

Winning Space Race with Data Science

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Outline

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

- Project Goal: Predict Falcon 9 first-stage landing success using features like payload mass, launch site, and booster version.
- Summary of methodologies:
- Data Collection: SpaceX API.
- Data Wrangling: Cleaned and preprocessed data.
- Models Used: Logistic Regression, Decision Trees, KNN, SVM.
- Evaluation: Accuracy and confusion matrix.
- Summary of all results:
- EDA Insights: Payload mass negatively impacts landing success.
- Best Model: Decision Tree with Accuracy: 0.9655172413793104 accuracy.

Introduction

- Project background and context:
- Overview of SpaceX: SpaceX is a private aerospace company, and Falcon 9 is one of its primary rockets.
- Objective: Predict the likelihood of successful landings for Falcon 9 first-stage rockets.
- Problems you want to find answers:
- What factors influence the success of Falcon 9's first-stage landing?
- Which machine learning model best predicts landing success?



Methodology

- Data Collection: SpaceX API.
- Data Wrangling: Cleaned and preprocessed data.
- Models Used: Logistic Regression, Decision Trees, KNN, SVM. Evaluation: Accuracy and confusion matrix.
- Key Results:
- EDA Insights: Payload mass negatively impacts landing success.
- Best Model: Decision Tree with Accuracy: 0.9655172413793104 accuracy.

Data Collection and Data Wrangling

Data Sources: SpaceX API (retrieved using requests library).

Data Preprocessing:

Handle missing values using mean imputation for numeric variables. Drop irrelevant columns and rows with multiple payloads/cores.

Feature engineering: Calculate payload mass, and extract launch year.

Libraries used: Pandas, NumPy, requests, SQL.

EDA and Interactive Visual Analytics Methodology

- EDA: Identify patterns and relationships in the data.
- **Tools:** Pandas for data manipulation, Matplotlib, Seaborn for visualization.
- Interactive Visual Analytics: Tools: Plotly for interactive charts, Folium for mapping.
- **Example:** Visualize success rates by launch site using an interactive map.

Predictive Analysis Methodology

 Model Selection: Try multiple classification models (Logistic Regression, Decision Trees, KNN, SVM).

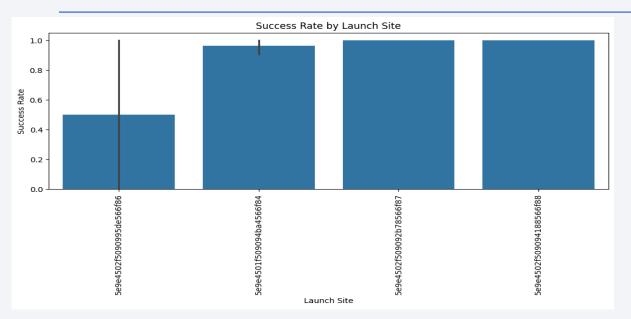
• Evaluation Metrics: Use accuracy, precision, recall, and confusion matrix to evaluate models.

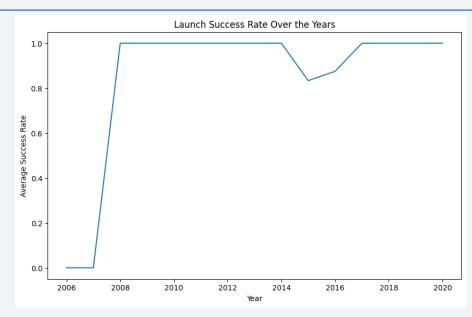
 Model Selection: Chose the best model (Logistic Regression in this case) based on highest accuracy.

