other models using Python

May 5, 2021

We found the following methods on this website: https://machinelearningmastery.com/bagging-and-random-forest-for-imbalanced-classification/

```
[1]: import numpy as np
     import pandas as pd
     from numpy import mean
     import matplotlib.pyplot as plt
     import seaborn as sns
     from sklearn.metrics import confusion_matrix
     from sklearn.datasets import make_classification
     from sklearn.model_selection import cross_val_score
     from sklearn.model_selection import RepeatedStratifiedKFold
     from sklearn.metrics import f1_score
     from sklearn.metrics import matthews_corrcoef
[2]: train = pd.read_csv("traindata.csv").iloc[:,1:]
     test = pd.read_csv("testdata.csv").iloc[:,1:]
[3]: train1 = pd.get_dummies(train,drop_first=True,columns=['Sex','Race',u
     →'Insurance', 'SEER.Registry', 'Subsite', 'AJCC.7.Stage', 'Chemotherapy', □

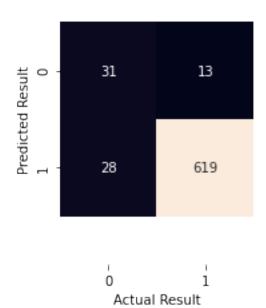
¬'follow'])
[4]: test1 = pd.get_dummies(test,drop_first=True,columns=['Sex','Race', 'Insurance',_
     →'SEER.Registry', 'Subsite', 'AJCC.7.Stage', 'Chemotherapy', 'follow'])
[5]: X = train1.iloc[:,:-1]
     y = train1['follow_1']
[6]: test_X = test1.iloc[:,:-1]
     test_y = test1['follow_1']
[7]: cv = RepeatedStratifiedKFold(n_splits=10, n_repeats=3, random_state=1)
[8]: def getResult(mod):
         fit = mod.fit(X,y)
         pred = mod.predict(test_X)
         scores = cross_val_score(mod, X, y, scoring='roc_auc', cv=cv, n_jobs=-1)
         print('Mean ROC AUC: %.3f' %mean(scores))
```

```
print('F1-score: %.3f' %f1_score(test_y, pred))
print('MCC: %.3f' %matthews_corrcoef(test_y, pred))
mat = confusion_matrix(pred, test_y)
mat
sns.heatmap(mat, square=True, annot=True, cbar=False, fmt='g')
plt.xlabel("Actual Result")
plt.ylabel("Predicted Result")
a, b = plt.ylim()
a += 0.5
b -= 0.5
plt.ylim(a, b)
plt.show()
```

1 Bagging for Imbalanced Classification

1.0.1 Standard Bagging

Mean ROC AUC: 0.800 F1-score: 0.968 MCC: 0.578



```
CPU times: user 241 ms, sys: 86.3 ms, total: 327 ms
```

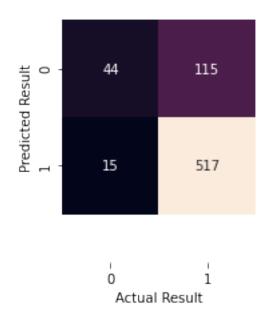
Wall time: 4.69 s

1.0.2 Bagging With Random Undersampling

[11]: from imblearn.ensemble import BalancedBaggingClassifier

[12]: %%time getResult(BalancedBaggingClassifier())

Mean ROC AUC: 0.834 F1-score: 0.888 MCC: 0.374



CPU times: user 169 ms, sys: 17.3 ms, total: 186 ms

Wall time: 801 ms

2 Random Forest for Imbalanced Classification

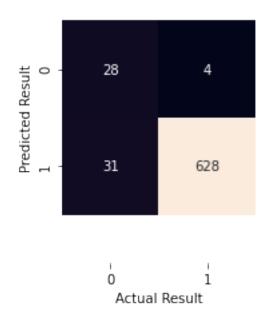
2.0.1 Standard Random Forest

[13]: from sklearn.ensemble import RandomForestClassifier

[14]: %time getResult(RandomForestClassifier(n_estimators=10))

Mean ROC AUC: 0.802 F1-score: 0.973

MCC: 0.623



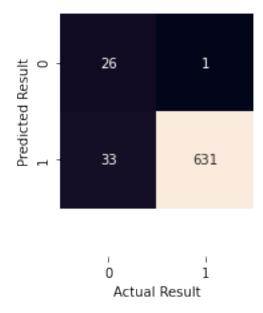
CPU times: user 143 ms, sys: 14.9 ms, total: 158 ms

Wall time: 470 ms

2.0.2 Random Forest With Class Weighting

[15]: %%time getResult(RandomForestClassifier(n_estimators=10, class_weight='balanced'))

Mean ROC AUC: 0.787 F1-score: 0.974



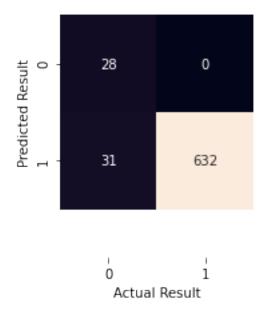
CPU times: user 211 ms, sys: 21.3 ms, total: 232 ms

Wall time: 517 ms

2.0.3 Random Forest With Bootstrap Class Weighting

[16]: \[\%\time \] getResult(RandomForestClassifier(n_estimators=10, \(\time \) \(\time \) class_weight='balanced_subsample'))

Mean ROC AUC: 0.789 F1-score: 0.976



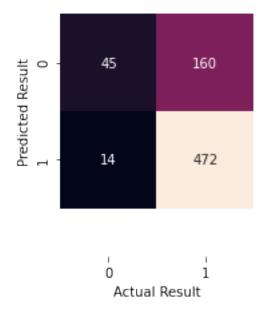
CPU times: user 152 ms, sys: 14.2 ms, total: 166 ms

Wall time: 526 ms

2.0.4 Random Forest With Random Undersampling

[17]: from imblearn.ensemble import BalancedRandomForestClassifier

Mean ROC AUC: 0.809 F1-score: 0.844



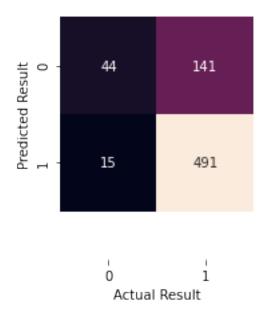
CPU times: user 149 ms, sys: 14 ms, total: 163 ms

Wall time: 685 ms

3 Easy Ensemble for Imbalanced Classification

3.0.1 Easy Ensemble

Mean ROC AUC: 0.834 F1-score: 0.863



CPU times: user 886 ms, sys: 26.7 ms, total: 913 ms

Wall time: 10.1 s