



PREDICTING PRICES OF FLIGHTS TO AND FROM ATL

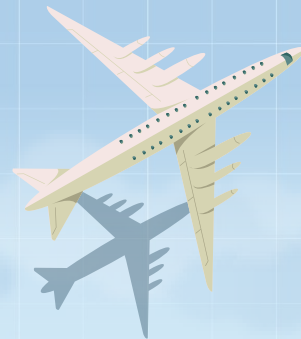
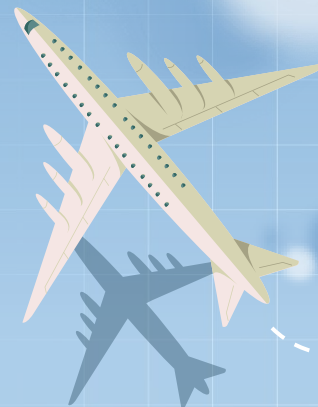
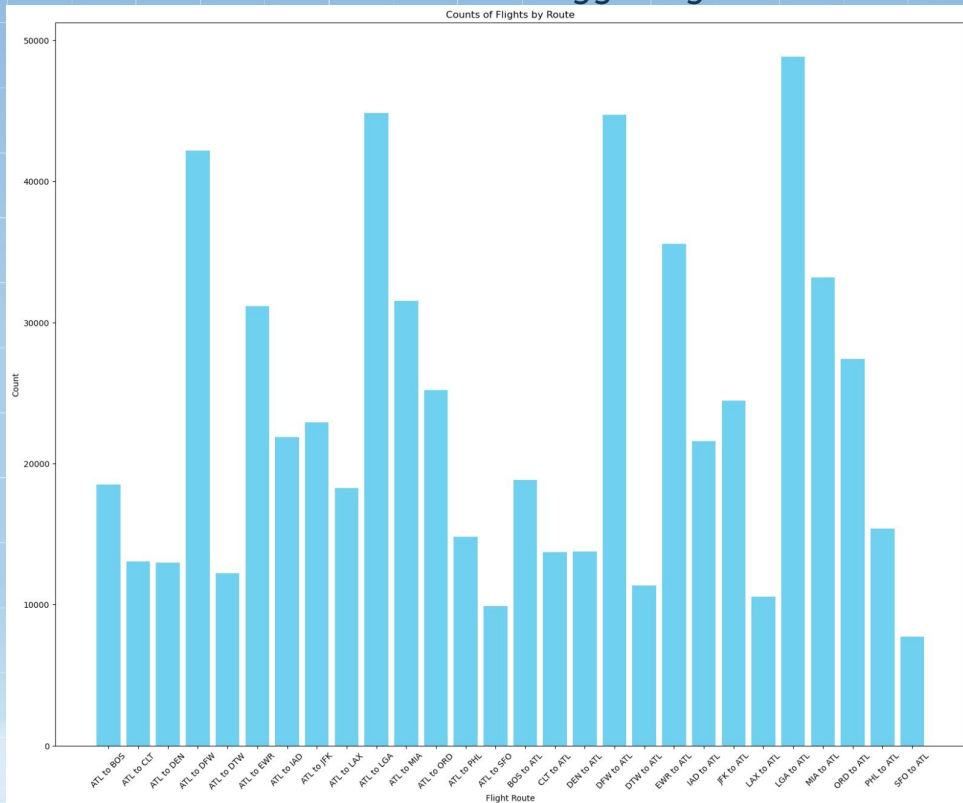
CS 4641 Group 31 Final Presentation

PROJECT DEFINITION

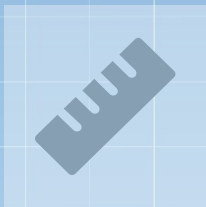
- Accurate flight cost prediction is vital in today's dynamic travel industry.
- Anticipating flight ticket prices significantly influences consumer choices and travel experiences.
- Our project aims to develop a machine learning model for effective flight cost prediction.
- Specifically geared towards Georgia Tech students, we focus on predicting costs for flights to and from ATL airport.
- To enhance specificity, our project excludes unrelated flights and is limited to economy-class seats.

