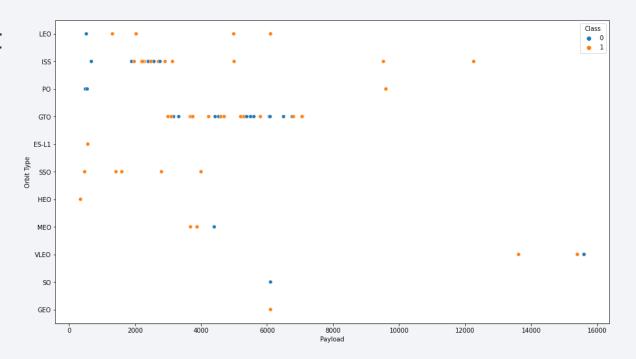
• In recent flight number, boosters have been launched for VLEO, which is very dangerous orbit. Means, boosters are stabilised in their operations.



Payload vs. Orbit Type

• In this chart, I couldn't find significant relations between payload and orbit type.

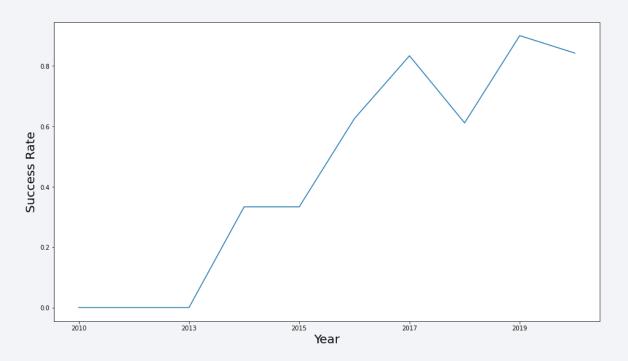
• This chart is not enough to reveal relations; Box-Whisker plot preferred.



Launch Success Yearly Trend

• This chart shows growing of success rate by every year.

• In addition, sns.regplot is much worth to estimate trend of success rate.



All Launch Site Names

- Launch Site
- 1. CCAFS LC-40
- 2. CCAFS SLC-40
- 3. KSC LC-39A
- 4. VAFB SLC-4E

select distinct launch_site from new;

• By using distinct clause, the query returns unique values only.

Launch Site Names Begin with 'CCA'

- Five records where launch sites begin with `CCA`
- 1. 2010-06-04
- 2. 2010-12-08
- 3. 2012-05-22
- 4. 2012-10-08
- 5. 2013-03-01
- select * from new where launch_site like '%CCA%' limit
 5;
- By using like, limit clause, the query returns 5 records.

Total Payload Mass

- The total payload carried by boosters from NASA (CRS)
- 45596kg

```
select sum(payload_mass__kg_) from new where
customer='NASA (CRS)' group by customer;
```

• By using sum, group by functions, the query returns sum value of NASA (CRS).

Average Payload Mass by F9 v1.1

- The avg payload mass carried by booster version F9 v1.1
- 2534kg

```
select avg(payload_mass__kg_) from new where
booster_version like 'F9 v1.1%';
```

By using avg, the query returns avg value.

First Successful Ground Landing Date

- The dates of the first successful landing outcome on ground pad
- 2015-12-22

```
select min(date) from new where
landing__outcome='Success (ground pad)';
```

• By using min clause on date, the query returns the earlist date value.

Successful Drone Ship Landing with Payload between 4000 and 6000

- The names of boosters which have successfully landed on drone ship and had 4000<payload mass<6000
- 1. F9 FT B1022
- 2. F9 FT B1026
- 3. F9 FT B1021.2
- 4. F9 FT B1031.2

```
select booster_version from new where
landing__outcome='Success (drone ship)' and
payload_mass__kg_>4000 and payload_mass__kg_<6000;</pre>
```

• By using AND in where clause, the query returns right values.

Total Number of Successful and Failure Mission Outcomes

- The total number of successful and failure mission outcomes
- Success: 100(99+1)
- Failure: 1

```
select count(mission_outcome) as success from new where
mission_outcome like '%Success%' group by
mission_outcome;
```

```
select count(mission_outcome) as failed from new where
mission_outcome like '%Failure%' group by
mission_outcome;
```

By using two queries, they returns right values.

