```
In [1]: #!unzip dank_data-master.zip
#!pip install tensorflow_addons
#!wget http://nlp.stanford.edu/data/glove.6B.zip
#!unzip glove*.zip
```

```
In [2]: | import glob
        import pandas as pd
        import warnings
        warnings.filterwarnings("ignore")
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        from tensorflow.keras.layers import Dense,Input,MaxPool1D,Activation,Dropout,F
        latten, Embedding, LSTM, concatenate
        from tensorflow.keras.models import Model
        import tensorflow as tf
        import numpy as np
        from tensorflow.keras.preprocessing.text import Tokenizer
        from sklearn.preprocessing import LabelEncoder
        from sklearn.preprocessing import StandardScaler
        import tensorflow_addons as tfa
        from tensorflow.keras.callbacks import LearningRateScheduler
        from tensorflow.keras.callbacks import ReduceLROnPlateau
        from tensorflow.keras.callbacks import ModelCheckpoint
        from tensorflow.keras.callbacks import EarlyStopping
        from sklearn.metrics import confusion matrix, accuracy score, f1 score
        import seaborn as sns
        import matplotlib.pyplot as plt
```

```
In [3]: training='/content/dank_data-master/data/training/*'
    test='/content/dank_data-master/data/test/*'
    validation='/content/dank_data-master/data/validation/*'
```

```
In [4]: training = glob.glob(training)
  test = glob.glob(test)
  validation = glob.glob(validation)
```

```
In [5]: final_dank=pd.read_csv('/content/dank_data-master/data/final_dank.csv')
    train_labels = [fn.split(',')[-1].split('.')[0].strip() for fn in training]
    validation_labels = [fn.split(',')[-1].split('.')[0].strip() for fn in validat
    ion]
    test_labels = [fn.split(',')[-1].split('.')[0].strip() for fn in test]
```

```
In [6]: for labels in train_labels:
            if labels==train_labels[0]:
                 train_data =final_dank[final_dank['id']==labels]
            else :
                 train_data = train_data.append(final_dank[final_dank['id'] == labels], sor
        t=False)
        for labels in validation_labels:
            if labels==validation_labels[0]:
                 val data =final_dank[final_dank['id']==labels]
            else :
                 val_data =val_data.append(final_dank[final_dank['id']==labels],sort=Fa
        lse)
        for labels in test_labels:
            if labels==test_labels[0]:
                 test_data =final_dank[final_dank['id']==labels]
            else :
                test_data =test_data.append(final_dank[final_dank['id']==labels],sort=
        False)
        print(train_data.shape)
        print(test_data.shape)
        print(val_data.shape)
        train_data.head(5)
        (3405, 68)
```

(1719, 68)

(1688, 68)

Out[6]:

	Unnamed: 0	level_0	index	author	awards	processed_words	created_utc
15987	32142	33669.0	33669.0	orby9990	0	['true', 'stori', 'btwthe', 'world', 'iranian'	1.584339e+09
58209	103760	8570.0	40223.0	DarkKnight_Jedi	0	['deserv', 'march', 'prove', 'cost', 'husband'	1.584860e+09
45465	76129	79789.0	11442.0	LifeAfterRedditFalls	0	['think', 'share', 'peac', 'final', 'option']	1.584535e+09
9635	22567	23666.0	23666.0	subbot9		['love', 'love', 'reddit']	1.584394e+09
40396	68463	71751.0	3404.0	CalmProfit	0	['behistorian', 'wonder', 'internet', 'respond	1.584577e+09
4							•

```
In [7]: | train_data_words=train_data['processed_words'].values
        validation_words=val_data['processed_words'].values
        test_data_words=test_data['processed_words'].values
        tokenizer = Tokenizer()
        tokenizer.fit_on_texts(train_data_words)
        vocab size=len(tokenizer.word index)
        encoded_Xtrain_words = [tf.keras.preprocessing.text.one_hot(d, vocab_size,filt
        ers='!"#$%&()*+,-./:;<=>?@[\\]^`{|}~\t\n') for d in train_data_words]
        encoded_validation_words = [tf.keras.preprocessing.text.one_hot(d, vocab_size,
        filters='!"#$%&()*+,-./:;<=>?@[\\]^`{|}~\t\n') for d in validation_words]
        encoded_Xtest_words = [tf.keras.preprocessing.text.one_hot(d, vocab_size,filte
        rs='!"#$%&()*+,-./:;<=>?@[\\]^`{|}~\t\n') for d in test_data_words]
        padded Xtrain words = tf.keras.preprocessing.sequence.pad sequences(encoded Xt
        rain_words, maxlen=20, padding='post')
        padded Xvalidation words = tf.keras.preprocessing.sequence.pad sequences(encod
        ed_validation_words, maxlen=20, padding='post')
        padded Xtest words = tf.keras.preprocessing.sequence.pad sequences(encoded Xte
        st words, maxlen=20, padding='post')
```

```
In [8]: embeddings_index = dict()
    f = open('/content/glove.6B.300d.txt')

for line in f:
    values = line.split()
    word = values[0]
    coefs = np.asarray(values[1:], dtype='float32')
    embeddings_index[word] = coefs

f.close()
    print('Loaded %s word vectors.' % len(embeddings_index))
```

