

# 🌟 Machine Learning 🌟

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## Steps

1. 🧩 **Problem statement**
2. 📊 **Data collection**
3. ✍️ **Data cleaning**
  - Check categorical values of columns if they have **error in names** using `count.value()`. If any, replace their names
  - Check the **missing value** and separate them in dataset for easiness
    - Separate numerical columns and categorical columns in two dataframe
    - Check a column has  $\leq 25$  unique value will be considered as discrete data, otherwise it will be numerical data
    - Replace missing value with mode for categorical & discrete data and median for numerical data.
  - Describe general statistics
  - Reducing **Number of column**
    - Drop any unnecessary categorical column
    - Combine similar numerical columns if it is possible. And drop their original columns
4. 🚀 **Model Training**
  - **Train test splitting of data**
    - Separate input as X and output as y using drop function
    - Split X into **X\_train & X\_test** and **y\_train & y\_test** using `train_test_split` function from `sklearn.model_selection`
      - mention test data will be 20% or you can adjust its amount based on test performance
  - **Encoding** using ColumnTransfer
    - Separate Categorical and Numerical feature of X
      - Apply OneHotEncoder for categorical feature of X if not many features, ltherwise use LabelEncoding
      - Apply StandardScaler for numerical feature of X
  - **Transformation and Dataframe**
    - Apply transformation on X\_train usinf `fit.transformer` # reason to masking
    - Convert X\_train in Dataframe
    - Apply transformer on X\_test
    - Conver X\_test into DataFrame
  - **Model Training Algorithms**
    - Run RandomForestClassifier on X\_train & and y\_train import from `sklearn.ensemble`
    - Glve X\_test and X\_train data to make prediction using `model.prediction` # we already know their prediction/output
  - **Performance metrics for both training and testing data**
    - Then we check training performance by giving y\_train & predict y\_train data to performance metrics
    - Then we check test performance by giving y\_test & predict y\_test data to performance metrics
    - For Classification Algorithms

- accuracy\_score
- classification\_report
- precision\_score,
- recall\_score
- f1\_score
- roc\_auc\_score
- roc\_curve

## 5. 🔑 **Hypertuning**

- Add parameters into algorithms using model list
- RandomizedSearchCV
- Then run algorithms
- Plotting with the ROC AUC Curve