Import libraries

```
In [1]: import pandas as pd
          import string
          import re
          import nltk
          import emoji
          import numpy
          import time
          from sklearn.metrics import f1_score
          from sklearn.metrics import make_scorer
          from sklearn.feature_extraction.text import CountVectorizer
          from sklearn.feature_extraction.text import TfidfVectorizer
          from sklearn import model_selection, svm
          from sklearn.preprocessing import LabelEncoder
          \textbf{from} \  \, \text{sklearn.metrics} \  \, \textbf{import} \  \, \text{classification\_report}, \  \, \text{confusion\_matrix}, \textbf{f1}\_\textbf{score}, \  \, \text{accuracy\_score}
          \textbf{from} \  \, \text{sklearn.metrics} \  \, \textbf{import} \  \, \text{confusion\_matrix}
          from sklearn.model_selection import GridSearchCV
          from sklearn.ensemble import RandomForestClassifier
          from nltk.tokenize import word_tokenize
          from nltk.stem.porter import PorterStemmer
          from sklearn.pipeline import Pipeline
          from sklearn.pipeline import make_pipeline
          import seaborn as sns
          import matplotlib.pyplot as plt
          import np
          from collections import Counter
          from textblob import TextBlob
          from langdetect import detect
```

Function to generate confusion matrix

```
def make_confusion_matrix(cf, group_names=None, categories='auto', count=True, percent=True, cbar=True, xyticks=True,
                            xyplotlabels=True,
                             sum_stats=True,
                            figsize=None,
                            cmap='Blues'
                            title=None):
    # CODE TO GENERATE TEXT INSIDE EACH SQUARE
    blanks = ['' for i in range(cf.size)]
    if group_names and len(group_names)==cf.size:
        group_labels = ["{}\n".format(value) for value in group_names]
         group_labels = blanks
         group\_counts = \hbox{\tt ["\{0:0.0f\}\n".format(value)} \ \textit{for} \ value \ \textit{in} \ cf.flatten()]
         group_counts = blanks
        group\_percentages = \hbox{\tt ["{0:.2\%}]".format}(value) \ \ \mbox{\tt for} \ \ value \ \ \mbox{\tt in} \ \ \mbox{\tt cf.flatten()/np.sum(cf)]}
         group_percentages = blanks
    box_labels = [f"{v1}{v2}{v3}".strip() for v1, v2, v3 in zip(group_labels,group_counts,group_percentages)]
    box_labels = np.asarray(box_labels).reshape(cf.shape[0],cf.shape[1])
    # SET FIGURE PARAMETERS ACCORDING TO OTHER ARGUMENTS
    if figsize==None:
         #Get default figure size if not set
         figsize = plt.rcParams.get('figure.figsize')
    if xvticks==False:
         #Do not show categories if xyticks is False
         categories=False
```

```
# MAKE THE HEATMAP VISUALIZATION
plt.figure(figsize=figsize)
sns.heatmap(cf,annot=box_labels,fmt="",cmap=cmap,cbar=cbar,xticklabels=categories,yticklabels=categories)
plt.ylabel('True label')
plt.xlabel('Predicted label')
plt.savefig('conf.pdf')

if title:
    plt.title(title)
```

Read, clean, prepare dataset

```
In [3]: # Read dataset
          test_tweets = pd.read_csv('mediaeval-2015-testset.txt', sep='\\t', engine='python', encoding='utf-8')
          train_tweets = pd.read_csv('mediaeval-2015-trainingset.txt', sep='\\t', engine='python', encoding='utf-8')
          def removeColumns(tweets):
             tweets.drop(["timestamp"],axis=1,inplace=True)
tweets.drop(["username"],axis=1,inplace=True)
             tweets.drop(["tweetId"],axis=1,inplace=True)
             tweets.drop(["userId"],axis=1,inplace=True)
             tweets.drop(["imageId(s)"],axis=1,inplace=True)
          # Remove Columns
          removeColumns(train_tweets)
          removeColumns(test_tweets)
          def converHumourLabel(tweets):
             tweets.loc[tweets['label'] == 'humor', 'label'] = 'fake'
          # Change 'humor' to 'fake'
          converHumourLabel(train_tweets)
          converHumourLabel(test_tweets)
```

```
def englishTweets(tweets):
    # Generate lang
    for value in tweets["preprocessed"]:
        if not (value == "en"):
             ls.append(detect(str(tweets)))
         else:
             ls.append("ND")
    tweets["lang"] = ls
    return tweets
def clean(text):
    # Remove URLs
    tweet = re.sub(r'http\S+', '', text)
    # Remove mentions
    tweet = re.sub(r'@\w+', '', tweet)
    # Remove all the special characters
    tweet = re.sub(r'\W', ' ', tweet)
    # Convert to lower case
    tweet = tweet.lower()
    return tweet
# Remove stop words
def remove_stopwords(text):
    event_words = ['hurricane', 'sochi',
                      ['hurricane', 'sochi',
'soldier', 'hurricanesandy',
'liberty', 'bringbackourgirls', 'jersey', 'manhattan','flood',
'nj', 'new', 'nyc', 'ny', 'york', 'statue',
'statueofliberty', 'shark', 'newyork',
'tomb', 'mh370', 'sandy', 'huracán', 'boston',
'columbianchemicals',
                        'flooding', 'cuba']
    twitter = ['via', 'photo', 'rt']
en_stop_words = nltk.corpus.stopwords.words('english')
     sp_stop_words = nltk.corpus.stopwords.words('spanish')
     stopword = en_stop_words + sp_stop_words + twitter + event_words
    text= ' '.join([word for word in text.split() if word not in stopword])
     return text
```

```
\textbf{def} \ \mathsf{stemming}(\mathsf{text}) \colon
   tokens = word_tokenize(text)
   stemmer = PorterStemmer()
   stems = [stemmer.stem(w) for w in tokens]
   return ' '.join(stems)
train_tweets['preprocessed'] = train_tweets['tweetText'].apply(clean)
test_tweets['preprocessed'] = test_tweets['tweetText'].apply(clean)
# train_tweets = englishTweets()
# Remove stopwords
train_tweets['preprocessed'] = train_tweets['preprocessed'].apply(remove_stopwords)
test_tweets['preprocessed'] = test_tweets['preprocessed'].apply(remove_stopwords)
# Perform stemming - ineffective
# train_tweets['preprocessed'] = train_tweets['preprocessed'].apply(stemming)
# test_tweets['preprocessed'] = test_tweets['preprocessed'].apply(stemming)
test_tweets.to_csv('test_tweets.csv')
train_tweets.to_csv('train_tweets.csv')
train_tweets.head()
```