DATASET

from a Kaggle Flight Prices Dataset



CLEANING



NORMALIZATION

- Numerical representations
- Feature extraction



PRUNING

- Removing features representing the same data
- Removing data irrelevant to our problem definition



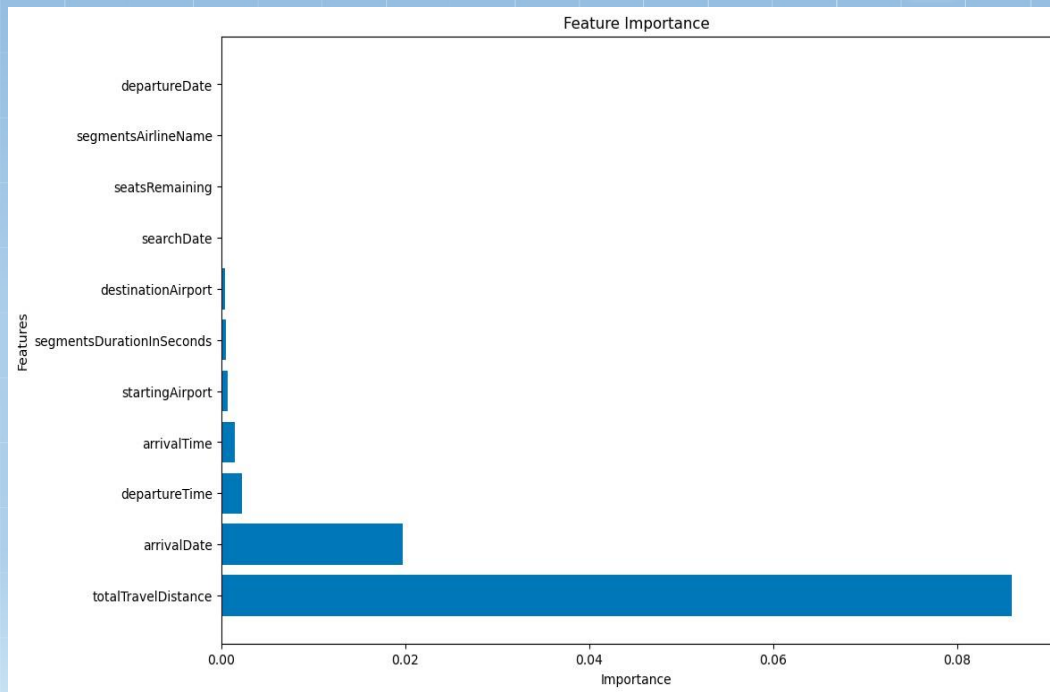
MISSING DATA

- Filling in missing values

PREPROCESSING

Lasso Feature Selection

- Determine feature importance
- Feature reduction





FINAL FEATURES

(in order of importance)

| | |
|----------------------------------|---|
| totalTravelDistance | Total travel distance in miles |
| arrivalDate | A number between 1-365 corresponding to the day of the year |
| departureTime | A number between 0-1439 corresponding to the time of the day in minutes |
| arrivalTime | A number between 0-1439 corresponding to the time of the day in minutes |
| startingAirport | Three-character IATA airport code for the initial location |
| segmentsDurationInSeconds | String containing the duration of the flight, in seconds, for the trip |
| destinationAirport | Three-character IATA airport code for the arrival location |

METHODS

01

KNN

Supervised

02

LINEAR REGRESSION

Supervised

03

RANDOM FOREST

Supervised



An illustration of two yellow and white commercial airplanes flying in a blue sky with soft, white clouds. A dashed white line traces a path through the sky, starting from the bottom left, looping around, and extending towards the top right. The background has a light blue grid pattern.

