

Flight Price Prediction

- Importing the important libraries for reading the dataset.
 - Because data is present in two different sets.
 - Joining the both dataset in single variable.
 - Checking for null values.
 - Checking the datatypes of dataset, in which form data is present.
 - All columns are present as object datatype except one.
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- Now, we import encoder for encoding the required columns.
 - Importing label encoder and ordinal encoder.
 - Because target variable is present as object datatype, so we have to encode it also.
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- Now we import libraries for visualization the dataset.
 - First, we plot distribution plot.
 - Second, we plot boxplot.
 - By distribution plot, data is shown which is not in normal shape.
 - By boxplot, some outliers are shown in dataset.
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- Now, we take out quantiles and inter quantile range, which help us to remove outliers.
 - After, removing outliers, again we check for data shape by distribution plot.
 - Now, after removing outliers, data starts looking in some normal shape.
 - By boxplot, it seems that a lot of outliers has been removed from dataset.
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- Now, we plot heatmap, for checking the correlation.
 - No such relation is shown by heatmap.
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- Now, its time to split the dataset into two variables.
 - Standardize the data and train test split.
 - Because target variable is present as continuous data so we apply Regression models on this dataset.

- Importing metrics for checking the model error.
- Now, first model, Linear Regression.
- By, Linear Regression, a very low score is given on train and test data both.
- Second model, Decision Tree.
- By second model, good accuracy is given on train data but on test data it became very low.
- Third model, Random Forest.
- By third model, good accuracy on train dataset is given, but low on test data.
- Fourth model, Support Vector Machine.
- By fourth model, a very-very low accuracy on both train and test data is given.
- Fifth model, Ada Boost.
- By fifth model, very low accuracy on both train and test dataset.
- Now, we do hyperparameter tuning on Random Forest and Ada Boost.
- After doing hyperparameter tuning, by Random Forest, accuracy on train dataset decreases and on test dataset increases.
- After doing hyperparameter tuning, by Ada Boost, accuracy remains same on both train and test dataset.
- Random Forest performed well, so we can save it by using pickle.