```
In [10]: import pandas as pd
         import numpy as np
         from sklearn.linear model import LogisticRegression
         from sklearn.feature_extraction.text import CountVectorizer, TfidfVector
         izer
         from sklearn import model selection, naive bayes, svm
         from sklearn.svm import SVC
         from sklearn.metrics import accuracy score, f1 score
         from tqdm import tqdm
         from sklearn.model_selection import train_test_split
         from sklearn.preprocessing import MaxAbsScaler
         from sklearn.model selection import GridSearchCV
         from sklearn.metrics import roc auc score
         import os.path
         import pickle
In [20]: | 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')
```

Before we do anything, we need to get the vectors. We can download one of the pre-trained models. We downloaded the pretrained model from http://nlp.stanford.edu/data/glove.6B.zip (http://nlp.stanford.edu/data/glove.6B.zip)

```
In [21]: import numpy as np
w2v = {}

f = open("../Word2Vec_Data/glove.6B.50d.txt", "rb")

for line in f:
    w2v[line.split()[0]] = np.array(line.split()[1:]).astype(np.float)
```

```
In [22]: words not found = 0
         train doc vectors = pd.DataFrame() # creating empty final dataframe
         if os.path.isfile('../Word2Vec Data/train doc vectors.pkl'):
             train_doc_vectors = pd.read_pickle('../Word2Vec_Data/train_doc_vecto
         rs.pkl')
         else:
             for doc in tqdm(X_train.values): # looping through each document and
         cleaning it
                 temp = pd.DataFrame() # creating a temporary dataframe(store va
         lue for 1st doc & for 2nd doc remove the details of 1st & proced through
         2nd and so on..)
                 word vec = np.zeros(50)
                 temp = temp.append(pd.Series(word vec), ignore index = True) # i
         f word is present then append it to temporary dataframe
                 for word in doc.split(" "): # looping through each word of a sin
         gle document and spliting through space
                     word = word.encode("utf-8")
                     try:
                         word vec = w2v[word] # if word is present in embeddings
         (goole provides weights associate with words (300)) then proceed
                         temp = temp.append(pd.Series(word_vec), ignore_index = T
         rue) # if word is present then append it to temporary dataframe
                     except:
                         word vec = np.zeros(50)
                         words_not_found += 1
                         temp = temp.append(pd.Series(word vec), ignore index = T
         rue) # if word is present then append it to temporary dataframe
                         pass
                 doc vector = temp.mean() # take the average of each column(w0, w
         1, w2,....w300)
                 train doc vectors = train doc vectors.append(doc vector, ignore
         index = True) # append each document value to the final dataframe
             train doc vectors.to pickle("../Word2Vec Data/train doc vectors.pkl"
         print(train doc vectors.shape)
```

(39912, 50)

```
In [23]: | words not found test = 0
         test doc vectors = pd.DataFrame() # creating empty final dataframe
         if os.path.isfile('../Word2Vec Data/test doc vectors.pkl'):
             test_doc_vectors = pd.read_pickle('../Word2Vec_Data/test_doc_vector
         s.pkl')
         else:
             for doc in tqdm(X test.values): # looping through each document and
          cleaning it
                 temp = pd.DataFrame() # creating a temporary dataframe(store va
         lue for 1st doc & for 2nd doc remove the details of 1st & proced through
         2nd and so on..)
                 word vec = np.zeros(50)
                 temp = temp.append(pd.Series(word vec), ignore index = True) # i
         f word is present then append it to temporary dataframe
                 for word in doc.split(" "): # looping through each word of a sin
         gle document and spliting through space
                     word = word.encode("utf-8")
                     try:
                         word vec = w2v[word] # if word is present in embeddings
         (goole provides weights associate with words (300)) then proceed
                         temp = temp.append(pd.Series(word_vec), ignore_index = T
         rue) # if word is present then append it to temporary dataframe
                     except:
                         word vec = np.zeros(50)
                         words_not_found_test += 1
                         temp = temp.append(pd.Series(word vec), ignore index = T
         rue) # if word is present then append it to temporary dataframe
                         pass
                 doc vector = temp.mean() # take the average of each column(w0, w
         1, w2,....w300)
                 test doc vectors = test doc vectors.append(doc vector, ignore in
         dex = True) # append each document value to the final dataframe
             test doc vectors.to pickle("../Word2Vec Data/test doc vectors.pkl")
         print(test doc vectors.shape)
```