Boosters Carried Maximum Payload

- The names of the booster 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 (total 12)

```
select distinct booster_version from new
where payload_mass__kg_=(select max(payload_mass__kg_)
from new);
```

 By using subquery, the query return right recordes which the maximum payload

2015 Launch Records

- The failed landing_outcomes in drone ship, their booster versions, and launch site names for in year 2015
- 1. 2015-01-10 F9 v1.1 B1012 CCAFS LC-40 Failure (drone ship)
- 2. 2015-04-14 F9 v1.1 B1015 CCAFS LC-40 Failure (drone ship)

```
select date, booster_version, launch_site,
landing__outcome from new where
landing__outcome='Failure (drone ship)' and
year(date)='2015'
```

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

• 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

```
select count(landing__outcome) as count,
landing__outcome

from new where date(date) between
'2010-06-04' and '2017-03-20'
group by landing__outcome
order by count(landing__outcome) desc;
```

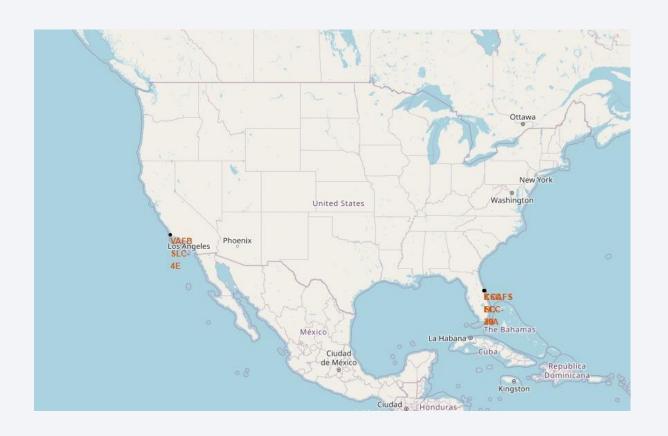
COUNT	landingoutcome
10	No attempt
5	Failure (drone ship)
5	Success (drone ship)
3	Controlled (ocean)
3	Success (ground pad)
2	Failure (parachute)
2	Uncontrolled (ocean)
1	Precluded (drone ship)



Folium Overview

• In this chart, locations of launch sites are shown as small black markers.

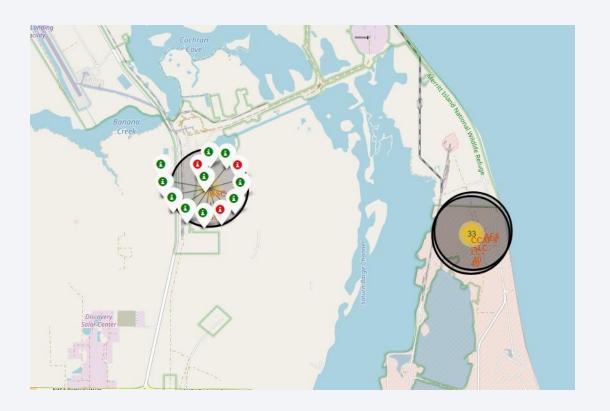
• Exclude overlapping issue, markers provide intuitive geographical information.



Folium Coloured Label

• In this chart, success or failure displayed as colour. Green means successed to retrieve boosters, Red means failed to do it.

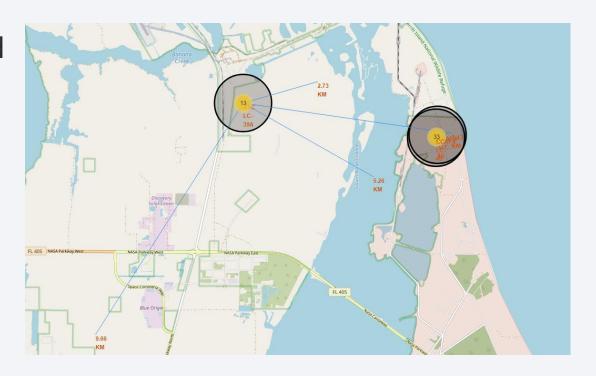
• Somehow, show success rate might better way than coloured label.

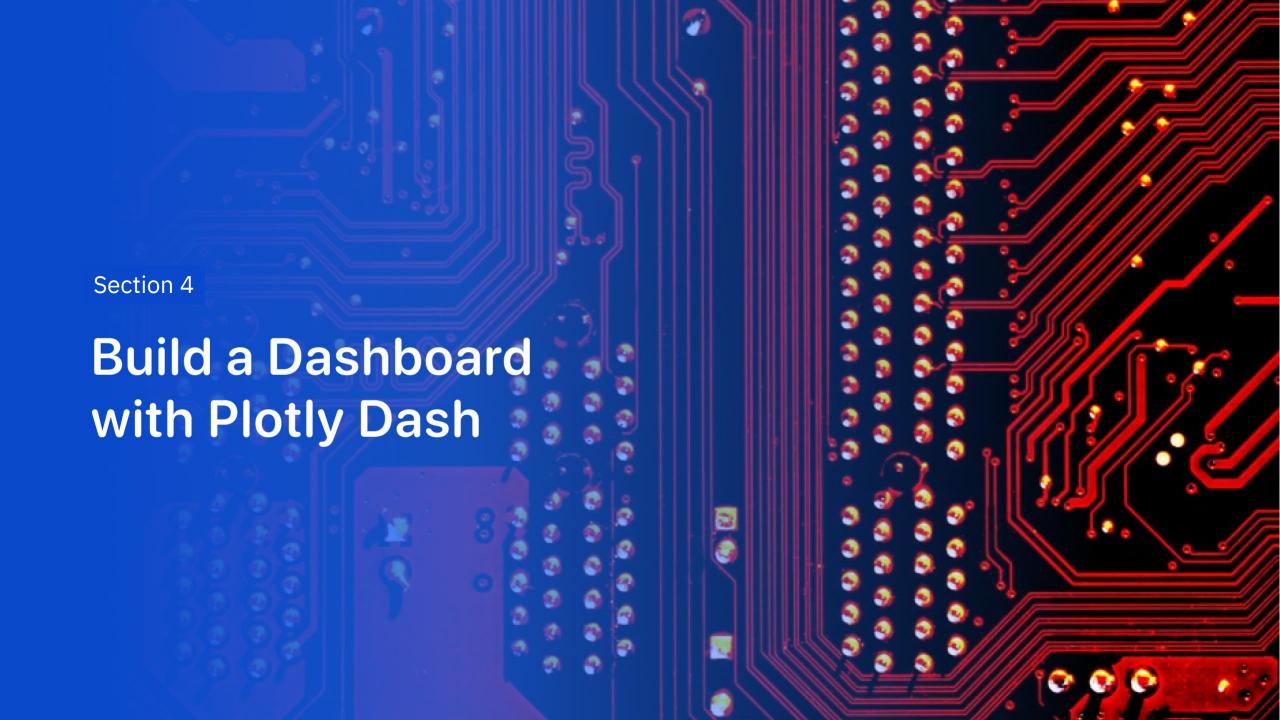


Folium Marker

• In this chart, nearest city, railway, and highway are displayed by lines and their distance from the KSC LC-39A site.

• To my point of view, Launch site is generally located far from city or civil infrastructure.



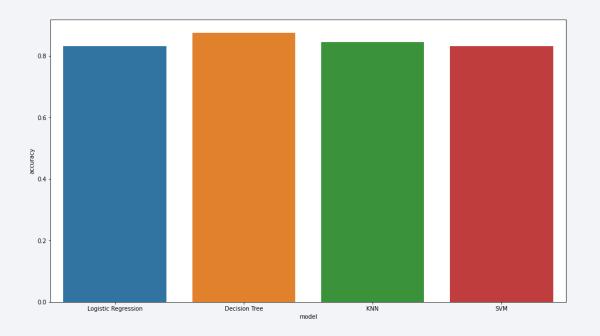




Classification Accuracy

• In this chart, there are four classification model used to classify landing success/failure.

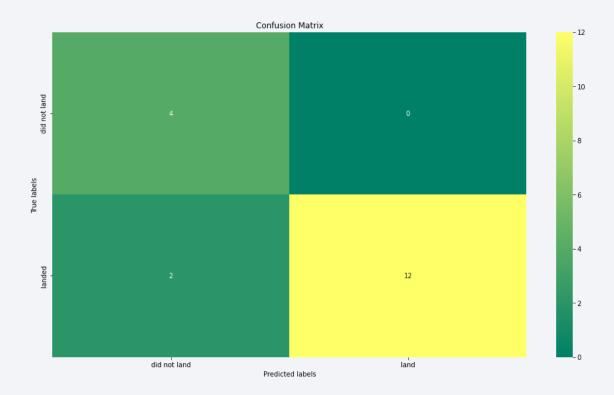
• The best performance was performed by Decision tree, with accuracy of 0.875.



Confusion Matrix

• This is a confusion matrix of the Decision tree model.

• Yellow means high observation, green means low observation.



Conclusions

- 1. Mission setting and booster performance effect on success. Especially: Mission orbit.
- 2. Space mission launch site is far from city, infrastructure in common.
- 3. SpaceX's success rate, include retrieving boosters, is going increased steadily.
- 4. Decision Tree was the best model for classification.

Appendix

• Tips for connect to Db2:

```
!pip install --upgrade ibm db
!pip install --upgrade ibm_db_sa
!pip install --upgrade SQLAlchemy
import ibm db
import ibm db sa
import ibm db dbi
import sqlalchemy
from sqlalchemy import *
import pandas as pd
%load ext sql
%reload_ext sql
%sql
ibm_db_sa://"ABC12345":passwordddddddddddddddddefghijklmnopqrstuvwxyzabcdefghijklmnopqrstuvwxyzabcdef.databases.app
domain.cloud:12345/bludb;security=SSL;
                                                                                                                              46
```

Appendix

All notebooks and dataset files:

Visit <u>EastPersiaLtd/IBM-DS</u> for further details, SQL queries, dataset, etc.

Or

git clone https://github.com/EastPersiaLtd/IBM-DS.git on git bash

Or

! git clone https://github.com/EastPersiaLtd/IBM-DS.git on Google Colaboratory

