Predict the success of SpaceX's Falcon 9 first-stage launches

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Table of contents

01

Introduction

03

Findings

02

Methodology

04

Conclusions



Introduction and Objectives

SpaceX's falcon 9

SpaceX is popular for her innovations and leading positions in the space industry. Many of their accomplishments include sending humans and spacecrafts to the Space and developing the world's most powerful rocket.

During 2013 and 2016, SpaceX conducted a series of controlled-descent flight tests for its Falcon 9 rocket.

SpaceX's Falcon 9 rocket was well-known for its ability to recover the first stage, which could save a lot of money for the company.



A falcon 9 rocket on a ground pad at Cape Canaveral AF Station

(source: wikipedia)

Objectives



Our aim

The aim in the project is to determine the price of each rocket launch by gathering information about SpaceX and creating dashboards for the team.



The goal

Successfully train a machine learning model to predict if SpaceX will reuse the first stage - which will determine the cost of each launch



Methodology

Data source

	Description	Link
SpaceX REST API	Open source REST API for launch, rocket, core, capsule, starlink, launchpad and landing pad data	http://github.com/r-spacex/Spac eX-API
Falcon 9 Launch record	Wikipedia launch record data, include launch outcomes, sites, customers, version and date	https://en.wikipedia.org/w/index. php?title=List_of_Falcon_9_and_Fa lcon_Heavy_launches&oldid=10276 86922

Tools



Python

Most of the work will be conducted by using Python.

Several python libraries will be used, such as pandas, numpy, matplotlib, dash, scikit-learn, ...



SQL

Using SQL queries to quickly find important exploratory insights

SQL syntax will be used directly in Python environment using *sqlalchemy*



Visualization libraries

To support the project's plotting, multiple Python visualization libraries will be used. Such as:

matplotlib, seaborn, folium, plotly

Data analysis process

Data collection API

(SpaceX REST API)

Requests

- Request to the SpaceX API
- Clean the requested data

Data wrangling

pandas, numpy

- Perform EDA to find patterns
- Determine model training labels

Data visualization

folium

- EDA by plotting geographical map
- Launch site locations analysis

ML Prediction

scikit-learn

- Data preparation labeling,: standardize, train/test split
- SVM, Decision Tree, Logistic Regression and kNN



Web Scraping

(Launch record)

BeautifulSoup

- Extract html table from website
- Parse the table and convert into pandas df

Descriptive Analysis

SQl, matplotlib, seaborn

- Use a MySQL local database
- Use SQL queries, plots to find correlations, insights

Dashboard

Plotly, Dash

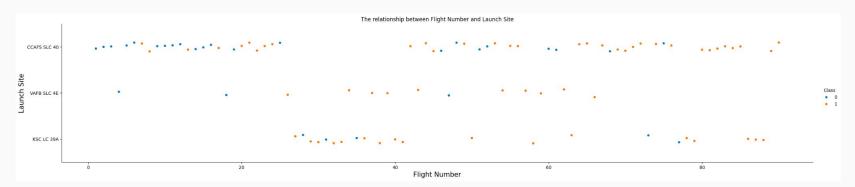
Create a live, interactive dashboard using plotly and dash

Report



Findings

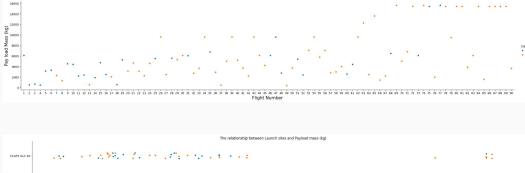
Exploratory research: Visualization

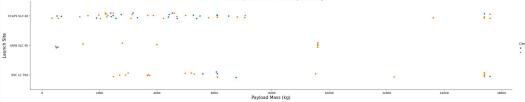


Later flights have higher success rate

- The later the flight, the higher the chance of success
- From flight 78th, the success rate is 100%

Exploratory research: Visualization



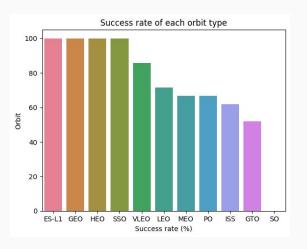


Higher payload mass results in better result

By observing the success rate and the relationship between Payload Mass and Number of Flights + Launch Site