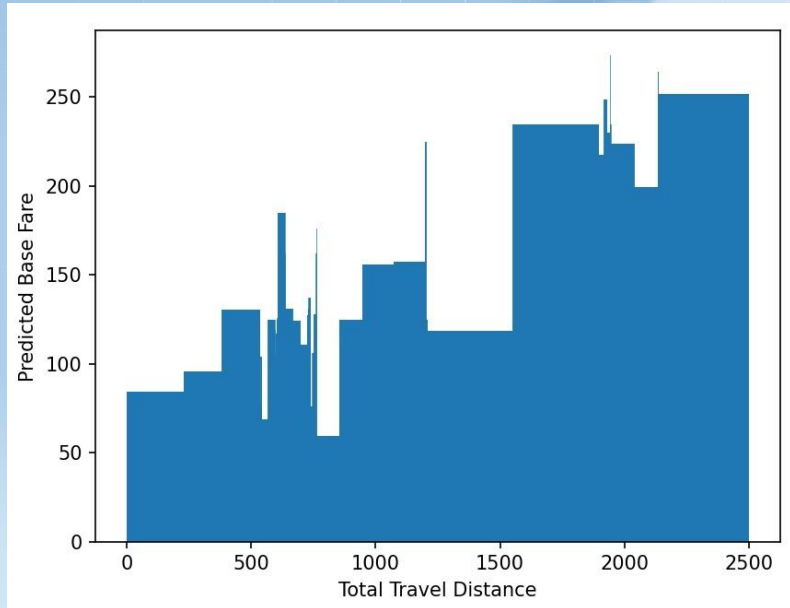
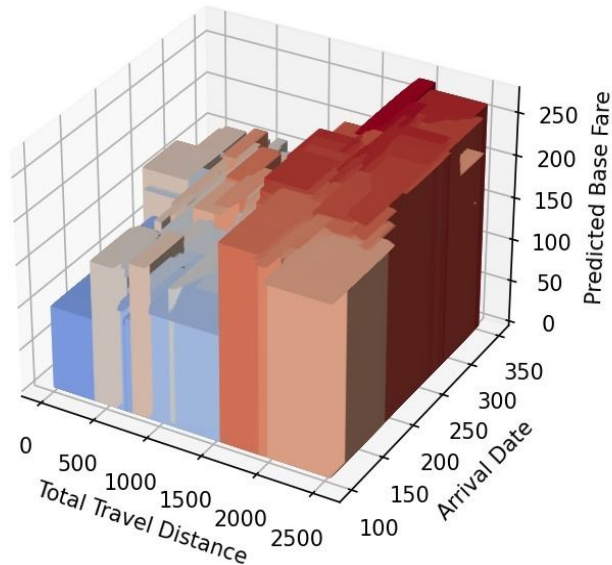
01

