## Predicting Falcon 9 First Stage Landing Success: A Data Science Approach to Cost Optimization in SpaceX Launches

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### OUTLINE



- Executive Summary
- Introduction
- Methodology
- Results
  - Visualization Charts
  - Dashboard
- Discussion
  - Findings & Implications
- Conclusion
- Appendix

### **EXECUTIVE SUMMARY**



• **Objective**: Predict if Falcon 9's first stage will land using machine learning.

#### • Data:

- Cleaned and standardized.
- Split into training (80%) and test (20%) data.
- Models Tested: Logistic Regression, SVM, Decision Tree, KNN.
- Best Model:
  - Logistic Regression: 83.33% accuracy.
  - **Best Hyperparameters**: C=0.01, penalty=12, solver=1bfgs.
- Conclusion:
  - Logistic Regression performed best, but false positives remain.

### INTRODUCTION



- **Objective**: Predict if Falcon 9's first stage will land successfully to assess launch costs.
- Importance: Accurate predictions help SpaceX and competitors optimize pricing and improve efficiency.
- Approach: Use machine learning models (Logistic Regression, SVM, Decision Trees, KNN) to classify landing success.
- **Data**: Historical rocket launch data including mission details and outcomes.

## **METHODOLOGY**



- Data Preprocessing:
- Standardized the dataset to scale the features.
- Split the data into training (80%) and testing (20%) sets.
- Model Selection:
- Used Logistic Regression, Support Vector Machines (SVM), Decision Trees, and K-Nearest Neighbors (KNN) for classification.
- Hyperparameter Tuning:
- Applied GridSearchCV to identify the best hyperparameters for each model.
- Model Evaluation:
- Used accuracy and confusion matrix to evaluate performance on test data.



## **RESULTS**

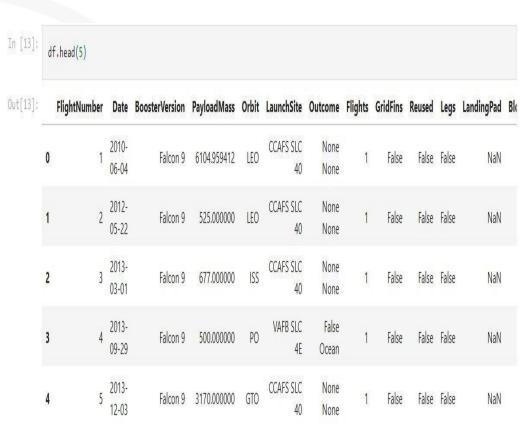
Model	Best Hyperparameters	Test Accuracy	<b>Key Observation</b>			
Logistic Regression	C=0.01, penalty='l2', solver='lbfgs'	83.33%	High accuracy with some false positives.			
Support Vector Machines	C=1.0, gamma=0.0316, kernel='sigmoid	84.82%	Performed well but with slightly higher complexity.			
Decision Tree Classifier	Multiple parameter optimizations	93.33%	High accuracy, but prone to overfitting.			
K-Nearest Neighbors	n_neighbors=5, algorithm='auto', p=2	76.67%	Lower accuracy compared to other models.			

### **Conclusion:**

• Decision Tree Classifier performed the best with 93.33% accuracy.

## Data Collection & Wrangling : Key Findings

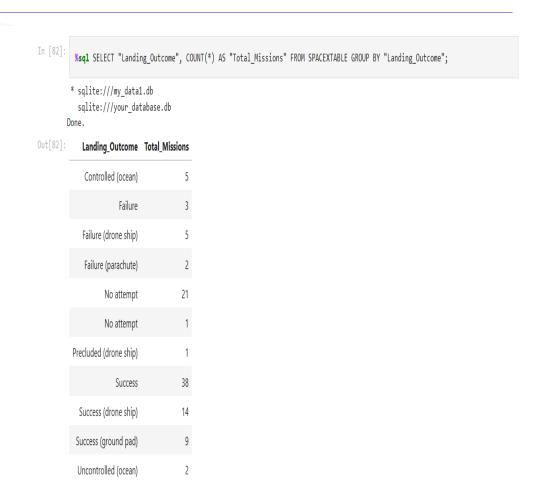
- Data Sources: Combined dataset\_part\_2.csv and dataset\_part\_3.csv.
- **Data Cleaning**: Verified no missing values; ensured data consistency.
- Feature Preparation: Standardized numerical features (e.g., Payload Mass) and encoded categorical variables.
- Target Variable: Created Class column (1 = Landed, 0 = Not Landed).
- Data Split: Divided into 80% training and 20% testing sets.



