

FALCON-9 PROJECT REPORT

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Aug-2025



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Executive Summary

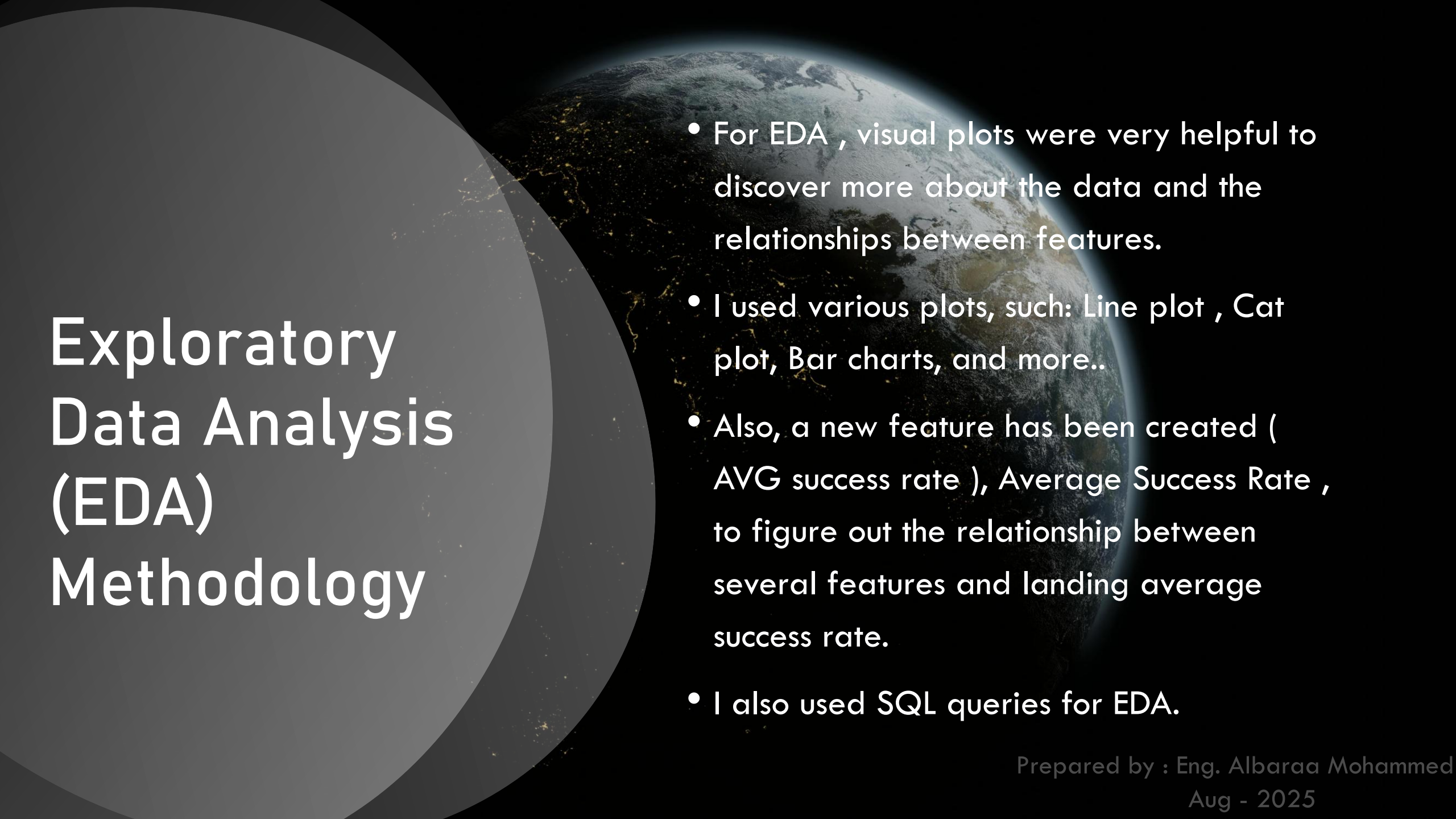
- In this project, I've built a ML model to determine whether the SpaceX Falcon-9 rocket's first stage will land successfully or not.
- The ML model used for this project is Classification. And to determine the accuracy of the model I used Confusion Matrix method.
- Several algorithms for Classification were used also, such as: Logistic Regression, KNN, SVM, Decision Trees. And compared between them to determine the best option and that option was (either Logistic Regression and SVM).

