Data 641 - Lab 3

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Exercise 1

Importing Base Modules and Data

```
[1]: import numpy as np
  import pandas as pd
  import re
  import string

[2]: dt_twt = pd.read_csv('COVID19_Dataset-text_labels_only.csv')
```

Exercise 2

Extracting and Normalizing Text

```
[3]: ## Removing Hashtags

dt_twt['No Hash'] = [re.sub('#', '', line) for line in dt_twt['Tweet']] ##

$\to Removing pound sign

## Example:

print('Original Tweet:\n', dt_twt['Tweet'].iloc[100])

print('\nTweet without hashtag:\n', dt_twt['No Hash'].iloc[100])
```

Original Tweet:

#coronavirus #virus Reasons to buy back stocks today 1500 new cases 60 deaths 2700 recovered #markets #portfolio

Tweet without hashtag:

coronavirus virus Reasons to buy back stocks today 1500 new cases 60 deaths 2700 recovered markets portfolio

```
[4]: ## Dropping words with less than 2 characters

from nltk import word_tokenize

twt_longs = []

for tweet in dt_twt['No Hash']:
```

```
twt_words = word_tokenize(tweet)
  twt_n = ' '.join([word for word in twt_words if re.match('\w{2,}|\d+',word)])
  twt_longs.append(twt_n)

dt_twt['No Short Word'] = twt_longs
```

```
[5]: dt_twt['No Short Word'].iloc[326]
```

[5]: '5 Rings Podcast Road To Tokyo March 10th 2020 COVID-19 and the Possibility of an Olympic Cancellation 24thminute KevLaramee coronavirus COVID19 Tokyo2020 IOC OlympicPodcast'

```
[6]: ## Lemmatizing
from nltk.stem import WordNetLemmatizer
lemmatizer = WordNetLemmatizer()

tw_temp=[]

for tweet in dt_twt['No Short Word']:
    twt_lm = [lemmatizer.lemmatize(word) for word in word_tokenize(tweet)]
    tw_temp.append(' '.join(twt_lm))

dt_twt['Lemmatized'] = tw_temp
```

```
[8]: (dt_twt['Lemmatized'].iloc[444])
```

[8]: 'State of emergency declared in New York a number of coronavirus patient climb'

Exercise 3

Instantiating Vectorizer (tf-idf, PCA, Sparse to Dense)

```
[11]: ## Instantiating PCA
from sklearn.decomposition import PCA
pca = PCA()

## Setting the range of components to try
```

```
ncomps = [5, 10, 20, 50, 75, 100]
[12]: ## Instantiating Dense Transformer
      from sklearn.base import TransformerMixin
      class SparseToDense(TransformerMixin):
          def fit(self, X, y = None, **fit_params):
              return self
          def transform(self, X, y = None, **fit_params):
              return X.toarray()
     Exercise 4
     Setting up ML Process (Transformation, Pipeline, SVM hyperparams, CV, etc.)
[13]: ## Setting up data set
      X = dt_twt['Lemmatized']
      y = dt_twt['Is_Unreliable']
      X_count = tfidf.fit_transform(X)
[14]: # Setting up the pipeline
      from sklearn.pipeline import Pipeline
      from sklearn.preprocessing import StandardScaler
      from sklearn.svm import SVC
      pipe = Pipeline([
          ('vectorize', tfidf),
          ('densify', SparseToDense()),
          ('scale', StandardScaler()),
          ('dim_red', pca),
          ('classify', SVC())
      ])
[15]: # SVC hyperparameters
      kernel = ['rbf', 'linear', 'poly', 'sigmoid'] ## kernels
      C = [0.001, 0.01, 0.1, 1, 10]
      # parameters' grid
      params = {
```

'dim_red__n_components': ncomps,

```
'classify__kernel': kernel,
          'classify__C': C
      }
[16]: # Setting Cross Validation scheme for inner and outer loops
      from sklearn.model_selection import cross_validate, KFold, GridSearchCV
      inner_cv = KFold(n_splits = 3, shuffle = True, random_state = 1)
      outer_cv = KFold(n_splits = 5, shuffle = True, random_state = 1)
      # Setting GridSearch for inner loop
      grid_SVC = GridSearchCV(pipe, params, cv = inner_cv)
      # Nested CV scores
      scores = cross_validate(grid_SVC, X = X, y = y, cv = outer_cv,
                              scoring = ['roc_auc', 'accuracy', 'f1', __
       return_estimator = True)
      print ('Done!')
[17]: | ## Model Performance: AUC, Accuracy, F1
      auc = scores['test_roc_auc']
      accuracy = scores['test_accuracy']
      f1 = scores['test_f1']
      precision = scores['test_precision']
      recall = scores['test_recall']
      estimators = scores['estimator']
[24]: ## Average scores for models
      print('Average Scores:\n')
      print('Accuracy: {} \nPrecision: {}\nRecall: {}\nF1: {}'.format(
          accuracy.mean().round(4), precision.mean().round(4), recall.mean().round(4), 
       \rightarrowf1.mean().round(4)))
     Average Scores:
     Accuracy: 0.7696
     Precision: 0.7771
```

Recall: 0.76 F1: 0.7661

```
[19]: ## Best Performance

for score in estimators:
    print(score.best_params_)
    print('\n')

{'classify__C': 0.01, 'classify__kernel': 'linear', 'dim_red__n_components':
100}

{'classify__C': 0.01, 'classify__kernel': 'linear', 'dim_red__n_components':
100}

{'classify__C': 0.01, 'classify__kernel': 'linear', 'dim_red__n_components':
100}

{'classify__C': 1, 'classify__kernel': 'sigmoid', 'dim_red__n_components': 50}

{'classify__C': 1, 'classify__kernel': 'linear', 'dim_red__n_components': 50}
```