## **SQL Analysis: Key Findings**

- Launch Success Rate: Queried data to calculate the proportion of successful landings.
- Feature Impact: Identified factors like Booster Version and Launch Site influencing outcomes.
- Launch Site Comparison: SQL queries revealed variations in success rates across launch sites.
- **Payload Analysis**: Filtered payload ranges and their impact on landing success.

SQL provided insights into patterns and relationships, forming the basis for machine learning predictions.





## **Exploratory Data Analysis (EDA) with Visualizations**

#### Correlation Analysis:

Heatmap visualizations highlighted relationships between features such as Payload Mass and success rate.

#### Landing Outcomes:

Bar plots displayed the distribution of successful and unsuccessful landings across launch sites.

#### Payload Distribution:

Histograms showed payload mass ranges for successful vs. unsuccessful landings.

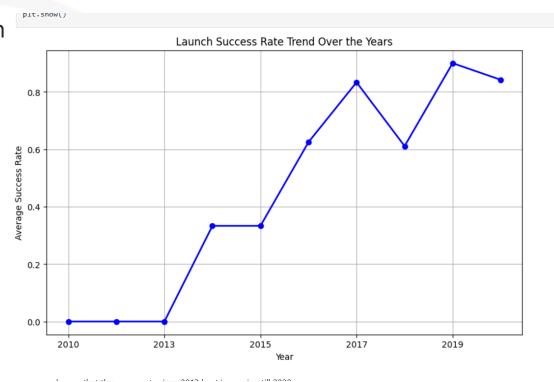
#### Temporal Trends:

Line charts analyzed trends over time, linking launch dates to landing success.

#### Feature Insights:

Scatter plots provided insights into Payload Mass and booster version impacts.

Visualizations revealed crucial patterns, guiding feature selection for machine learning.



## **Interactive Visualization and Dashboard**

cess Launches By Site

#### Interactive Maps:

Visualized launch sites and their success rates using interactive maps (e.g., Folium) for better spatial analysis.

#### Success Rate Filters:

Dashboards included filters for booster versions, launch sites, and payload mass ranges to explore success patterns dynamically.

### • Customizable Graphs:

Users could switch between bar charts, scatter plots, and line graphs to analyze trends interactively.

Payload and Outcome Correlation:

Interactive scatter plots allowed users to investigate the relationship between payload mass and success probability.

### Real-Time Insights:

Dashboards enabled users to explore the dataset dynamically, gaining actionable insights for predictive modeling.

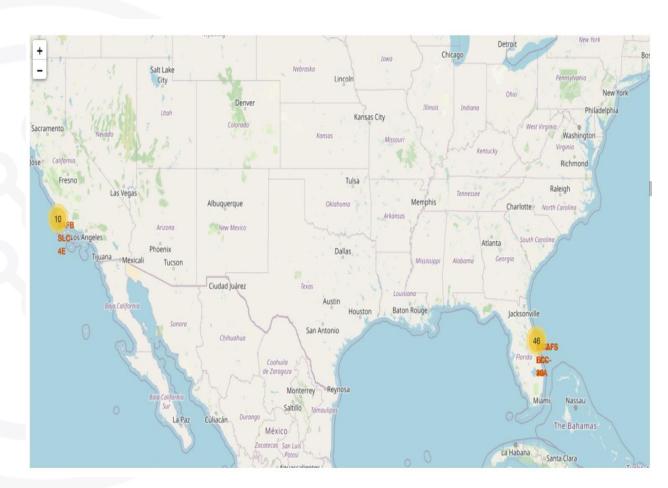
This feature significantly enhanced exploratory analysis, making data insights more accessible and user-friendly.