Introduction

- SpaceX has launched Falcon-9 rocket on their website, the rocket comes with a great price (62million\$), comparing to other manufacturer which their cost nearby(1 65million\$).
- The reason for this is that SpaceX can use the first stage of the rocket again.

Data Collection Methodology

- The data used for this project from SpaceX official website.
- The methodology to get this data is API .
- To prepare the data well, I used some functions to get important features in consist way, like:(Booster Version, Payload Mass, Cores, and Launch Site).
- Also, since the ML is classification, I had to convert all object data into a numerical data to prepare it for analysis, I create a new feature columns called Class that represent weather the launch landed or not by (0 & 1).



Exploratory Data Analysis (EDA) Methodology

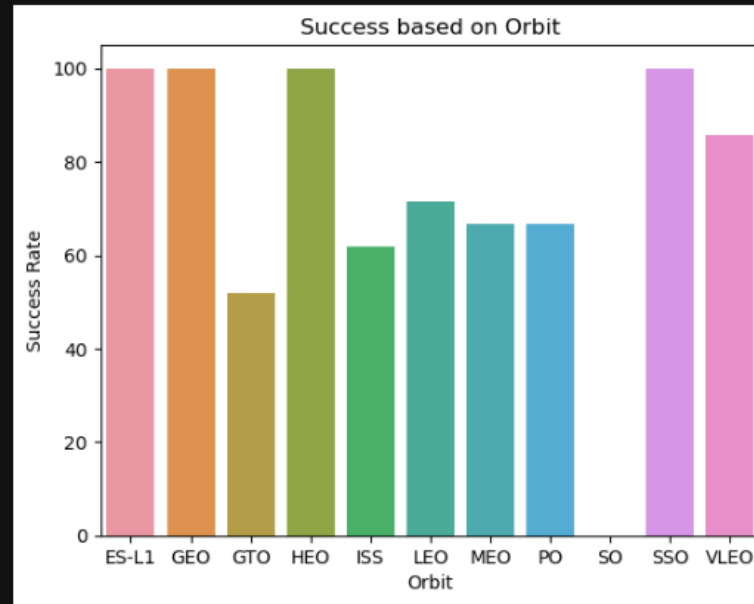
- For EDA , visual plots were very helpful to discover more about the data and the relationships between features.
- I used various plots, such: Line plot , Cat plot, Bar charts, and more..
- Also, a new feature has been created (AVG success rate), Average Success Rate , to figure out the relationship between several features and landing average success rate.
- I also used SQL queries for EDA.

Predictive Model Methodology

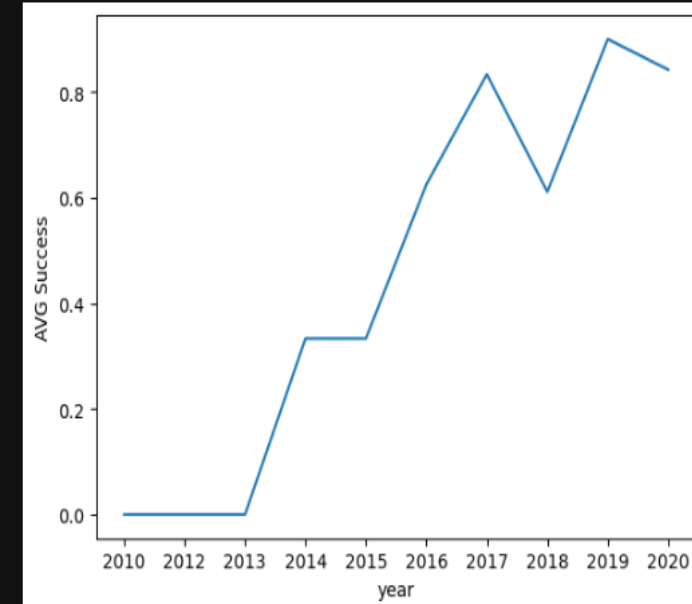
- The predictive model was Classification. I've used several algorithm such: KNN, SVM, Decision Trees, and Logistic Regression.
- To compare between them, I've created a function that draw a confusion matrix to ease the process of analysis.
- Also, Cross-Validation was used using Grid Search to better overcome overfitting and underfitting.

