



# Imbalanced-Learn module in Python

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Imbalanced-Learn is a Python module that helps in balancing the datasets which are highly skewed or biased towards some classes. Thus, it helps in resampling the classes which are otherwise oversampled or undersampled. If there is a greater imbalance ratio, the output is biased to the class which has a higher number of examples. The following dependencies need to be installed to use imbalanced-learn:

- `scipy(>=0.19.1)`
- `numpy(>=1.13.3)`
- `scikit-learn(>=0.23)`
- `joblib(>=0.11)`
- `keras 2` (optional)
- `tensorflow` (optional)

To install imbalanced-learn just type in :

```
pip install imbalanced-learn
```

The resampling of data is done in 2 parts:

**Estimator:** It implements a fit method which is derived from *scikit-learn*. The data and targets are both in the form of a 2D array



```
estimator = obj.fit(data, targets)
```

**Resampler:** The *fit\_resample* method resample the data and targets into a dictionary with a key-value pair of *data\_resampled* and *targets\_resampled*.

```
data_resampled, targets_resampled = obj.fit_resample(data,
targets)
```

The Imbalanced Learn module has different algorithms for oversampling and undersampling:

We will use the built-in dataset called the *make\_classification* dataset which return

- **x**: a matrix of n\_samples\*n\_features and
- **y**: an array of integer labels.

Click [dataset](#) to get the dataset used.

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## Python3

```
# import required modules
from sklearn.datasets import make_classification

# define dataset
x, y = make_classification(n_samples=10000,
                           weights=[0.99],
                           flip_y=0)

print('x:\n', x)
print('y:\n', y)
```

### Output:

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```

x:
[[ 0.71276923 -0.98326057 -0.87568931 ... -1.99993483  0.9751252
 -0.90000921]
 [-1.65450542 -0.88195312  0.95582941 ...  0.07119132  1.30072855
 -0.3771063 ]
 [ 2.30941943  0.65107945 -0.2355681 ...  0.09448597 -0.74031559
 0.61787822]
 ...
 [ 0.62834832 -0.39341776  1.4454335 ... -0.54503017 -0.30142052
 -1.07822178]
 [ 0.97680739  0.15958695  0.5623564 ...  0.5974245 -0.15265927
 -1.66262164]
 [ 0.02772241 -1.45204323  0.55036507 ...  0.40683996 -0.69263495
 0.64249002]]
y:
[0 0 0 ... 0 0 0]

```

Below are some programs in which depict how to apply oversampling and undersampling to the dataset:

## Oversampling

- **Random Over Sampler:** It is a naive method where classes that have low examples are generated and randomly resampled.

## Syntax:

```
from imblearn.over_sampling import RandomOverSampler
```

**Parameters(optional):** *sampling\_strategy='auto', return\_indices=False, random\_state=None, ratio=None*

### **Implementation:**

```
oversample = RandomOverSampler(sampling_strategy='minority')
X_oversample,Y_oversample=oversample.fit_resample(X,Y)
```

**Return Type:***a matrix with the shape of n\_samples\*n\_features*

## Example:

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```
# import required modules
from sklearn.datasets import make_classification
from imblearn.over_sampling import RandomOverSampler

# define dataset
x, y = make_classification(n_samples=10000,
                          weights=[0.99],
                          flip_y=0)

oversample = RandomOverSampler(sampling_strategy='minority')
x_over, y_over = oversample.fit_resample(x, y)

# print the features and the labels
print('x_over:\n', x_over)
print('y_over:\n', y_over)
```

### Output:

```
x_over:
[[-0.47817272 -0.62714208  0.16687505 ... -0.89802198  0.05962375
  0.62797162]
 [ 1.92254368 -0.42039287 -0.92273141 ...  0.64676004 -0.31657257
 -3.59310973]
 [ 0.71079186 -0.05674072  0.0524251  ... -0.13068423 -0.12462373
 -0.3223906 ]
 ...
 [ 1.93014401  0.06327391  0.81191734 ...  0.59143051 -0.00555849
  0.63403685]
 [ 0.55141876 -0.69346537 -0.34922318 ... -0.17817397  0.32749715
  1.45827921]
 [-0.73560282  1.62320154  0.38547395 ... -1.35220893  0.46219923
 -2.04995979]]
y_over:
[0 0 0 ... 1 1 1]
```

- **SMOTE, ADASYN:** Synthetic Minority Oversampling Technique (SMOTE) and the Adaptive Synthetic (ADASYN) are 2 methods used in oversampling. These also generate low examples but ADASYN takes into account the density of distribution to distribute the data points evenly.