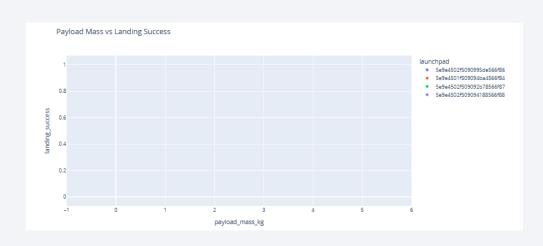
EDA with Visualization Results

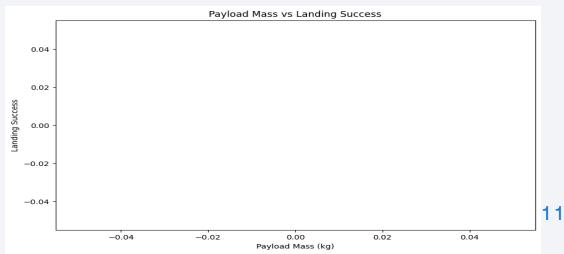
- Flight Number vs. Launch Site (scatter plot).
- Payload vs. Launch Site (scatter plot).
- Success Rate vs. Orbit Type (bar chart).
- Flight Number vs. Orbit Type (scatter plot).
- Payload vs. Orbit Type (scatter plot).
- Launch Success Yearly Trend (line chart).

EDA with Visualization Results









EDA with SQL Results

Present 10 SQL queries related to your analysis:

```
static_fire_date_utc static_fire_date_unix tbd net window \
0 2006-03-17T00:00:00.000Z
                                  1.142554e+09 False False
                     None
                                           NaN False False
                     None
                                           NaN False False
  2008-09-20T00:00:00.000Z
                                  1.221869e+09 False False
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                   rocket success \
0 5e9d0d95eda69955f709d1eb
1 5e9d0d95eda69955f709d1eb
2 5e9d0d95eda69955f709d1eb
3 5e9d0d95eda69955f709d1eb
4 5e9d0d95eda69955f709d1eb
                                         details crew ships ... \
0 Engine failure at 33 seconds and loss of vehicle [] [] ...
1 Successful first stage burn and transition to ... [] [] ...
2 Residual stage 1 thrust led to collision betwe... [] [] ...
  Ratsat was carried to orbit on the first succe... [] [] ...
                                            None [] [] ...
  links.reddit.media links.reddit.recovery links.flickr.small \
              None
                                   None
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   links.flickr.original
                                                        links.presskit \
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                    [] http://www.spacex.com/press/2012/12/19/spacexs...
                              links.webcast links.youtube_id \
0 https://www.youtube.com/watch?v=0a 00nJ Y88
                                                0a 00nJ Y88
                                                Lk4z02wP-Nc
1 https://www.youtube.com/watch?v=Lk4zQ2wP-Nc
2 https://www.youtube.com/watch?v=v0w9p3U8860
                                                v@w9p3U8860
3 https://www.youtube.com/watch?v=dLQ2tZEH6G0
                                                dLQ2tZEH6G0
4 https://www.voutube.com/watch?v=vTaIDooc80g
                                                vTaIDooc80g
0 https://www.space.com/2196-spacex-inaugural-fa...
  https://www.space.com/3590-spacex-falcon-1-roc...
   http://www.spacex.com/news/2013/02/11/falcon-1...
              https://en.wikipedia.org/wiki/Ratsat
  http://www.spacex.com/news/2013/02/12/falcon-1...
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links.wikipedia fairings
              https://en.wikipedia.org/wiki/DemoSat
              https://en.wikipedia.org/wiki/DemoSat
2 https://en.wikipedia.org/wiki/Trailblazer (sat...
              https://en.wikipedia.org/wiki/Ratsat
             https://en.wikipedia.org/wiki/RazakSAT
[5 rows x 42 columns]
Warning: 'payload mass kg' column is all NaN. Cannot fill missing values with mean.
      static_fire_date_utc static_fire_date_unix tbd net window \
0 2006-03-17T00:00:00.000Z
                                  1.142554e+09 False False
                                       NaN False False
3 2008-09-20T00:00:00.000Z
                                  1.221869e+09 False False 0.0
4 None
                                           NaN False False 0.0
5 2010-03-13T00:00:00.000Z
                                  1.268438e+09 False False 0.0
                   rocket success \
0 5e9d0d95eda69955f709d1eb False
1 5e9d0d95eda69955f709d1eb
3 5e9d0d95eda69955f709d1eb
4 5e9d0d95eda69955f709d1eb
5 5e9d0d95eda69973a809d1ec True
                                         details crew ships ... \
0 Engine failure at 33 seconds and loss of vehicle [] [] ...
1 Successful first stage burn and transition to ... [] [] ...
3 Ratsat was carried to orbit on the first succe... []
                                            None [] [] ...
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                                   links.presskit \
4 http://www.spacex.com/press/2012/12/19/spacexs...
5 http://forum.nasaspaceflight.com/index.php?act...
                              links.webcast links.youtube id \
0 https://www.youtube.com/watch?v=0a 00nJ Y88
1 https://www.youtube.com/watch?v=Lk4zQ2wP-Nc
                                                Lk4z02wP-Nc
3 https://www.voutube.com/watch?v=dLO2tZEH6G0
4 https://www.voutube.com/watch?v=vTaIDooc8Og
                                                yTaIDooc80g
5 https://www.youtube.com/watch?v=nxSxgBKlYws
                                                nxSxgBK1Yws
                                    links.article \
0 https://www.space.com/2196-spacex-inaugural-fa...
1 https://www.space.com/3590-spacex-falcon-1-roc...
              https://en.wikipedia.org/wiki/Ratsat
4 http://www.spacex.com/news/2013/02/12/falcon-1...
5 http://www.spacex.com/news/2013/02/12/falcon-9...
```

