## **Combination Sampling**

Implement the SMOTEENN technique with the credit card default data. Then estimate a logistic regression model and report the classification evaluation metrics.

In\_balance\_limit is the log of the maximum balance they can have on the card; 1 is female, 0 male for sex; the education is denoted: 1 = graduate school; 2 = university; 3 = high school; 4 = others; 1 is married and 0 single for marriage; default\_next\_month is whether the person defaults in the following month (1 yes, 0 no).

```
In [1]: import pandas as pd
         from path import Path
         from collections import Counter
 In [7]: data = Path('../Resources/cc default.csv')
         df = pd.read csv(data)
         df.head()
 Out[7]:
             ID In_balance_limit sex education marriage age default_next_month
             1
                     9.903488
                              1
                                       2
                                               0
                                                   24
                                                                   1
          1
             2
                    11.695247
                                       2
                                               1
                                                   26
                                                                   1
                    11.407565
                                       2
                                                   34
                                                                   0
                    10.819778
                                       2
                                                                   0
                                               0
                                                   37
            5
                    10.819778
 In [8]: x cols = [i for i in df.columns if i not in ('ID', 'default next month')]
         X = df[x cols]
         y = df['default next month']
 In [9]: x cols
 Out[9]: ['In balance limit', 'sex', 'education', 'marriage', 'age']
In [10]: Counter(y)
Out[10]: Counter({1: 6636, 0: 23364})
In [11]: # Normal train-test split
         from sklearn.model_selection import train_test_split
         X_train, X_test, y_train, y_test = train_test_split(X, y, random_state=1)
```

## **Combination Sampling with SMOTEENN**

```
In [12]: # Use the SMOTEENN technique to perform combination sampling on the data
# Count the resampled classes
from imblearn.combine import SMOTEENN

smote_enn = SMOTEENN(random_state=0)
X_resampled, y_resampled = smote_enn.fit_resample(X, y)
Counter(y_resampled)
Out[12]: Counter({0: 10148, 1: 7645})
```

## **Logistic Regression**

```
In [13]: # Fit a Logistic regression model using random undersampled data
    from sklearn.linear_model import LogisticRegression
    model = LogisticRegression(solver='lbfgs', random_state=1)
    model.fit(X_resampled, y_resampled)
Out[13]: LogisticRegression(random_state=1)
```

## **Evaluation Metrics**

Out[15]: 0.5944929241791752

In [16]: # Print the imbalanced classification report
from imblearn.metrics import classification\_report\_imbalanced

print(classification\_report\_imbalanced(y\_test, y\_pred))

		pre	rec	spe	f1	geo	iba
sup							
5832 1668	0	0.83	0.72	0.47	0.77	0.58	0.35
	1	0.32	0.47	0.72	0.38	0.58	0.33
avg / total 7500		0.71	0.66	0.53	0.68	0.58	0.34

In [ ]: