Apply+SVM+to+Amazon+reviews+data+set+bow+tfidf

June 1, 2018

In [1]: !pip install imblearn

Collecting imblearn

!pip install gensim

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```
Collecting imbalanced-learn (from imblearn)
 Downloading https://files.pythonhosted.org/packages/80/a4/900463a3c0af082aed9c5a43f4ec317a9469
    100% || 153kB 3.2MB/s ta 0:00:01
Requirement already satisfied: numpy in /usr/local/lib/python3.6/site-packages (from imbalanced-
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.6/site-packages (from imba
Requirement already satisfied: scipy in /usr/local/lib/python3.6/site-packages (from imbalanced-
Installing collected packages: imbalanced-learn, imblearn
Successfully installed imbalanced-learn-0.3.3 imblearn-0.0
You are using pip version 9.0.1, however version 10.0.1 is available. You should consider upgradi
Collecting gensim
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Requirement already satisfied: six>=1.5.0 in /usr/local/lib/python3.6/site-packages (from gensim
Requirement already satisfied: numpy>=1.11.3 in /usr/local/lib/python3.6/site-packages (from ger
Collecting smart-open>=1.2.1 (from gensim)
  Downloading https://files.pythonhosted.org/packages/4b/69/c92661a333f733510628f28b8282698b62cd
Requirement already satisfied: scipy>=0.18.1 in /usr/local/lib/python3.6/site-packages (from ger
Collecting boto>=2.32 (from smart-open>=1.2.1->gensim)
  Downloading https://files.pythonhosted.org/packages/bd/b7/a88a67002b1185ed9a8e8a6ef15266728c23
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Collecting bz2file (from smart-open>=1.2.1->gensim)
  Downloading https://files.pythonhosted.org/packages/61/39/122222b5e85cd41c391b68a99ee296584b2a
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Collecting boto3 (from smart-open>=1.2.1->gensim)
  Downloading https://files.pythonhosted.org/packages/25/54/288f3d87d055440e20eb7a9dce58fdd005c9
    100% || 133kB 9.3MB/s eta 0:00:01
Collecting s3transfer<0.2.0,>=0.1.10 (from boto3->smart-open>=1.2.1->gensim)
```

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Downloading https://files.pythonhosted.org/packages/da/aa/b227500e26dbbd95bd6cda78cf784f769bba

Downloading https://files.pythonhosted.org/packages/81/a7/4179e6ebfd654bd0eac0b9c06125b8b4c96a

Collecting botocore<1.11.0,>=1.10.30 (from boto3->smart-open>=1.2.1->gensim)

```
Downloading https://files.pythonhosted.org/packages/b7/31/05c8d001f7f87f0f07289a5fc0fc3832e9a5
Collecting docutils>=0.10 (from botocore<1.11.0,>=1.10.30->boto3->smart-open>=1.2.1->gensim)
  Downloading https://files.pythonhosted.org/packages/36/fa/08e9e6e0e3cbd1d362c3bbee8d01d0aedb21
    100% || 552kB 2.6MB/s eta 0:00:01
Requirement already satisfied: python-dateutil<3.0.0,>=2.1; python_version >= "2.7" in /usr/loca
Building wheels for collected packages: smart-open, bz2file
  Running setup.py bdist_wheel for smart-open ... done
  Stored in directory: /root/.cache/pip/wheels/b1/9e/7d/bb3d3b55c597e72617140a0638c06382a5f17283
  Running setup.py bdist_wheel for bz2file ... done
  Stored in directory: /root/.cache/pip/wheels/81/75/d6/e1317bf09bf1af5a30befc2a007869fa6e1f516b
Successfully built smart-open bz2file
Installing collected packages: boto, bz2file, jmespath, docutils, botocore, s3transfer, boto3, s
Successfully installed boto-2.48.0 boto3-1.7.30 botocore-1.10.30 bz2file-0.98 docutils-0.14 gens
You are using pip version 9.0.1, however version 10.0.1 is available. You should consider upgradi
In [1]: from sklearn.model_selection import train_test_split
        from sklearn.grid_search import GridSearchCV
        from sklearn.grid_search import RandomizedSearchCV
        from scipy.stats import uniform
        from scipy.stats import norm
        from imblearn.over_sampling import SMOTE
        import sqlite3
        import pandas as pd
        import numpy as np
        from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.feature_extraction.text import CountVectorizer
        import gensim
        from sklearn.metrics import classification_report, accuracy_score, confusion_matrix
        from sklearn.preprocessing import StandardScaler
/usr/local/lib/python3.6/site-packages/sklearn/cross_validation.py:41: DeprecationWarning: This
  "This module will be removed in 0.20.", DeprecationWarning)
/usr/local/lib/python3.6/site-packages/sklearn/grid_search.py:42: DeprecationWarning: This modul
  DeprecationWarning)
In [2]: con = sqlite3.connect('final.sqlite') # this is cleaned dataset
        final = pd.read_sql_query("""
        SELECT Score, Text_not_included
```

Collecting jmespath<1.0.0,>=0.7.1 (from boto3->smart-open>=1.2.1->gensim)