K NEAREST NEIGHBORS

Accuracy: $\approx 90\%$

KNN: RESULTS & ANALYSIS

- MSE: 208.86 - 211.54
- R-squared: 0.9041 - 0.9065





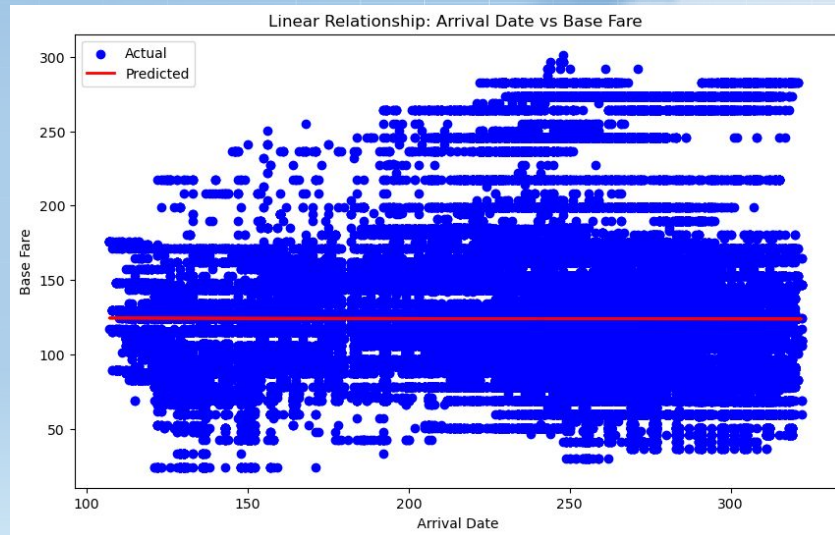
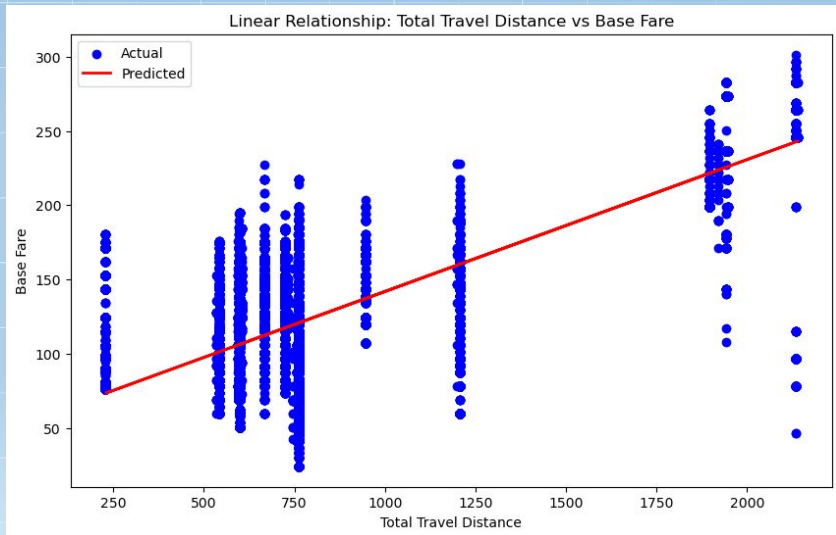
02

LINEAR REGRESSION

Accuracy: $\approx 50\%$

LR: RESULTS & ANALYSIS

- MSE: 1076
- R-squared: 0.51
- Reason for inaccuracy: non-linear relationships for some features



An illustration of two yellow and white commercial airplanes flying in a blue sky with soft white clouds. A dashed white line winds through the sky, starting from the bottom left, looping around the bottom airplane, and extending towards the top right. The background has a light blue grid pattern.

