

# Explanation of used python library/necessity/ purpose.

## **1. from sklearn.model\_selection import train\_test\_split**

Used to split arrays or matrices into random train and test subsets

parameters are-

\*arrays sequence of indexables with same length / shape[0]

test\_size float, int or None, optional (default=None) etc.

Return type - List containing train-test split of inputs.

## **2. from sklearn.neural\_network import MLPClassifier**

It is multi-layer Perceptron classifier.

Model optimizes the log-loss function using LBFGS or stochastic gradient descent.

## **3. from sklearn.model\_selection import StratifiedKFold**

Stratified K-Folds cross-validator

Provides train/test indices to split data in train/test sets.

This cross-validation object is a variation of KFold that returns stratified folds. The folds are made by preserving the percentage of samples for each class.

## **4. from sklearn.model\_selection import GridSearchCV**

Exhaustive search over specified parameter values for an estimator.

Important members are fit, predict.

GridSearchCV implements a “fit” and a “score” method. It also implements “predict”, “predict\_proba”, “decision\_function”, “transform” and “inverse\_transform” if they are implemented in the estimator used.

The parameters of the estimator used to apply these methods are optimized by cross-validated grid-search over a parameter grid.

## **5. from tpot import TPOTClassifier**

TPOT is meant to be an assistant that gives an ideas on how to solve a particular machine learning problem by exploring pipeline configurations that we might have never considered, then leaves the fine-tuning to more constrained parameter tuning techniques such as grid search.

## **6. % matplotlib inline**

The % matplotlib inline is an example of predefined magic function in lpythonand

frequently used in interactive environments like jupyter notebook. `% matplotlib inline` makes your plot outputs appear and be stored within the notebook.

## **7. from scipy import signal**

Signal processing toolbox currently contains some filtering functions, a limited set of filter design tools, and a few B-spline interpolation algorithms for 1- and 2-D data.

## **8. import pickle**

Pickling is the serializing and de-serializing of python objects to a byte stream. Unpickling is the opposite.

## **9. sklearn.model\_selection import cross\_val\_score**

Evaluate a score by cross-validation

## **10. from sklearn import svm**

The implementation is based on libsvm. The fit time scales at least quadratically with the number of samples and may be impractical beyond tens of thousands of samples. For large datasets consider using `sklearn.svm.LinearSVC` or `sklearn.linear_model.SGDClassifier` instead, possibly after a `sklearn.kernel_approximation.Nystroem` transformer.

The multiclass support is handled according to a one-vs-one scheme.

## **11. from sklearn.metrics import precision\_recall\_fscore\_support**

Compute precision, recall, F-measure and support for each class

The precision is the ratio  $tp / (tp + fp)$   $tp$  is the number of true positives and  $fp$  the number of false positive. precision is intuitively the ability of the classifier not to label as positive a sample that is negative.

The recall is the ratio  $tp / (tp + fn)$   $tp$  is the number of true positives and  $fn$  the number of false negative. The recall is intuitively the ability of the classifier to find all the positive sample.