```
In [4]: import pandas as pd
        import numpy as np
        from sklearn.feature extraction.text import CountVectorizer, TfidfVector
        from sklearn import model selection, naive bayes, svm
        from sklearn.svm import SVC
        from sklearn.model selection import GridSearchCV
        from sklearn.metrics import accuracy score, f1 score, roc auc score
        from tqdm import tqdm
        from sklearn.model_selection import train_test_split
        from sklearn.preprocessing import MaxAbsScaler
        import os.path
        import pickle
In [5]: X train = pd.read pickle('../../Preprocessing/Data/X train.pkl')
        X_test = pd.read_pickle('../../Preprocessing/Data/X_test.pkl')
        y_train = pd.read_pickle('../../Preprocessing/Data/y_train.pkl')
        y test = pd.read pickle('../../Preprocessing/Data/y test.pkl')
In [6]: vec = TfidfVectorizer(ngram range=(1, 2), min df=3,
                              max df=0.9, strip accents='unicode', use idf=1, sm
        ooth idf=1, sublinear tf=1)
        df = pd.read pickle('../../Preprocessing/Data/preprocess pickle.pkl')
        df['comment text final'] = [" ".join(text) for text in df['comment text
        final'].values]
        vec = vec.fit(df['comment text final'])
In [7]: train_term_doc = vec.transform(X_train)
        test term doc = vec.transform(X test)
In [8]: | scaler = MaxAbsScaler()
        train term doc = scaler.fit transform(train term doc)
        test term doc = scaler.fit transform(test term doc)
In [6]: if os.path.isfile('Models/svm poly.sav'):
            clf = pickle.load(open('Models/svm poly.sav', 'rb'))
        else:
            clf = svm.SVC(kernel = 'poly', gamma='scale', degree=1, verbose=1)
            clf.fit(train_term_doc, y_train['toxic'])
            pickle.dump(clf, open('Models/svm poly.sav', 'wb'))
In [7]: | svm pred = clf.predict(test term doc)
```

```
In [8]: print('Accuracy SVM Linear Kernel:', accuracy_score(y_test['toxic'], svm
         print('F1 Score SVM Linear Kernel:', f1_score(y_test['toxic'], svm_pred
         print('ROC-AUC Score SVM Linear Kernel:', roc auc score(y test['toxic'],
         svm pred))
         Accuracy SVM Linear Kernel: 0.9064041914644692
         F1 Score SVM Linear Kernel: 0.7991704867932765
         ROC-AUC Score SVM Linear Kernel: 0.8492133295324785
In [9]:
        if os.path.isfile('Models/svm rbf.sav'):
             clf rbf = pickle.load(open('Models/svm rbf.sav', 'rb'))
         else:
             clf_rbf = svm.SVC(kernel = 'rbf', gamma='scale', verbose=1)
             clf rbf.fit(train_term_doc, y_train['toxic'])
             pickle.dump(clf, open('Models/svm rbf.sav', 'wb'))
         [LibSVM]
         svm rbf pred = clf rbf.predict(test term doc)
In [10]:
         print('Accuracy SVM RBF Kernel:', accuracy_score(y_test['toxic'], svm rb
In [11]:
         print('F1 Score SVM RBG Kernel:', f1 score(y test['toxic'], svm_rbf_pred
         print('ROC-AUC Score SVM RBF Kernel:', roc auc score(y test['toxic'], sv
         m rbf pred))
         Accuracy SVM RBF Kernel: 0.8759346864031741
         F1 Score SVM RBG Kernel: 0.6976571216065452
         ROC-AUC Score SVM RBF Kernel: 0.7728846276718616
 In [3]: n folds = 5
         c_{vals} = np.power(float(10), range(-7, 7 + 1))
         degree of poly kernel = [1, 2, 3]
         param grid = {'C': c vals, 'degree' : degree of poly kernel}
         grid search = GridSearchCV(svm.SVC(kernel='poly', gamma='scale'), param
         grid, cv=n folds, iid=False, n jobs = -1, verbose=10)
```

```
In [17]: if os.path.isfile('Models/svm poly cv.sav'):
             grid search = pickle.load(open('Models/svm poly cv.sav', 'rb'))
         else:
             grid_search.fit(train_term_doc, y_train['toxic'])
             pickle.dump(grid search, open('Models/svm poly cv.sav', 'wb'))
         Fitting 5 folds for each of 33 candidates, totalling 165 fits
         [Parallel(n jobs=-1)]: Using backend LokyBackend with 12 concurrent wor
         kers.
                                                      elapsed:
         [Parallel(n jobs=-1)]: Done
                                       1 tasks
                                                                 3.5min
         [Parallel(n jobs=-1)]: Done
                                                       elapsed:
                                                                 3.6min
                                       8 tasks
         [Parallel(n_jobs=-1)]: Done 17 tasks
                                                       elapsed:
                                                                 6.9min
         [Parallel(n jobs=-1)]: Done
                                      26 tasks
                                                      elapsed: 10.6min
         [Parallel(n_jobs=-1)]: Done
                                      37 tasks
                                                       elapsed: 18.4min
         [Parallel(n_jobs=-1)]: Done 48 tasks
                                                       elapsed: 26.9min
         [Parallel(n_jobs=-1)]: Done 61 tasks
                                                       elapsed: 41.9min
         [Parallel(n jobs=-1)]: Done 74 tasks
                                                      elapsed: 65.1min
         [Parallel(n_jobs=-1)]: Done 89 tasks
                                                      elapsed: 97.1min
                                                       elapsed: 134.7min
         [Parallel(n_jobs=-1)]: Done 104 tasks
         [Parallel(n jobs=-1)]: Done 121 tasks
                                                       elapsed: 182.5min
                                                       elapsed: 234.5min
         [Parallel(n_jobs=-1)]: Done 138 tasks
         [Parallel(n jobs=-1)]: Done 159 out of 165 | elapsed: 291.8min remainin
         g: 11.0min
         [Parallel(n jobs=-1)]: Done 165 out of 165 | elapsed: 304.3min finished
Out[17]: GridSearchCV(cv=5, error_score='raise-deprecating',
                      estimator=SVC(C=1.0, cache size=200, class weight=None, co
         ef0=0.0,
                                    decision function shape='ovr', degree=3,
                                    gamma='scale', kernel='poly', max iter=-1,
                                    probability=False, random state=None, shrink
         ing=True,
                                    tol=0.001, verbose=False),
                      iid=False, n jobs=-1,
                      param grid={'C': array([1.e-05, 1.e-04, 1.e-03, 1.e-02, 1.
         e-01, 1.e+00, 1.e+01, 1.e+02,
                1.e+03, 1.e+04, 1.e+05]),
                                   'degree': [1, 2, 3]},
                      pre_dispatch='2*n_jobs', refit=True, return_train_score=Fa
         lse,
                      scoring=None, verbose=10)
         svm poly grid search predict = grid search.predict(test term doc)
In [18]:
In [20]: f1_score(y_test['toxic'], svm_poly_grid_search_predict)
Out[20]: 0.7991704867932765
```