

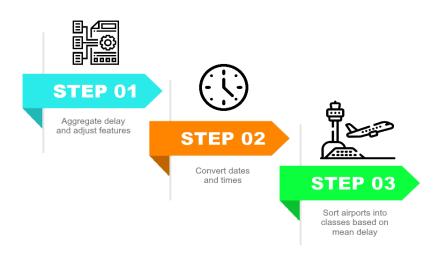
Predicting Flight Delay for August 2022

A penalized regression approach

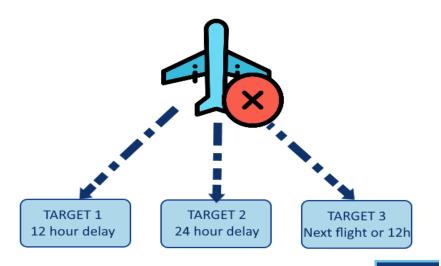


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Quick Preprocessing Overview



What about cancelled flights?



Feature adjustment

Dropped features

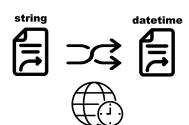
- Delays/Cancelled
- Arrival/Departure times, Taxi/Wheel deployment, Elapsed/Air time
- Flight and Tail number, Cancellation reason

Reason

- Replacement
- Would not be available before departure of future flights
- Irrelevant to the predictions

Date and time conversion

- Workable format
- Adjust to timezone
- Day of the week



Airport classes

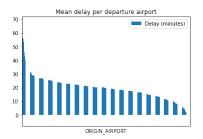


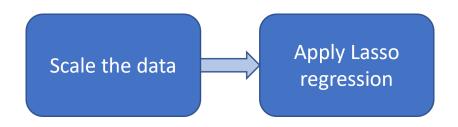
Figure: Mean delay per departure airport



Figure: Mean delay per arrival airport

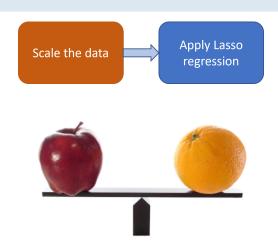
Top and bottom 5% of airports \rightarrow High and Low class resp. Middle 90% and missing obs \rightarrow Moderate class

The model



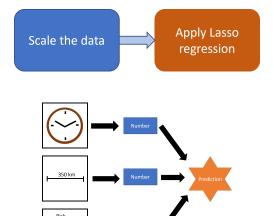
Model and model performance

The model



 $apples- and- oranges.\ \textit{Locatify}.\ https://locatify.com/apples-to-oranges-4/.$

The model



Performance

Performance measures (on test set)

- $R^2 \approx 0.44$
- MSE ≈ 6974

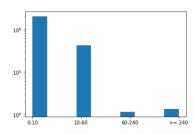


Figure: The distribution of the errors

Prediction

Predictions on delay time (min)

- 0-10 (1.46%)
- 10-30 (65.46%)
- 30-60 (33.03%)
- 60+ (0.05%)

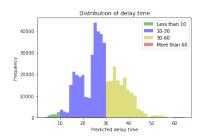


Figure: The distribution of delay time

Prediction

Predictions by airline

- HA: highest (29.6 min)
 - ...
- B6: lowest (26.6 min)

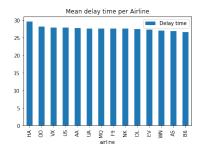


Figure: Mean delay time per airline

Notes and suggestions for improvement



Notes and suggestions for improvement

- 1. Cancellation reasons?
- 2. Peak season or off season?
- 3. Public Holidays?
- 4. Subjectivity of classes?
- 5. Group variables?

Conclusion



Takeaways

- Usefulness of this model
- Why did we pick this model?
- Limitations of Model
- Important Assumptions/ Considerations

Next Steps



- Practical Use Cases
- Additional Data supplementation
- Suggested next steps