fave place world

[3]:	tweetText	label	preprocessed
0	¿Se acuerdan de la película: "El día después d	fake	acuerdan película día después mañana recuerda
1	@milenagimon: Miren a Sandy en NY! Tremenda i	fake	miren tremenda imagen parece día independencia
2	2 Buena la foto del Huracán Sandy, me recuerda a	fake	buena foto recuerda película día independencia
3	Scary shit #hurricane #NY http://t.co/e4JLBUfH	fake	scary shit

4 My fave place in the world #nyc #hurricane #sa... fake

Hyperparameter tuning

```
In [44]: # Define x and y train and test set
          x_train = train_tweets['preprocessed']
          x_test = test_tweets['preprocessed']
          y_train = train_tweets['label']
          y_test = test_tweets['label']
           # classifier
          pipeline = Pipeline([
              ('tfidf', TfidfVectorizer(max_df=0.5, min_df=2, ngram_range=(1,1))),
              ('clf', svm.SVC(kernel='linear',C=1.9, gamma='auto')),
          parameters = {
              'tfidf__max_features': (3600,10000)
          f1_scorer = make_scorer(f1_score, average='micro', pos_label="fake")
          \ensuremath{\textit{\#}} find the best parameters for both the feature extraction and the
           # classifier
          grid_search = GridSearchCV(pipeline, parameters, n_jobs=-1, cv=10 , verbose=1,scoring=f1_scorer)
          print("Performing grid search...")
          print("pipeline:", [name for name, _ in pipeline.steps])
          print("parameters:")
          print(parameters)
          t0 = time.time()
          grid_search.fit(x_train, y_train)
          print("done in %0.3fs" % (time.time() - t0))
          print()
          print("Best score: %0.3f" % grid_search.best_score_)
          print("Best parameters set:")
          best_parameters = grid_search.best_estimator_.get_params()
          for param_name in sorted(parameters.keys()):
              print("\t%s: %r" % (param_name, best_parameters[param_name]))
         Performing grid search...
         pipeline: ['tfidf', 'clf']
         parameters:
         {'tfidf_max_features': (3600, 10000)}
         Fitting 10 folds for each of 2 candidates, totalling 20 fits
         [Parallel(n_jobs=-1)]: Using backend LokyBackend with 12 concurrent workers.
         [Parallel(n_jobs=-1)]: Done 18 out of 20 | elapsed: 14.4s remaining: 1.6s
        [Parallel(n_jobs=-1)]: Done 20 out of 20 | elapsed: 14.5s finished
         done in 19.590s
         Best score: 0.846
         Best parameters set:
                tfidf max features: 3600
         predictions = grid_search.predict(x_test)
         print("micro F1 score: " + str(f1_score(y_test, predictions, average='micro')))
         micro F1 score: 0.8508331129330866
```

Train and evaluate model

```
In [4]: # Define x and y train and test set
        x_train = train_tweets['preprocessed']
        x_test = test_tweets['preprocessed']
        y_train = train_tweets['label']
        y_test = test_tweets['label']
        # TF-IDF vectorizer TfidfVectorizer()
        # Tfidf_vect = TfidfVectorizer(max_features=3600, min_df=2, max_df=0.5, ngram_range=(1,1))
        # Tfidf_vect = TfidfVectorizer()
        # unigrams performs very well
        # Tfidf_vect = CountVectorizer(max_features=3600, min_df=2, ngram_range=(1,1))
        Tfidf_vect = TfidfVectorizer(max_features=3600, min_df=2, max_df=0.5, ngram_range=(1,1))
        Tfidf_vect.fit(train_tweets['preprocessed'])
        train_x_tfidf = Tfidf_vect.transform(x_train)
        test_x_tfidf = Tfidf_vect.transform(x_test)
        labels = ['True Neg','False Pos','False Neg','True Pos']
categories = ['Fake', 'Real']
        # Classifier - Support Vector Machine
        SVM = svm.SVC(C=1.9, kernel='linear')
        SVM.fit(train_x_tfidf, y_train)
svm_predictions = SVM.predict(test_x_tfidf)
        svm_cf_matrix = confusion_matrix(y_test, svm_predictions)
        micro F1-score: 0.8804549061094947
```

macro F1-score: 0.8589963182538833 2250 2000 True Neg False Pos 2402 63.53% 162 Fake - 1750 4.28% - 1500 Frue label - 1250 - 1000 False Neg True Pos 927 - 750 290 Real 24.52% 7.67% - 500 -250 Fake Real

Predicted label