### **Bagging of Extrapolation-SMOTE SVM**

The Bagging of Extrapolation-SMOTE SVM (BEBS) algorithm integrates borderline information into the SMOTE algorithm. This algorithm capitalizes on the fact that samples near the decision boundary contain more discriminative information. The skew of the boundary is rectified by constructing synthetic samples using these samples.

**Source**: Wang, Q., Luo, Z., Huang, J., Feng, Y., & Liu, Z. (2017). A Novel Ensemble Method for Imbalanced Data Learning: Bagging of Extrapolation-SMOTE SVM. Computational intelligence and neuroscience, 2017(1), 1827016.

Parameters: n estimator: int (default=10)

The number of nearest neighbors to search for.

# **Examples:**

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

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 = BEBS(n_estimator = 50)
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

# 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.