(19659, 50)

```
In [24]: train_doc_vectors.fillna(0)
    test_doc_vectors.fillna(0)
    scaler = MaxAbsScaler()
    # using averaged word embeddings
    train_term_doc = scaler.fit_transform(train_doc_vectors)
    test_term_doc = scaler.fit_transform(test_doc_vectors)
```

```
In [25]: def logistic regression with CV(label):
         if os.path.isfile('Models/ridge_lr_' + label + '_w2v.sav') and os.pa
th.isfile('Models/lasso_lr_' + label + '_w2v.sav'):
                 ridge_logistic_regressor_grid_cv = pickle.load(open('Models/ridg
         e_lr_' + label + '_w2v.sav', 'rb'))
                  lasso_logistic_regressor_grid_cv = pickle.load(open('Models/lass
         o_lr_' + label + '_w2v.sav', 'rb'))
             else:
                  ridge_logistic_regressor = LogisticRegression(penalty="12", solv
         er="liblinear", max_iter = 2000)
                  lasso logistic regressor = LogisticRegression(penalty="11", solv
         er="liblinear", max iter = 2000)
                  ridge logistic regressor grid cv = GridSearchCV(estimator=ridge
         logistic regressor,
                                                                param_grid={'C':np.
         logspace(-4, 4, 20)}, cv= 5, iid=False)
                  lasso logistic regressor grid cv = GridSearchCV(estimator=lasso_
         logistic regressor,
                                                                param grid={'C':np.
         logspace(-4, 4, 20)}, cv= 5, iid=False)
                 ridge logistic regressor grid cv.fit(train term doc, y train[lab
         el])
                  lasso logistic regressor grid cv.fit(train term doc, y train[lab
         el])
                 pickle.dump(ridge logistic regressor grid cv, open('Models/ridge
         lr ' + label + ' w2v.sav', 'wb'))
                 pickle.dump(lasso logistic regressor grid cv, open('Models/lasso
         <u>lr_'</u> + label + '_w2v.sav', 'wb'))
             ridge train pred = ridge logistic regressor grid cv.predict(train te
         rm doc)
             lasso_train_pred = lasso_logistic_regressor_grid_cv.predict(train_te
         rm doc)
             ridge test pred = ridge logistic regressor grid cv.predict(test term
         doc)
             lasso test pred = lasso logistic regressor grid cv.predict(test term
         _doc)
             print(label + " Ridge Train Accuracy - " + str(ridge logistic regres
         sor grid cv.score(train term doc, y train[label])))
             print(label + " Lasso Train Accuracy - " + str(lasso logistic regres
         sor grid cv.score(train term doc, y train[label])) + '\n')
             print(label + " Ridge Train F1 Score - " + str(f1 score(y train[labe
         1], ridge train pred)))
             print(label + " Lasso Train F1 Score - " + str(f1 score(y train[labe
         1], lasso train pred)) + '\n')
             print(label + " Ridge Train ROC-AUC Score - " + str(roc_auc_score(y_
         train[label], ridge train pred)))
              print(label + " Lasso Train ROC-AUC Score - " + str(roc auc score(y
         train[label], lasso train pred)) + '\n')
```

11/17/2019 Logistic Regression

```
print(label + " Ridge Test Accuracy - " + str(ridge_logistic_regress
or_grid_cv.score(test_term_doc, y_test[label])))
    print(label + " Lasso Test Accuracy - " + str(lasso_logistic_regress
or_grid_cv.score(test_term_doc, y_test[label])) + '\n')

    print(label + " Ridge Test F1 Score - " + str(f1_score(y_test[label]), ridge_test_pred)))
    print(label + " Lasso Test F1 Score - " + str(f1_score(y_test[label]), lasso_test_pred)) + '\n')

    print(label + " Ridge Test ROC-AUC Score - " + str(roc_auc_score(y_test[label]), ridge_test_pred)))
    print(label + " Lasso Test ROC-AUC Score - " + str(roc_auc_score(y_test[label]), lasso_test_pred)) + '\n\n')
```

11/17/2019 Logistic Regression

/Users/abhay/anaconda3/lib/python3.7/site-packages/sklearn/base.py:306: UserWarning: Trying to unpickle estimator LogisticRegression from versi on 0.21.3 when using version 0.21.2. This might lead to breaking code or invalid results. Use at your own risk.