EDA with Visualization Results

```
: sns.barplot(data=success,x='Orbit',y='Success Rate')
plt.title('Success based on Orbit')
plt.xlabel('Orbit')
plt.ylabel('Success Rate')
plt.show()
```



```
9]: sns.lineplot(data=suc,x='Year',y='AVG Success')
plt.xlabel('year')
plt.ylabel('AVG Success')
plt.show()
```

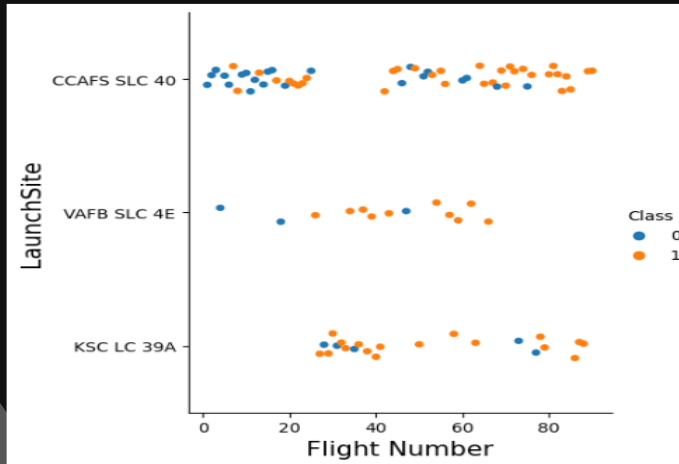


As we can see , Orbit has clear affect on the Success rate of the landing process And GTO is the least one.

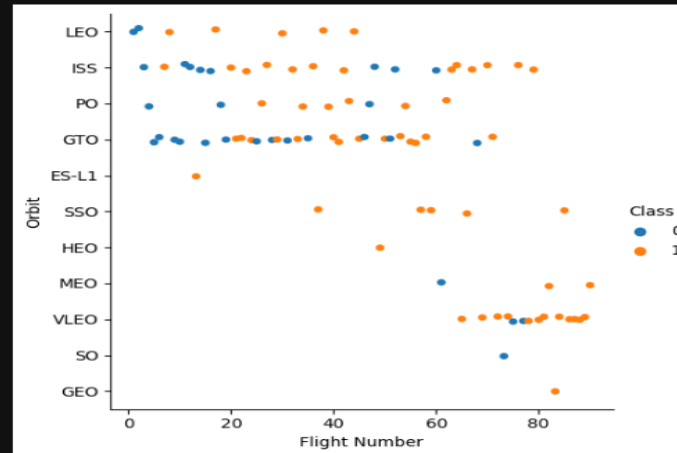
From 2013 AVG success rate increased significantly.

EDA with Visualization Results

```
[99]: sns.catplot(y='LaunchSite',x='FlightNumber',data=df3,hue='Class')
plt.xlabel('Flight Number',fontsize=15)
plt.ylabel('LaunchSite',fontsize=15)
plt.show()
```