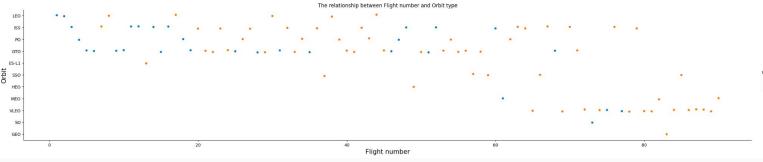
- Success rate for launches from >8000kg payload is very high (~87%)
- Two sites KSC LC-39A and VAFB SLC 4E has success rate of 77% while the other has 60%

Exploratory research: Visualization



Orbit types

- For orbits that have at least 5 landings, SSO is the only orbit that has a success rate of 100%
- VLEO and LEO has significantly high success rate
 ~70-80%, and their success rate seems to correlate to the number of flights
- GTO orbit has a low success rate of 50%



Exploratory research: SQL

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

Count	Landing_Outcome	
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)	

Various landing outcomes, but there's always a successful mission

- Many landing outcomes: There are 8 types of landing outcomes
- **No attempt** happened the most (10 times).
- However, landing outcomes related to **drone ship** happened the most (11 times)
- High mission success rate
 (100/101), include the mission with payload status unclear

Exploratory research: Mapping





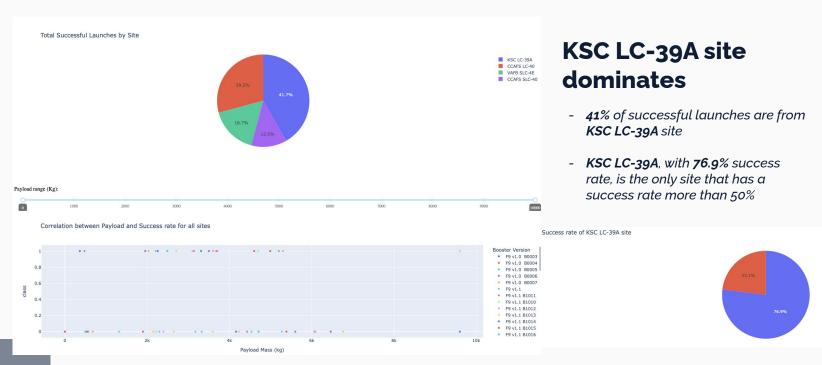


Launch sites has notable geographical characteristics

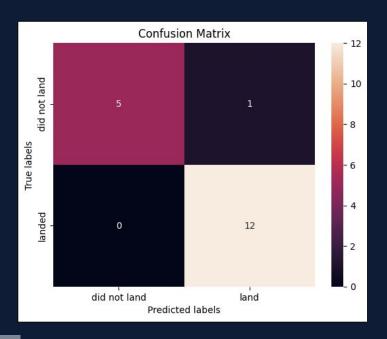
By observing:

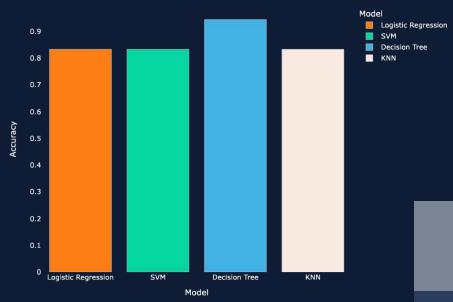
- Launch sites are close to railways, highways and coastline, which is important and convenient for transportation and logistics
- Launch sites are far away from cities, which is necessary for safety and security reasons

Dashboard: Mapping



ML Predicting: Decision Tree





ML Predicting: Decision Tree

Classification Tree has the highest predicting accuracy

- Out of 18 samples, the model predicted 17 correct with a success rate of 94%
- In the wrong prediction:
 - The model predicted the positive class incorrectly, which was a False Positive (FP case)
 - The model predicted a successfully landed case, but the actual value was a 'did not land' case
- The other 3 methods: SVM, logistic regression and k nearest neighbor has a similar success rate of 83.3%



Conclusions

Conclusions

Prediction method: Decision Tree

The test shows an outstanding result of 94% accuracy for decision tree classification. Which should be used for future prediction.

Geographical aspects

Geography shows an essential impact to the launch sites. Most sites were placed close to transportations for logistics and were far from urban areas.

Promising future

Recent testing results show positive outcomes, which is an improvement compared to the past. Most missions were completed successfully.