UserWarning)

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UserWarning)

```
toxic Ridge Train Accuracy - 0.8567097614752456
toxic Lasso Train Accuracy - 0.8567348165965123
toxic Ridge Train F1 Score - 0.6864755221753194
toxic Lasso Train F1 Score - 0.687199124726477
toxic Ridge Train ROC-AUC Score - 0.7755787448157945
toxic Lasso Train ROC-AUC Score - 0.7762289078191099
toxic Ridge Test Accuracy - 0.8598097563456941
toxic Lasso Test Accuracy - 0.8585889414517524
toxic Ridge Test F1 Score - 0.6866757617098681
toxic Lasso Test F1 Score - 0.6821404070432198
toxic Ridge Test ROC-AUC Score - 0.7756724605128862
toxic Lasso Test ROC-AUC Score - 0.7724120192205299
severe toxic Ridge Train Accuracy - 0.9737422329124072
severe_toxic Lasso Train Accuracy - 0.9739426738825416
severe toxic Ridge Train F1 Score - 0.2305433186490455
severe_toxic Lasso Train F1 Score - 0.25501432664756446
severe toxic Ridge Train ROC-AUC Score - 0.5719634302498421
severe toxic Lasso Train ROC-AUC Score - 0.581664532521584
severe toxic Ridge Test Accuracy - 0.9719721247265883
severe toxic Lasso Test Accuracy - 0.9722264611628262
severe toxic Ridge Test F1 Score - 0.26238286479250333
severe toxic Lasso Test F1 Score - 0.2680965147453083
severe toxic Ridge Test ROC-AUC Score - 0.5891942359176479
severe toxic Lasso Test ROC-AUC Score - 0.5911558941605408
obscene Ridge Train Accuracy - 0.9050160352776108
obscene Lasso Train Accuracy - 0.9050160352776108
obscene Ridge Train F1 Score - 0.5899405083829097
obscene Lasso Train F1 Score - 0.5898517797251974
obscene Ridge Train ROC-AUC Score - 0.7285465099794247
obscene Lasso Train ROC-AUC Score - 0.7284726721426786
obscene Ridge Test Accuracy - 0.8960272648659647
obscene Lasso Test Accuracy - 0.8957729284297269
obscene Ridge Test F1 Score - 0.5537117903930131
obscene Lasso Test F1 Score - 0.5521311475409836
obscene Ridge Test ROC-AUC Score - 0.7115051210483617
obscene Lasso Test ROC-AUC Score - 0.7106106666655351
```

```
threat Ridge Train Accuracy - 0.9914812587692925
threat Lasso Train Accuracy - 0.9914060934054921
threat Ridge Train F1 Score - 0.0
threat Lasso Train F1 Score - 0.017191977077363897
threat Ridge Train ROC-AUC Score - 0.5
threat Lasso Train ROC-AUC Score - 0.5043359535262604
threat Ridge Test Accuracy - 0.9929803143598351
threat Lasso Test Accuracy - 0.99287857978534
threat Ridge Test F1 Score - 0.0
threat Lasso Test F1 Score - 0.0410958904109589
threat Ridge Test ROC-AUC Score - 0.5
threat Lasso Test ROC-AUC Score - 0.5107414980077197
insult Ridge Train Accuracy - 0.8977500501102426
insult Lasso Train Accuracy - 0.8978252154740429
insult Ridge Train F1 Score - 0.5065892878732923
insult Lasso Train F1 Score - 0.5070116054158608
insult Ridge Train ROC-AUC Score - 0.6860466086747585
insult Lasso Train ROC-AUC Score - 0.6862513307318064
insult Ridge Test Accuracy - 0.8924156874713871
insult Lasso Test Accuracy - 0.8925174220458822
insult Ridge Test F1 Score - 0.4729628706703215
insult Lasso Test F1 Score - 0.4731987035651958
insult Ridge Test ROC-AUC Score - 0.6679969588333052
insult Lasso Test ROC-AUC Score - 0.6680556546130787
identity hate Ridge Train Accuracy - 0.9760222489476849
identity hate Lasso Train Accuracy - 0.9760222489476849
identity hate Ridge Train F1 Score - 0.0
identity hate Lasso Train F1 Score - 0.0
identity hate Ridge Train ROC-AUC Score - 0.5
identity hate Lasso Train ROC-AUC Score - 0.5
identity hate Ridge Test Accuracy - 0.9772114553130882
identity hate Lasso Test Accuracy - 0.9772114553130882
identity hate Ridge Test F1 Score - 0.0
identity hate Lasso Test F1 Score - 0.0
identity hate Ridge Test ROC-AUC Score - 0.5
identity hate Lasso Test ROC-AUC Score - 0.5
```

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'precision', 'predicted', average, warn_for)

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In []: