```
class imbbag.SMOTEBag(n_estimator=10, k_neighbors= 5, estimator=DecisionTreeClassifier())
```

## **SMOTE Bagging classifier.**

The SMOTE Bagging algorithm combines SMOTE (Synthetic Minority Over-sampling Technique) with bagging to handle imbalanced datasets..

Parameters: n estimator: int (default=10)

The number of nearest neighbors to search for.

k neighbors: int (default=10)

The number of nearest neighbors used to generate synthetic samples.

estimator : object (default= DecisionTreeClassifier())

An instance of a base classifier used in the ensemble.

# **Examples:**

```
from sklearn.tree import DecisionTreeClassifier
from sklearn. model_selection import train_test_split
from ImbBag import SMOTEBag

dataframe = read_csv('dataset.csv')
data = dataframe.values
X = data[:,:-1]
Y = data[:,:-1]
# split the dataset into training and test sets
X_train ,X_test ,y_train ,y_test = train_test_split (X, y, test_size =0.2)
# instantiate the imbalance bagging classifier, training,
prediction
cls = SMOTEBag(n_estimator = 50, estimator =
DecisionTreeClassifier())
clf.fit(X_train , y_train)
y_pred = clf.predict(X_test)
```

#### **Methods**

fit(self, X_train, y_train)	Fit the model.
predict(self, X)	Predict the class label for sample X
predict_proba(self, X)	Estimate the probability of X belonging to each class-labels.

## fit(self, X\_train,y\_train)

Parameters X train: numpy.ndarray of shape (n samples, n features)

The features to train the model.

y\_train: numpy.ndarray of shape (n\_samples,)

An array-like with the class labels of all samples in X train.

**Returns: self** 

## predict(self, X):

Parameters X: numpy.ndarray of shape (n samples, n features)

All the samples we want to predict the label for.

Returns: numpy.ndarray

A 1D array of shape (, n\_samples), containing the predicted class labels for all instances in X.

## predict proba(self, X):

Parameters X: numpy.ndarray of shape (n samples, n features)

All the samples we want to predict the label for.

Returns: numpy.ndarray

A 2D array of shape (n\_samples, n\_classes). Where each i-th row contains len(self.target\_value) elements, representing the probability that the i-th sample of X belongs to a certain class label.