```
FROM reviews
        """, con)[:2000]
In [3]: for i, seq in enumerate(final['Text_not_included']):
          final['Text_not_included'][i]=final['Text_not_included'][i].decode('UTF-8')
        X_train, X_test, y_train , y_test = train_test_split(final['Text_not_included'], final['
0.1 Generate Count BoW vectors
In [4]: count_vect = CountVectorizer(ngram_range=(1,2) )
        count_vect.fit(X_train)
        bow_train=count_vect.transform(X_train)
        bow_test=count_vect.transform(X_test)
0.2 Generate TF IDF vectors
In [5]: tf_idf_vect=TfidfVectorizer(ngram_range=(1,2), min_df=10, dtype=float)
       tf_idf_vect.fit(X_train)
        tf_idf_train=tf_idf_vect.transform(X_train)
        tf_idf_test=tf_idf_vect.transform(X_test)
/usr/local/lib/python3.6/site-packages/sklearn/feature_extraction/text.py:1089: FutureWarning: (
  if hasattr(X, 'dtype') and np.issubdtype(X.dtype, np.float):
0.3 Upsampling followed by standardization
In [6]: # Upsampling minority class
        over_sampler = SMOTE(ratio='minority')
        bow_train_resampled, y_train_resampled = over_sampler.fit_sample(bow_train, y_train)
        tf_idf_train_resampled, y_train_resampled = over_sampler.fit_sample(tf_idf_train, y_trai
        scaler_bow=StandardScaler(with_mean=False)
        scaler_tf_idf=StandardScaler(with_mean=False)
        scaler_bow.fit(bow_train_resampled)
        scaler_tf_idf.fit(tf_idf_train_resampled)
        bow_train_scaled=scaler_bow.transform(bow_train_resampled)
        tf_idf_train_scaled=scaler_tf_idf.transform(tf_idf_train_resampled)
        bow_test_scaled=scaler_bow.transform(bow_test)
        tf_idf_test_scaled=scaler_tf_idf.transform(tf_idf_test)
/usr/local/lib/python3.6/site-packages/sklearn/utils/validation.py:444: DataConversionWarning: I
  warnings.warn(msg, DataConversionWarning)
```

In [7]: from sklearn.svm import SVC

0.4 Classification using count Bow

```
In [16]: tuned_parameters = {'C': np.linspace(10.0, 20, 10, dtype=float), 'gamma': np.linspace(
                 #Using GridSearchCV
                 gscv = GridSearchCV(SVC(), tuned_parameters, scoring = 'accuracy', cv=5)
                 print(gscv.fit(bow_train_scaled, y_train_resampled))
                 tuned_parameters = {'C' : uniform(10,20), 'gamma' : uniform(0,1)}
                 #Using RandomizedSearchCV
                 rscv = RandomizedSearchCV(SVC(), tuned_parameters, scoring = 'accuracy', cv=5, n_iter=2
                 print(rscv.fit(bow_train_scaled, y_train_resampled))
GridSearchCV(cv=5, error_score='raise',
              estimator=SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
   decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max_iter=-1, probability=False, random_state=None, shrinking=True,
   tol=0.001, verbose=False),
             fit_params={}, iid=True, n_jobs=1,
             param_grid={'C': array([10.
                                                                        , 11.11111, 12.22222, 13.33333, 14.44444, 15.55556,
             16.66667, 17.77778, 18.88889, 20. ]), 'gamma': array([0.001, 0.112, 0.223, 0.334, 0.4
             pre_dispatch='2*n_jobs', refit=True, scoring='accuracy', verbose=0)
RandomizedSearchCV(cv=5, error_score='raise',
                   estimator=SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
   decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max_iter=-1, probability=False, random_state=None, shrinking=True,
   tol=0.001, verbose=False),
                   fit_params={}, iid=True, n_iter=20, n_jobs=1,
                   param_distributions={'C': <scipy.stats._distn_infrastructure.rv_frozen object at 0x7f5
                   pre_dispatch='2*n_jobs', random_state=None, refit=True,
                   scoring='accuracy', verbose=0)
In [17]: predictions = gscv.best_estimator_.predict(bow_test_scaled)
                 print(classification_report(y_test, predictions))
                 print(confusion_matrix(y_test, predictions).T)
                 tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
                 print("TPR = {}\n TNR = {}\n FPR = {}\n FNR = {}\".format(tp/(fn+tp), tn/(tn+fp), fp/(tn+fp), fp/(tn
                 predictions = rscv.best_estimator_.predict(bow_test_scaled)
                 print(classification_report(y_test, predictions))
                 print(confusion_matrix(y_test, predictions).T)
                 tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
```

```
 print("TPR = {} \setminus n \ TNR = {} \setminus n \ FPR = {} \setminus n \ FNR = {} \mid .format(tp/(fn+tp), \ tn/(tn+fp), \ fp/(tn+fp), 
                                    precision
                                                                        recall f1-score
                                                                                                                             support
                                                                              0.01
                                                                                                          0.02
                                                                                                                                           87
        negative
                                                  0.11
                                                                                                          0.87
        positive
                                                  0.78
                                                                              0.97
                                                                                                                                         313
avg / total
                                                  0.63
                                                                             0.77
                                                                                                          0.68
                                                                                                                                         400
ΓΓ 1
                     87
  [ 86 305]]
TPR = 0.9744408945686901
  TNR = 0.011494252873563218
  FPR = 0.9885057471264368
  FNR = 0.025559105431309903
                                    precision
                                                                        recall f1-score
                                                                                                                              support
        negative
                                                  0.00
                                                                              0.00
                                                                                                          0.00
                                                                                                                                           87
        positive
                                                  0.78
                                                                              1.00
                                                                                                          0.88
                                                                                                                                         313
avg / total
                                                  0.61
                                                                             0.78
                                                                                                          0.69
                                                                                                                                         400
[[ 0 0]]
  [ 87 313]]
TPR = 1.0
  TNR = 0.0
  FPR = 1.0
  FNR = 0.0
/usr/local/lib/python3.6/site-packages/sklearn/metrics/classification.py:1135: UndefinedMetricWa
      'precision', 'predicted', average, warn_for)
In [18]: print(gscv.best_estimator_)
SVC(C=10.0, cache_size=200, class_weight=None, coef0=0.0,
     decision_function_shape='ovr', degree=3, gamma=0.001, kernel='rbf',
     max_iter=-1, probability=False, random_state=None, shrinking=True,
     tol=0.001, verbose=False)
In [19]: print(rscv.best_estimator_)
SVC(C=11.996771023901712, cache_size=200, class_weight=None, coef0=0.0,
     decision_function_shape='ovr', degree=3, gamma=0.14509256003855375,
     kernel='rbf', max_iter=-1, probability=False, random_state=None,
```

shrinking=True, tol=0.001, verbose=False)

0.5 Classification using TF IDF