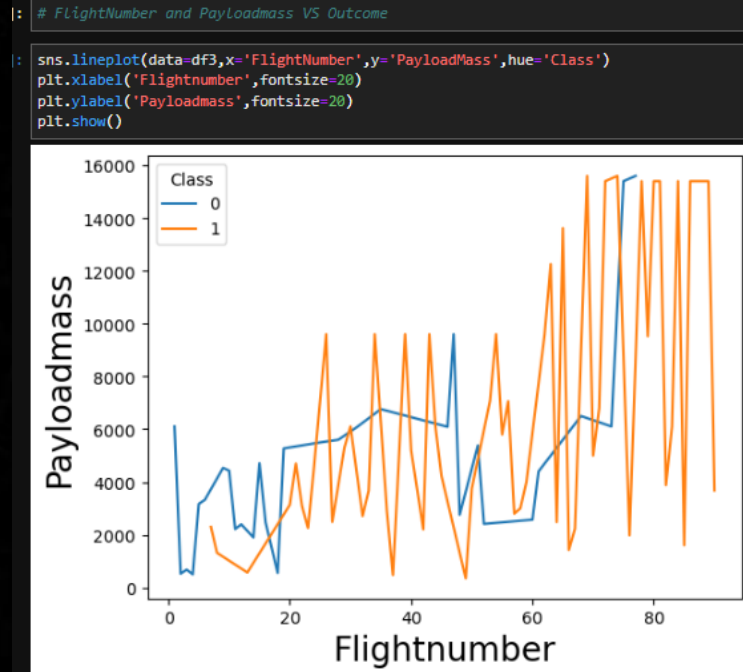
```
[103]: sns.catplot(y='Orbit',x='FlightNumber',data=df3,hue='Class')
plt.xlabel('Flight Number')
plt.ylabel('Orbit')
plt.show()
```



We can see here that GTO orbit has the most flights and we can say the success and failure were almost equal. Flights to SSO orbit always succeed landing.

CCAFS SLC 40 has the most flights and after 50 flights the success rate increased.

EDA with Visualization Results



The effect of payload to success rate was clear if the payload above 12 Ton and flights above 75.



SQL Results for EDA

```
[78]: %sql SELECT * from SPACEXTBL WHERE (Launch_Site) LIKE 'CCA*' LIMIT 5;
```

```
* sqlite:///my_data1.db  
Done.
```

```
[78]:
```

	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	7:44:00	F9 v1.0 B0005	CCAFS LC-40	Dragon demo flight C2	525	LEO (ISS)	NASA (COTS)	Success	No attempt
	2012-10-08	0:35:00	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

```
[92]: %sql SELECT DISTINCT Launch_Site  
FROM SPACEXTBL;
```

```
* sqlite:///my_data1.db  
Done.
```

```
[92]:
```

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

```
[89]: %sql  
SELECT  
    substr(Date, 6, 2) AS Month,  
    Booster_Version,  
    Launch_Site,  
    Landing_Outcome  
FROM SPACEXTBL  
WHERE Landing_Outcome = 'Failure (drone ship)'  
    AND substr(Date, 0, 5) = '2015';
```

```
* sqlite:///my_data1.db  
Done.
```

```
[89]:
```

Month	Booster_Version	Launch_Site	Landing_Outcome
01	F9 v1.1 B1012	CCAFS LC-40	Failure (drone ship)
04	F9 v1.1 B1015	CCAFS LC-40	Failure (drone ship)

```
[84]: %sql SELECT DISTINCT(Booster_Version) FROM SPACEXTBL  
WHERE Landing_Outcome='Success (drone ship)'  
AND PAYLOAD_MASS_KG_ BETWEEN 4000 AND 6000;
```

```
* sqlite:///my_data1.db  
Done.
```

```
[84]:
```

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

We notice here that CCAFS LC-40 have the biggest customer such, SpaceX and NASA.

And, there are several launch sites as shown above.

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SQL Results for EDA

```
[82]: %sql select sum(PAYLOAD_MASS_KG_) from spacextbl where Customer='SpaceX';
* sqlite:///my_data1.db
Done.
[82]: sum(PAYLOAD_MASS_KG_)
185220
```

```
[83]: %sql SELECT MIN(Date) AS First_Success_landing_pad
FROM SPACEXTBL
WHERE Landing_Outcome ='Success (ground pad)';
* sqlite:///my_data1.db
Done.
[83]: First_Success_landing_pad
2015-12-22
```

```
[80]: %sql select sum(PAYLOAD_MASS_KG_) from spacextbl where Customer='NASA (CRS)';
* sqlite:///my_data1.db
Done.
[80]: sum(PAYLOAD_MASS_KG_)
45596
```

We notice that SpaceX have much higher payload comparing to NASA, about 121%.

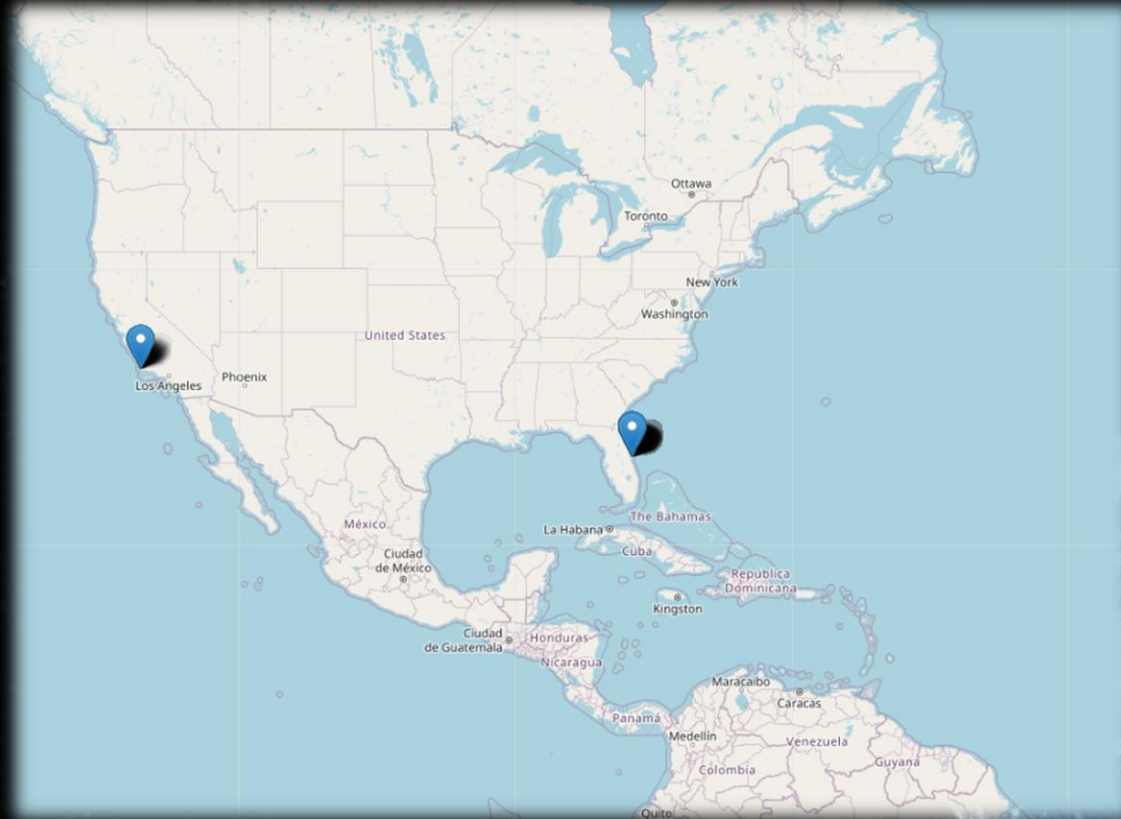
Also, some landing outcomes for several locations

First success ground pad location was in 2015

```
F9 FT B1031.2
[85]: %sql SELECT Landing_Outcome,COUNT(*)
FROM SPACEXTBL
GROUP BY Landing_Outcome;
* sqlite:///my_data1.db
Done.
[85]:
```

Landing_Outcome	COUNT(*)
Controlled (ocean)	5
Failure	3
Failure (drone ship)	5
Failure (parachute)	2
No attempt	21
No attempt	1
Precluded (drone ship)	1
Success	38
Success (drone ship)	14
Success (ground pad)	9
Uncontrolled (ocean)	2

Folium Results



We notice from this map, that launch sites are explicitly located at eastern and western coasts of US.

Predictive Model Results

Logistic Regression

The model performs very well and give overall accuracy 94%, also since the problem is False Positive the confusion matrix did well as well.

```
[195]: logreg_cv.fit(Xtrain, Ytrain)

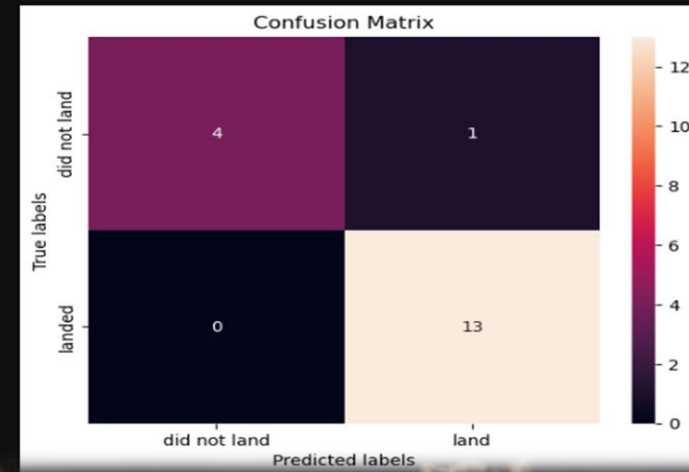
[195]: > GridSearchCV
> estimator: LogisticRegression
> LogisticRegression

[196]: print('the parameters of this model ',logreg_cv.best_params_)
print('the accuracy of these parmaeters ',logreg_cv.best_score_)
the parameters of this model {'C': 0.1, 'penalty': 'l2', 'solver': 'lbfgs'}
the accuracy of these parmaeters 0.8285714285714286

[197]: print('The Accuracy of test data for this model (logistic regression) is ',logreg_cv.score(Xtest,Ytest))
The Accuracy of test data for this model (logistic regression) is 0.9444444444444444

[198]: yhat=logreg_cv.predict(Xtest)

[199]: plot_confusion_matrix(Ytest,yhat)
```



Predictive Model Results

The model performs very well and give overall accuracy 94%, also since the problem is False Positive the confusion matrix did well as well.

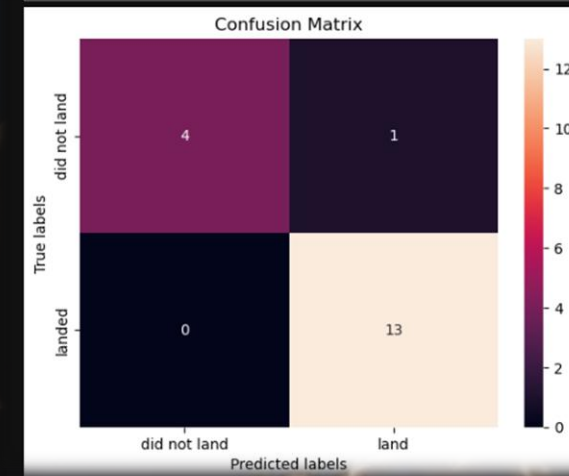
```
217: > GridSearchCV
      > estimator: SVC
      > SVC

218: print('the best hyperparameters ',svm_cv.best_params_)
      print('the accuracy of hyperparameters ',svm_cv.best_score_)
      the best hyperparameters {'C': 0.03162277660168379, 'gamma': 0.001, 'kernel': 'linear'}
      the accuracy of hyperparameters 0.8285714285714286

219: print('the accuracy of the test data for this model (SVM) is ',svm_cv.score(Xtest,Ytest))
      the accuracy of the test data for this model (SVM) is 0.9444444444444444

220: yhat2=svm_cv.predict(Xtest)

221: plot_confusion_matrix(Ytest,yhat)
```



Predictive Model Results

The model performs good and give overall accuracy 83%, also since the problem is False Positive the confusion matrix did well as well.

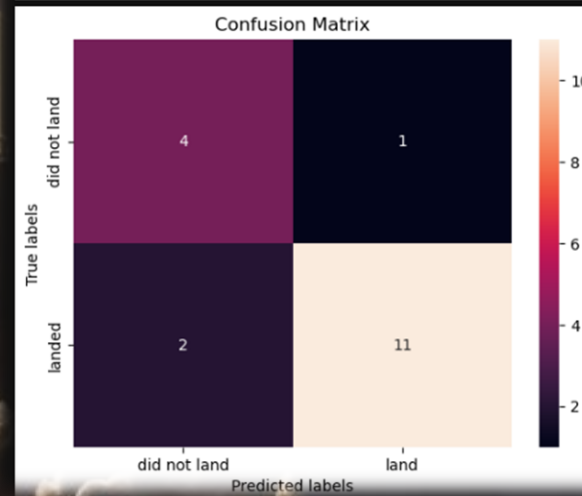
```
GridSearchCV
> estimator: DecisionTreeClassifier
  DecisionTreeClassifier

3]: print('the best parameters here are ',tree_cv.best_params_)
   print('the accuracy here ',tree_cv.best_score_)
the best parameters here are {'criterion': 'entropy', 'max_depth': 18, 'max_features': 'sqrt', 'min_samples_leaf': 2, 'min_samples_split': 10, 'splitter': 'best'}
the accuracy here 0.8857142857142858

4]: print('the accuracy of test data for this model (Tree) is ',tree_cv.score(Xtest,Ytest))
the accuracy of test data for this model (Tree) is 0.8333333333333334

5]: yhat3=tree_cv.predict(Xtest)

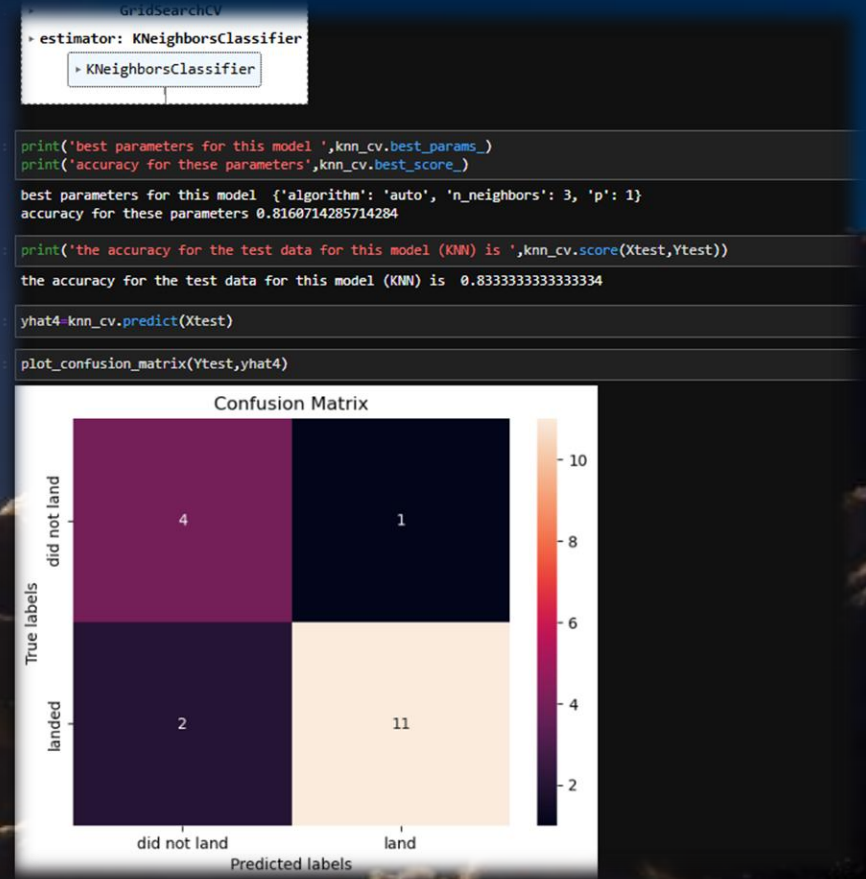
6]: plot_confusion_matrix(Ytest,yhat3)
```