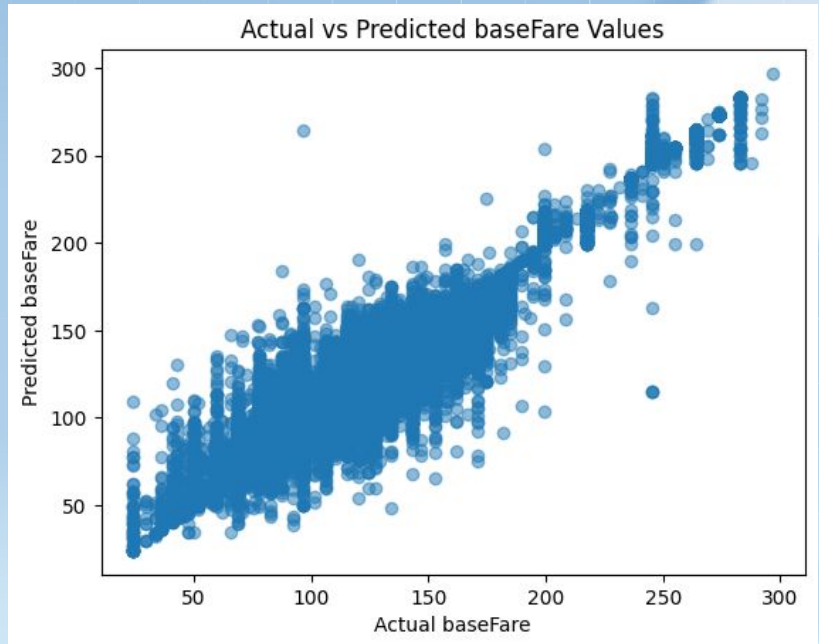
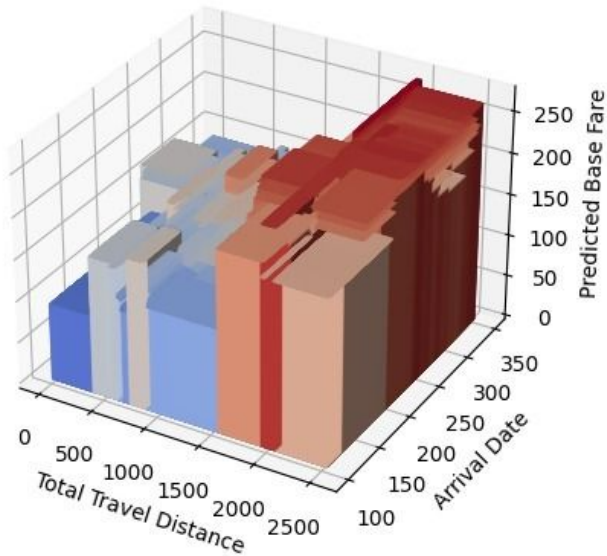
03

RANDOM FOREST

Accuracy: $\approx 96\%$

RF: RESULTS & ANALYSIS

- MSE: 88.46
- R-squared: 0.9599



MODEL COMPARISON

| Aspect | KNN | Linear Regression | Random Forest |
|------------------------|--|---|---|
| Non-linearity Handling | Effective in capturing complex, non-linear relationships in flight features. | Assumes a linear relationship between variables, may struggle with non-linear correlations. | Capable of capturing non-linear relationships, robust performance observed. |
| Prediction Accuracy | Achieved over 90% accuracy with optimized k-value and feature selection. | Achieved approximately 50% accuracy, reliable predictor despite non-linear correlations. | Achieved outstanding accuracy up to 95.99%, with consideration for computational costs. |
| Interpretability | Limited interpretability due to the complex nature of the algorithm. | May struggle with interpretability, especially in the presence of non-linear correlations. | Limited interpretability, challenging to explain the decision-making process. |
| Overfitting Concerns | Prone to overfitting, especially with small k-values. Requires careful tuning. | Potential for overfitting, especially with correlated features. | May exhibit overfitting, but observed performance on testing data alleviated concerns. |



DISCUSSION & CONCLUSION



DISCUSSION

- KNN: showed high accuracy, but lacked interpretability
- Linear Regression: not reliable w/ less accuracy
- Random Forest: high accuracy, but at the expense of some interpretability

FUTURE WORK / NEXT STEPS

Test on future flight prices and explore prices for flight routes outside Atlanta



REFERENCES

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- M. Mulkalla, Deepika and A. Joshi, "Predicting the Fare of a Flight Ticket with Machine Learning Algorithms," *2022 International Conference on Cyber Resilience (ICCR)*, Dubai, United Arab Emirates, 2022, pp. 1-5, doi: 10.1109/ICCR56254.2022.9995886.
- S. N. Prasath, S. Kumar M and S. Eliyas, "A Prediction of Flight Fare Using K-Nearest Neighbors," *2022 2nd International Conference on Advance Computing and Innovative Technologies in Engineering (ICACITE)*, Greater Noida, India, 2022, pp. 1347-1351, doi: 10.1109/ICACITE53722.2022.9823876.