## **Interactive Visualization with Folium**

- •Mapped all SpaceX launch sites with interactive markers for success/failure rates.
- •Integrated hover tooltips showing payload mass and mission outcomes.
- •Added dynamic layers to categorize launches by outcome.
- •Enabled zoom and navigation for spatial pattern analysis.

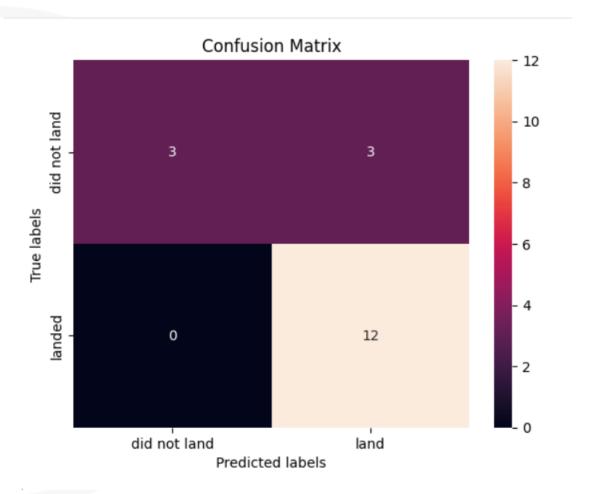




## **Machine Learning Predictions Lab**

#### **❖ Data Preprocessing**:

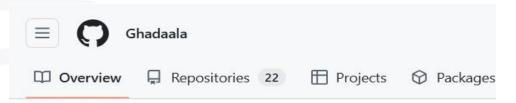
- &. Standardized the dataset using consistency.
- &. Split data into training (80%) and test (20%) sets.
- **❖ Model Training**:
- &. Used Logistic Regression, Support Vector Machines (SVM), Decision Trees, and K-Nearest Neighbors (KNN).
- GridSearchCV : for optimal model configurations.
- **Best Model Performance:**
- &. Logistic Regression achieved the best accuracy on test data (83.33%).
- &. Confusion matrix analysis highlighted minimal false positives and high true positive rates.
- Insights:
- &. Predictive models effectively classify whether the first stage will land, assisting in cost-saving strategies.