Interactive Map with Folium Results

Use Folium to create an interactive map showing launch sites and success rates.

Include proximity analysis (distance to infrastructure).

•Include charts like a pie chart for launch success rates and scatter plots for payload vs. outcome.

Plotly Dash Dashboard Results

• Created a Plotly Dash dashboard to display key metrics interactively.

• Included charts like a pie chart for launch success rates and scatter plots for payload vs. outcome.

Predictive Analysis Results

- Present results for each classification model (e.g., Logistic Regression, Decision Tree, KNN, SVM).
- Showcase confusion matrices for each model to compare their performance. Include performance metrics: Accuracy, Precision, Recall, and F1 Score.
- Comparison: Discuss which model performed best based on the evaluation metrics
- Model Comparison: Decision Tree: Shows the highest accuracy at Decision
 Tree with Accuracy: 0.9655172413793104
- Decision Tree: Performed reasonably but overfitted slightly on training data. KNN: Good accuracy but struggled with large datasets.
- SVM: Performed well but with higher training time. Confusion Matrix for Logistic Regression: Use scikit-learn to generate and display the confusion matrix.

Conclusion

- Summary of Findings:
- EDA Insights: We found that payload mass is inversely correlated with landing success, and certain launch sites have higher success rates. SQL Insights: The successful launch sites were identified, and queries revealed payload trends for successful landings.
- Predictive Model Results: The decision tree model performed best with 96% accuracy, effectively predicting landing success.
- Future Work: Explore adding weather data and other external factors for more accurate predictions. Try more advanced machine learning models (e.g., Random Forest, XGBoost) to improve prediction accuracy. Real-time prediction: Implement a real-time prediction system for upcoming Falcon 9 launches.

Creativity and Insights

- Innovative Visualizations: Created an interactive map using Folium to visualize launch sites and their success rates. Developed a Plotly Dash dashboard to explore key metrics interactively.
- Extra Insights: Discovered that newer booster versions (e.g., Falcon 9 v1.1) had a better chance of successful landings, which could be useful for future launch planning.
- Visualized correlations between launch site locations and success rates, showing that proximity to specific infrastructure (e.g., coastlines) could influence landing success.

Final Thoughts

- Reflection on Challenges: The main challenge was dealing with missing data and imbalanced class distribution for landing success. Model tuning was also a challenge, but it was a valuable learning experience.
- What I Learned: The importance of data wrangling in making data ready for modeling. How different machine learning algorithms behave when dealing with classification problems like this one.
- **Practical Applications:** The insights from this analysis could help SpaceX improve their landing strategy by prioritizing specific launch sites or adjusting their booster designs.
- The predictive model could be used to provide early warnings for potential landing failures, helping improve launch safety and cost-efficiency.

References

- **Coursera**. (n.d.). *IBM Data Science Professional Certificate*. Retrieved from https://www.coursera.org/professional-certificates/ibm-data-science
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Thank you