```
In [20]: tuned_parameters = {'C': np.linspace(10.0, 20, 20, dtype=float), 'gamma': np.linspace(
                #Using GridSearchCV
                gscv = GridSearchCV(SVC(), tuned_parameters, scoring = 'accuracy', cv=5)
                print(gscv.fit(tf_idf_train_scaled, y_train_resampled))
                tuned_parameters = {'C' : uniform(10,20), 'gamma' : uniform(0,1)}
                #Using RandomizedSearchCV
                rscv = RandomizedSearchCV(SVC(), tuned_parameters, scoring = 'accuracy', cv=5, n_iter=1
                print(rscv.fit(tf_idf_train_scaled, y_train_resampled))
GridSearchCV(cv=5, error_score='raise',
             estimator=SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
   decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max_iter=-1, probability=False, random_state=None, shrinking=True,
   tol=0.001, verbose=False),
            fit_params={}, iid=True, n_jobs=1,
            13.15789, 13.68421, 14.21053, 14.73684, 15.26316, 15.78947,
             16.31579, 16.84211, 17.36842, 17.89474, 18.42105, 18.94737,
            19.47368, 20.
                                          ]), 'gamma': array([0.001, 0.112, 0.223, 0.334, 0.445, 0.556, 0.667, 0.
             1. ])},
            pre_dispatch='2*n_jobs', refit=True, scoring='accuracy', verbose=0)
RandomizedSearchCV(cv=5, error_score='raise',
                  estimator=SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
   decision_function_shape='ovr', degree=3, gamma='auto', kernel='rbf',
   max_iter=-1, probability=False, random_state=None, shrinking=True,
   tol=0.001, verbose=False),
                  fit_params={}, iid=True, n_iter=15, n_jobs=1,
                  param_distributions={'C': <scipy.stats._distn_infrastructure.rv_frozen object at 0x7f5
                  pre_dispatch='2*n_jobs', random_state=None, refit=True,
                  scoring='accuracy', verbose=0)
In [21]: predictions = gscv.best_estimator_.predict(tf_idf_test_scaled)
                print(classification_report(y_test, predictions))
                print(confusion_matrix(y_test, predictions).T)
                tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
                print("TPR = {}\n TNR = {}\n FPR = {}\n FNR = {}\".format(tp/(fn+tp), tn/(tn+fp), fp/(tn+fp), fp/(tn
                predictions = rscv.best_estimator_.predict(tf_idf_test_scaled)
                print(classification_report(y_test, predictions))
                print(confusion_matrix(y_test, predictions).T)
                tn, fp, fn, tp = confusion_matrix(y_test, predictions).ravel()
```

```
 print("TPR = {} \setminus n \ TNR = {} \setminus n \ FPR = {} \setminus n \ FNR = {} \mid .format(tp/(fn+tp), \ tn/(tn+fp), \ fp/(tn+fp), 
                                                                 precision
                                                                                                                                   recall f1-score
                                                                                                                                                                                                                                    support
              negative
                                                                                          0.60
                                                                                                                                             0.03
                                                                                                                                                                                                0.07
                                                                                                                                                                                                                                                            87
              positive
                                                                                          0.79
                                                                                                                                             0.99
                                                                                                                                                                                                0.88
                                                                                                                                                                                                                                                        313
                                                                                                                                                                                                0.70
                                                                                                                                                                                                                                                        400
avg / total
                                                                                          0.75
                                                                                                                                            0.79
[[ 3
                                       2]
    [ 84 311]]
TPR = 0.9936102236421726
    TNR = 0.034482758620689655
    FPR = 0.9655172413793104
     FNR = 0.006389776357827476
                                                                  precision
                                                                                                                                   recall f1-score
                                                                                                                                                                                                                                    support
              negative
                                                                                          0.00
                                                                                                                                             0.00
                                                                                                                                                                                                0.00
                                                                                                                                                                                                                                                            87
              positive
                                                                                          0.78
                                                                                                                                              1.00
                                                                                                                                                                                                0.88
                                                                                                                                                                                                                                                        313
                                                                                          0.61
                                                                                                                                            0.78
                                                                                                                                                                                                0.69
                                                                                                                                                                                                                                                        400
avg / total
[[ 0 0]
    [ 87 313]]
TPR = 1.0
   TNR = 0.0
    FPR = 1.0
    FNR = 0.0
```

In [22]: gscv.best_estimator_

/usr/local/lib/python3.6/site-packages/sklearn/metrics/classification.py:1135: UndefinedMetricWaterics of the control of the c

shrinking=True, tol=0.001, verbose=False)

0.6 Conclusions

Both BoW and TFIDF although provide excellent TPR, are failing at TNR. Also given large amount of time taken to train, SVMs combined with such high dimensional representations are not a good choice for text classification. Somewhat decent results are given by gamma: $0.001\ 10 < C < 20$