## **DASHBOARD**

### https://github.com/Ghadaala



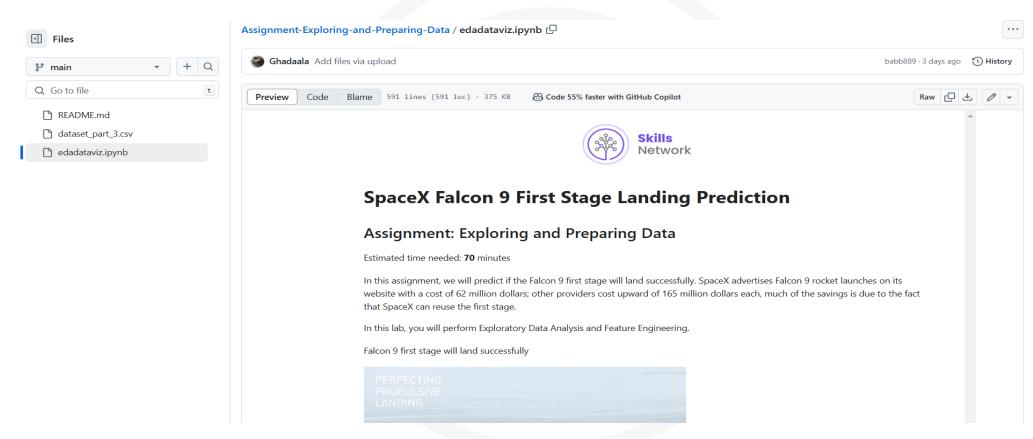




Ghada ALabdullah Ghadaala

## **Assignment: Exploring and Preparing Data**

https://github.com/Ghadaala/Assignment-Exploring-and-Preparing-Data

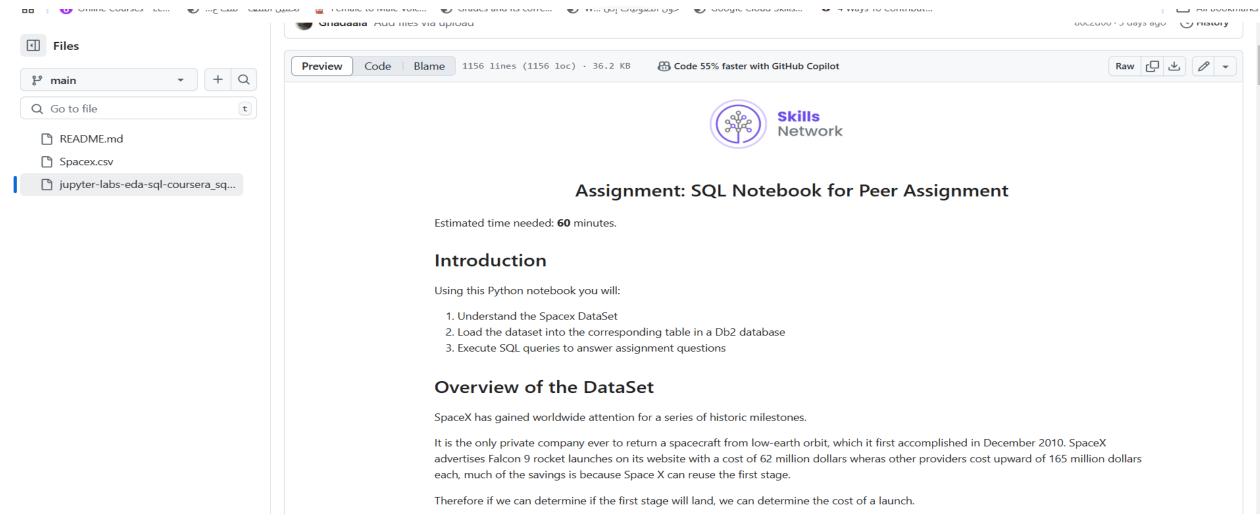






## Assignment: SQL Notebook for Peer Assignment

https://github.com/Ghadaala/Assignment-SQL-Notebook-for-Peer-Assignment/blob/main/jupyter-labs-eda-sql-coursera\_sqllite%20(1).ipynb

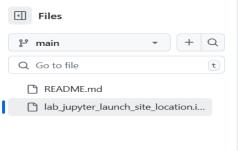






### Interactive Visual Analytics with Folium

https://github.com/Ghadaala/Hands-on-Lab-Interactive-Visual-Analytics-with-Folium/blob/main/lab\_jupyter\_launch\_site\_location.ipynb



