

Project Design Phase-II

Solution Requirements(Functional & Non-Functional)

Date	17 October 2022
Team ID	PNT2022TMID34340
Project Name	Project - Developing a Flight Delay prediction model using Machine Learning
Maximum Marks	2 Marks

Abstract:

Predicting flight delays accurately is essential for building a more effective airline industry. All participants in commercial aviation must consider their predictions while making decisions. Additionally, the complexity of the air transportation system, the variety of forecast approaches, and the abundance of flight data made it difficult to construct precise prediction models for flight delays. To optimize flight operations and reduce delays, the model's primary goal is to estimate flight delays accurately.

Flight arrival delays can be predicted using a machine-learning algorithm. Rows of feature vectors, such as departure date, delay, travel time between the two airports, and scheduled arrival time, provide input to our algorithm. The decision tree classifier is then used to determine whether or not the flight arrival will be delayed. A flight is deemed to be delayed when there is more than a 15-minute gap between the scheduled and actual arrival timings. Additionally, we contrast the decision tree classifier with logistic regression and a straightforward neural network for various figures of merit.

LITERATURE SURVEY

Literature	Methodology	Merits	Demerits
Study of Flight Departure Delay and Causal Factor Using Spatial Analysis	Assuming delay as a spatially dependent variable, finds a delay distribution pattern to predict delay	Considers spatial factors, people, day types, and time ranges of a day to contribute to the prediction	Some of the attributes considered cannot be obtained on large scale in real-time
Flight Delay Prediction System	Supervised Machine Learning algorithm using Naive Bayes	Considers independence among the predictors making the system scalable. Good for real-time prediction	Does not take into account the impact of unprecedented reasons such as major calamities in flight delays
Flight delay forecasting and analysis of direct and indirect factors	LSTM network with an attention mechanism to remember spatial dependencies	Direct and indirect causing factors are weighed differently.	Air interaction of flights not taken into account
A deep learning approach to flight delay prediction	A deep RNN and LSTM approach to prediction; uses limited data attributes	Predicting two sections namely day prediction and flight prediction seems more reasonable and can give more insights for the airport managers to make necessary arrangements	Air traffic/flight interaction doesn't play great roles. Biased towards weather attributes.
Flight delay prediction based on aviation big data and machine learning	Comparison of LSTM and Random forest; Uses ADS-B data for improved accuracy.	The use of ADS-B can be seen as promising. Showed that LSTM suffers from over fitting on test set.	Deployment of ADS-B is hectic. More data handling takes place.
Prediction of weather induced airline delays based on machine learning algorithms	Experimented with predicting delay using supervised machine learning algorithms. Uses SMOTE for weaker class sampling	Found weather causes to be amounting to a significant percent of delay	Not taking spatial dependencies into account. Amount of delay could have been found.

Flight delay prediction from a spatial and temporal perspective	ST-Random Forest for flight delay prediction using spatial features of aviation network and temporal correlation of weather condition and airport crowded on flight delays.	A real-time, highly accurate prediction system that guarantees the influence of the air traffic network in the prediction.	Over fitting might occur due to LSTM
Predicting flight delay based on multiple linear regression	A multiple linear regression algorithm to predict delay	Both airline and weather features are taken into consideration. The methodology used in this gives better results compared to the Naive- Bayes and C4.5 approach	Predicts only the flights which are delayed above 30 minutes
Airline Flight Delay Prediction Using Machine Learning Models	Comparison among 7 classification machine learning algorithms	Among the considered algorithms using 4 Performance indicators decision tree was found to be the best in predicting flight delays	The data imbalance issue even though handled through weighted evaluation methods does have A significant effect onto`performance of the algorithms

References :

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