

# Approach for the competition

## EDA notes:

- Skewness of dependent column (Price)
- Possible outliers in price, Levy, Mileage attributes (Data points that lie outside the 99<sup>th</sup> percentiles).
- Finally, by visualization of the outliers, managed to delete some from Price and Levy.
- Idea regarding duplicates in dataset.
- Matching ID's in Train and Test set
- Low value counts on some production year attributes
- Having Engine volume and type, on single column
- Reason for special character in Levy column.

## Pre-processing steps:

- Replacing the Missing values in Levy column with different strategy (MIN, MAX, MEAN, -1, 0)
- Replacing with 0 value worked well for all the models based on trial and error method
- Separating Engine volume and Engine Type
- Removing the outliers (based on trial and error)
- Tried scaling the data (didn't work)

## Modelling:

- Checking the performance across all the regression models
- Checking the performance across all the regression models by taking log transformation of dependent variable (Price)
- Selecting best performing models on baseline.
- Tuning the best performance models (Did not work well after tuning, so manually tuned for few parameters only)
- Selecting cross-validation strategy.
- Feature Engineering (aggregation features with group by)
- Making the subset of features for different models based on cross validation score
- Keep track of CV score (since for small variation in local CV, there was large variation on Public Leader-board)
- Selected 3 models' Random forest, Extra Trees, Light GBM
- Made 3 subset of best features for 3 models.
- Check the feature importance plots each time.
- Check performance of individual models on Public LB (70% test data).
- Blend the models with weights based of Individual performance on Public LB
- There was data leak in train and test (didn't use that because it may lead to legality issue)

Due to limited submissions(3/day) couldn't try some things

## Things for further try:

- Frequency encoding the data
- Other possible transformations for dependent variable
- Categorical-to- Categorical feature creation.
- Tuning models for different subset of features.
- Generating predictions at each fold