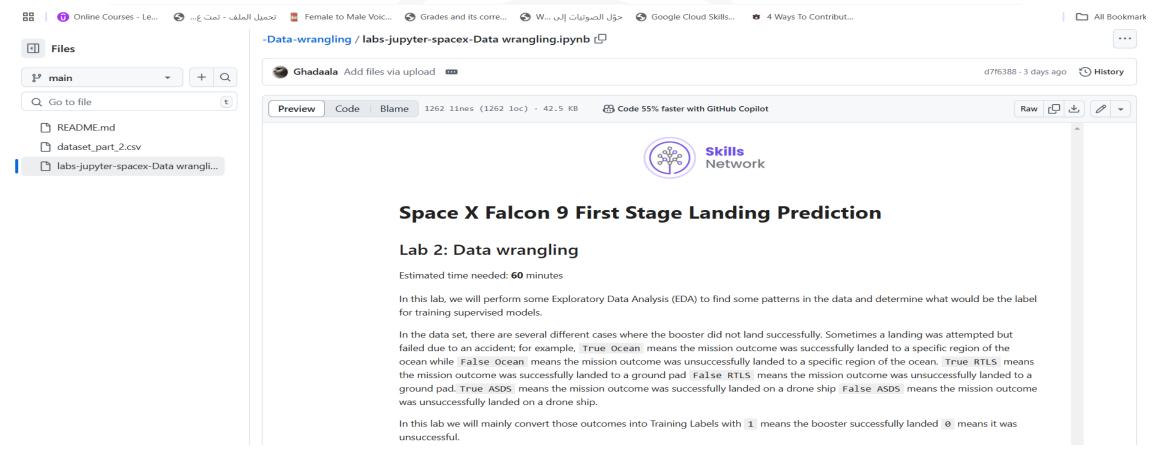






## **Data wrangling**

https://github.com/Ghadaala/-Data-wrangling/blob/main/labs-jupyter-spacex-Data%20wrangling.jpynb

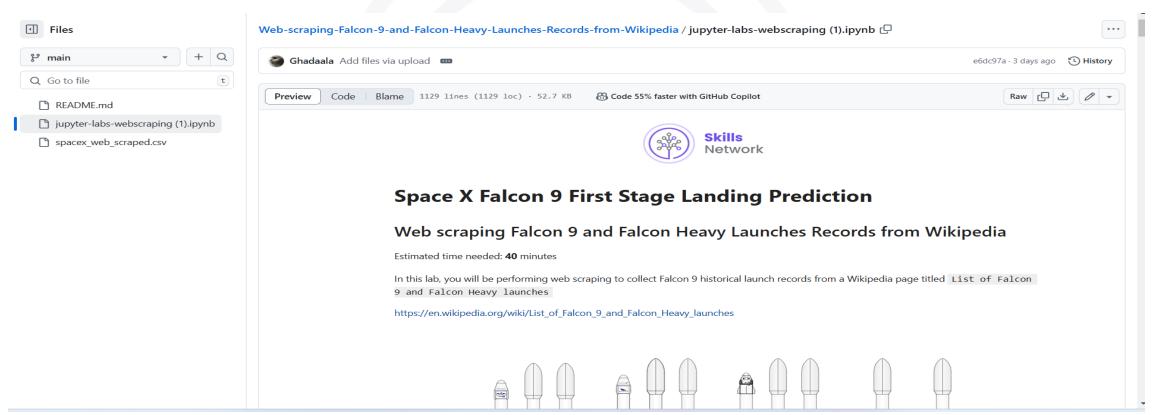






# Web scraping Falcon 9 and Falcon Heavy Launches Records from Wikipedia

https://github.com/Ghadaala/-Data-wrangling/blob/main/labs-jupyter-spacex-Data%20wrangling.ipynb

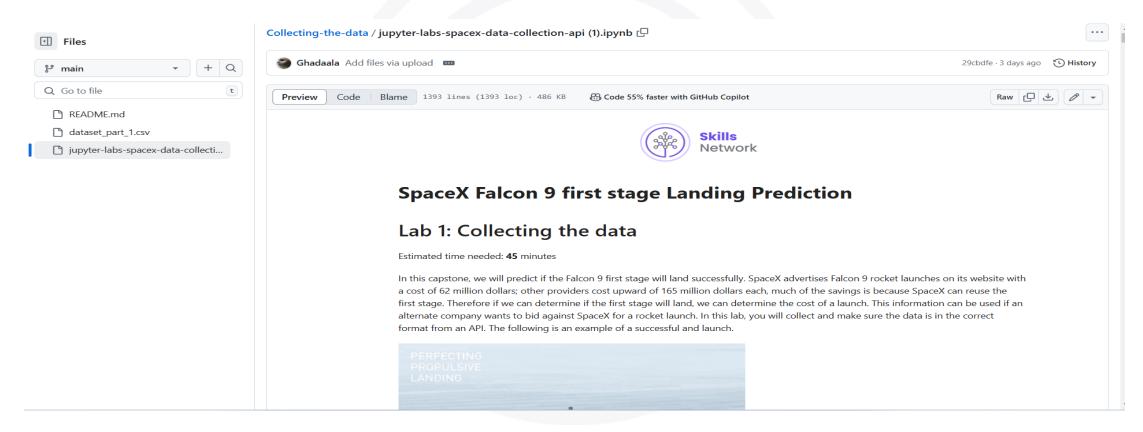






## Collecting the data

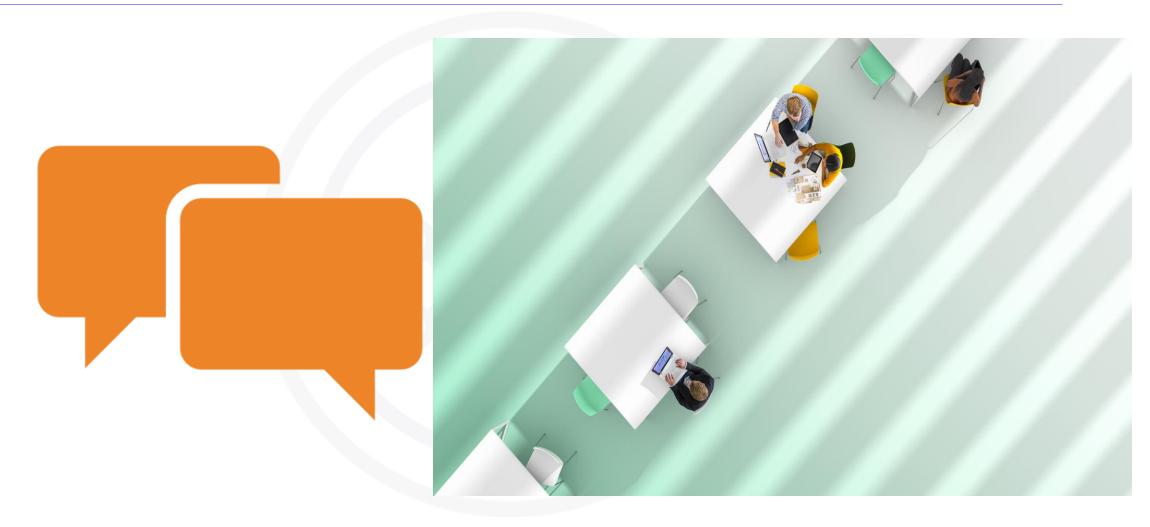
https://github.com/Ghadaala/Collecting-the-data/blob/main/jupyter-labs-spacex-data-collection-api%20(1).ipynb







## **DISCUSSION**







### **OVERALL FINDINGS & IMPLICATIONS**

### **Findings**

- Decision Tree Classifier: Best performer with 93.33% accuracy.
- Logistic Regression: Reliable with 83.33% accuracy.
- **SVM**: Competitive at **84.82%** accuracy but resource-heavy.
- KNN: Least effective at 76.67% accuracy

### **Implications**

- Cost Prediction: Supports pricing strategies for rocket launches.
- Model Choice: Decision Tree is ideal for accuracy; Logistic Regression for simplicity.
- Business Impact: Enhances decision-making for space providers and clients.

### CONCLUSION



- Best Model: Decision Tree with 93.33% accuracy.
- **Key Insight**: Models can predict rocket landing reliability, aiding cost efficiency.
- **Business Impact**: Supports competitive bidding and strategic decision-making.
- Future Direction: Explore more data and model refinements.

### **APPENDIX**



- Sample SQL queries or Python code snippet
- > References to any external sources or datasets used.
- videos that can help explain key points in project

## > Sample SQL queries or Python code snippet

Edit View Run Kernel Tabs C Task 1 Q Filter files by name Display the names of the unique launch sites in the space mission / module\_2 / SQLLIte / Name Last Modified [76]: # Run this if you're using SQLAlchemy connection Image: Ima 2 days ago %sql SELECT DISTINCT "Launch Site" FROM SPACEXTABLE; my\_data1.db 2 days ago your\_datab... 3 days ago \* sqlite:///my data1.db sqlite:///your\_database.db Done. [76]: Launch Site CCAFS LC-40 VAFB SLC-4E KSC LC-39A CCAFS SLC-40





## References to any datasets used.

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## videos that can help explain key points in project