### Syntax:

```
from imblearn.over_sampling import SMOTE, ADASYN
```

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*Parameters(optional):\*, sampling\_strategy='auto', random\_state=None, n\_neighbors=5, n\_jobs=None*

**Implementation:**

```
smote = SMOTE(ratio='minority')
X_smote,Y_smote=smote.fit_resample(X,Y)
```

*Return Type:a matrix with the shape of n\_samples\*n\_features*

**Example:**

---

## Python3

```
# import required modules
from sklearn.datasets import make_classification
from imblearn.over_sampling import SMOTE

# define dataset
x, y = make_classification(n_samples=10000, weights=[0.99], flip_y=0)
smote = SMOTE()
x_smote, y_smote = smote.fit_resample(x, y)

# print the features and the labels
print('x_smote:\n', x_smote)
print('y_smote:\n', y_smote)
```

**Output:**

```

x_smote:
[[ 0.07757261  0.55707706 -1.28435379 ...  0.79248002  1.79293049
  0.64704489]
 [-1.62850262 -1.49229817  0.92872149 ...  1.77990205  1.87848699
  0.83031737]
 [ 0.53977421 -0.32822633 -0.90911299 ...  0.30618373  1.15298085
  1.30019422]
 ...
 [ 0.79362245  0.07473455  1.69905552 ...  0.31442711 -1.36481352
 -1.06840418]
 [ 0.72376507  1.31578618 -0.82658488 ... -0.52831636  1.13885669
 -0.41189875]
 [-1.05447793 -0.77821703 -0.10061748 ... -0.41375807 -0.68706483
 -1.38604644]]
y_smote:
[0 0 0 ... 1 1 1]

```

## Undersampling

- **Edited Nearest Neighbours:** This algorithm removes any sample which has labels different from those of its adjoining classes.

### Syntax:

*from imblearn.under\_sampling import EditedNearestNeighbours*

**Parameters(optional):** *sampling\_strategy='auto', return\_indices=False, random\_state=None, n\_neighbors=3, kind\_sel='all', n\_jobs=1, ratio=None*

### Implementation:

*en = EditedNearestNeighbours()*

*X\_en, Y\_en=en.fit\_resample(X, y)*

**Return Type:***a matrix with the shape of n\_samples\*n\_features*

### Example:

## Python3

```
# import required modules
```

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```
# define dataset
x, y = make_classification(n_samples=10000, weights=[0.99], flip_y=0)
en = EditedNearestNeighbours()
x_en, y_en = en.fit_resample(x, y)

# print the features and the labels
print('x_en:\n', x_en)
print('y_en:\n', y_en)
```

## Output:

```
x_en:
[[ 0.22785828 -0.85084888  1.38166738 ... -1.68751472  0.75997958
 -0.90306857]
 [-1.05263874 -1.88084168 -1.14235759 ...  0.74286784  1.21184758
  0.06549804]
 [-0.61894734  0.63828964  1.16209772 ... -0.57984432 -0.52139758
  1.71656638]
 ...
 [ 0.7000294  -0.63800371  0.00634002 ... -0.36264562  0.02097873
 -1.19811301]
 [-0.77416036 -0.94446611  0.91688717 ... -0.83353398  0.58838661
  1.2912737 ]
 [-1.35477743  0.23246413  0.57379621 ... -0.6267911  0.95713282
 -0.91976794]]
y_en:
[0 0 0 ... 1 1 1]
```

- **Random Under Sampler:** It involves sampling any random class with or without any replacement.

## Syntax:

*from imblearn.under\_sampling import RandomUnderSampler*

**Parameters(optional):** *sampling\_strategy='auto', return\_indices=False, random\_state=None, replacement=False, ratio=None*

### **Implementation:**

*undersample = RandomUnderSampler()*

*X\_under, y\_under = undersample.fit\_resample(X, y)*

**Return Type:** *a matrix with the shape of n\_samples\*n\_features*

Example:

# Python3

```
# import required modules

from sklearn.datasets import make_classification
from imblearn.under_sampling import RandomUnderSampler

# define dataset

x, y = make_classification(n_samples=10000,
                           weights=[0.99],
                           flip_y=0)

undersample = RandomUnderSampler()

x_under, y_under = undersample.fit_resample(x, y)

# print the features and the labels

print('x_under:\n', x_under)
print('y_under:\n', y_under)
```



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