Predictive Model Results

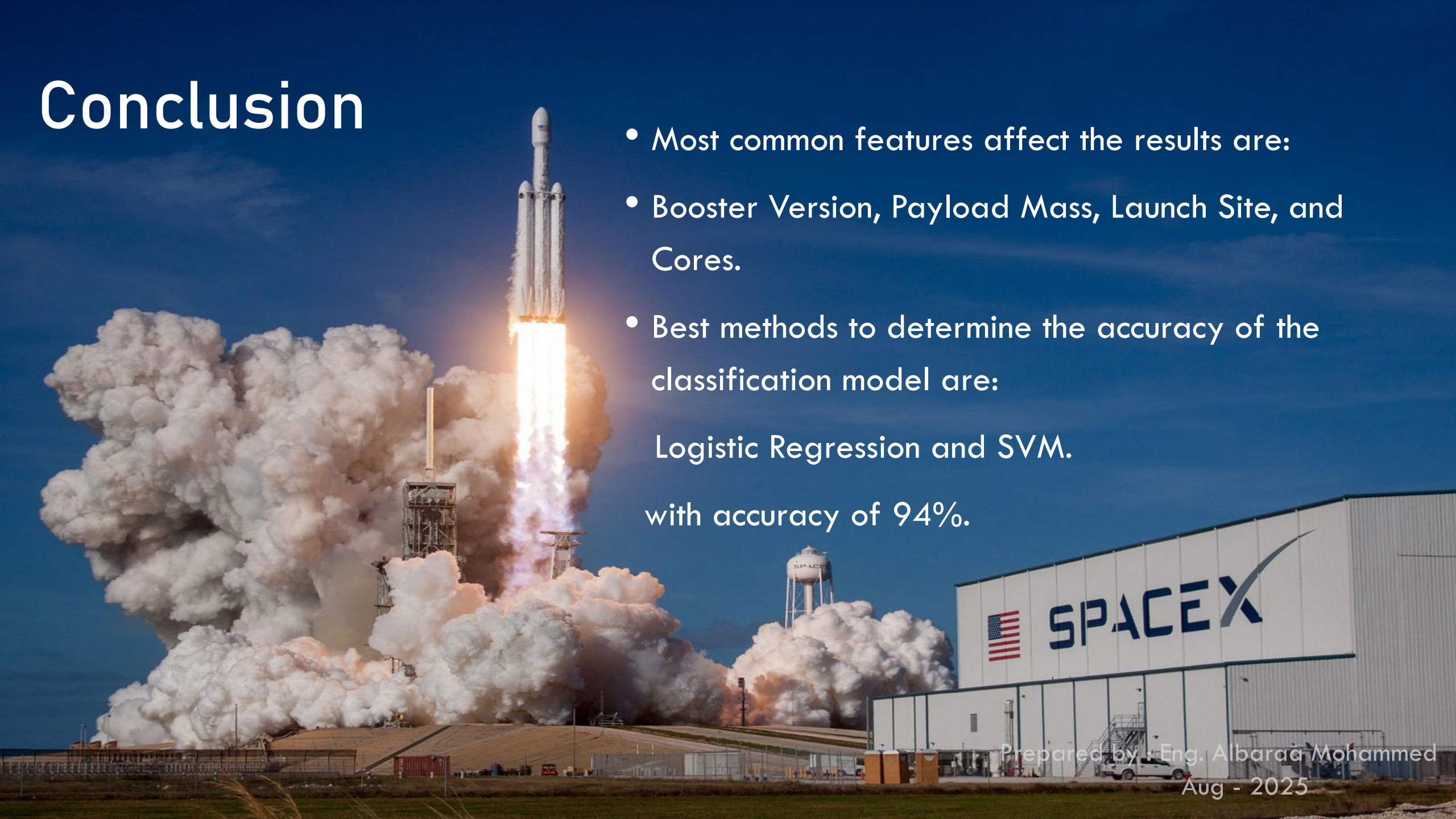
KNN

The model performs good and give overall accuracy 83%, also since the problem is False Positive the confusion matrix did well as well.



Conclusion

- Most common features affect the results are:
- Booster Version, Payload Mass, Launch Site, and Cores.
- Best methods to determine the accuracy of the classification model are:
Logistic Regression and SVM.
with accuracy of 94%.



Appendix

- <https://github.com/Albaraaakarman/Falcon-9-First-Stage-Classfier->





Thank You !

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