Loaded 400000 word vectors.

```
In [9]: embedding_matrix = np.zeros((vocab_size+1, 300))
    for word, i in tokenizer.word_index.items():
        embedding_vector = embeddings_index.get(word)
        if embedding_vector is not None:
        embedding_matrix[i] = embedding_vector
```

```
In [10]:
         labelencoder = LabelEncoder()
         labelencoder.fit(train_data['subreddit'].values)
         subreddit train=labelencoder.transform(train_data['subreddit'].values).reshape
         (-1,1)
         subreddit validation=labelencoder.transform(val data['subreddit'].values).resh
         ape(-1,1)
         subreddit test=labelencoder.transform(test data['subreddit'].values).reshape(-
         1, 1)
         print(subreddit_train.shape)
         print(subreddit test.shape)
         print(subreddit_validation.shape)
         (3405, 1)
         (1719, 1)
         (1688, 1)
In [11]: labelencoder = LabelEncoder()
         labelencoder.fit(train data['is nsfw'].values)
         is_nsfw_train=labelencoder.transform(train_data['is_nsfw'].values).reshape(-1,
         is nsfw validation=labelencoder.transform(val data['is nsfw'].values).reshape(
         -1,1)
         is nsfw test=labelencoder.transform(test data['is nsfw'].values).reshape(-1,1)
         print(is_nsfw_train.shape)
         print(is nsfw test.shape)
         print(is nsfw validation.shape)
         (3405, 1)
         (1719, 1)
         (1688, 1)
In [12]: | scaler = StandardScaler()
         scaler=scaler.fit(train data['created utc'].values.reshape(-1, 1))
         created_utc_train=scaler.transform(train_data['created_utc'].values.reshape(-1
         , 1))
         created utc validation=scaler.transform(val data['created utc'].values.reshape
         (-1, 1)
         created utc test=scaler.transform(test data['created utc'].values.reshape(-1,
         1))
         print(created utc train.shape)
         print(created_utc_test.shape)
         print(created utc validation.shape)
         (3405, 1)
         (1719, 1)
         (1688, 1)
```

```
In [13]:
         time_of_day_train=(train_data['time_of_day'].values).reshape(-1,1)
         time_of_day_validation=(val_data['time_of_day'].values).reshape(-1,1)
         time_of_day_test=(test_data['time_of_day'].values).reshape(-1,1)
         print(time_of_day_train.shape)
         print(time_of_day_validation.shape)
         print(time_of_day_test.shape)
         (3405, 1)
         (1688, 1)
         (1719, 1)
In [14]:
         scaler = StandardScaler()
         scaler=scaler.fit(train_data['subscribers'].values.reshape(-1, 1))
         subscribers_train=scaler.transform(train_data['subscribers'].values.reshape(-1
         , 1))
         subscribers_validation=scaler.transform(val_data['subscribers'].values.reshape
         (-1, 1)
         subscribers_test=scaler.transform(test_data['subscribers'].values.reshape(-1,
         1))
         print(subscribers train.shape)
         print(subscribers_validation.shape)
         print(subscribers_test.shape)
         (3405, 1)
         (1688, 1)
         (1719, 1)
```

```
In [15]: | #words embeding Layer
         words =Input(shape=(20,),name="words")
         embeding=Embedding(vocab_size+1,300,weights=[embedding_matrix],input_length=20
         ,trainable=False)(words)
         lstm layer=LSTM(100)(embeding)
         flatten1 = Flatten(data_format='channels_last')(lstm_layer)
         #categore data
         subreddit train layer =Input(shape=(subreddit_train.shape[1],),name="subreddit
          _train_layer")
         flatten2= Flatten(data format='channels last')(subreddit train layer)
         ###
         is_nsfw_train_layer =Input(shape=(is_nsfw_train.shape[1],),name="is_nsfw_train
          _layer")
         flatten3 = Flatten(data_format='channels_last')(is_nsfw_train_layer)
         ####
         time_of_day_train_layer =Input(shape=(time_of_day_train.shape[1],),name="time_
         of_day_train_layer")
         flatten4 = Flatten(data_format='channels_last')(time_of_day_train_layer)
         #numeric data
         created utc train layer =Input(shape=(created utc train.shape[1],),name="creat
         ed utc train layer")
         created utc dence = Dense(units=3,activation='relu',kernel initializer=tf.kera
         s.initializers.glorot_normal(seed=33))(created_utc_train_layer)
         #numeric data
         subscribers_train_layer =Input(shape=(subscribers_train.shape[1],),name="subsc
         ribers train layer")
         subscribers dence = Dense(units=3,activation='relu',kernel initializer=tf.kera
         s.initializers.glorot_normal(seed=33))(subscribers_train_layer)
         #concat layer
         concatenated = concatenate([subscribers train layer,created utc dence,flatten4
         ,flatten3,flatten2,flatten1],axis = -1)
         dense_layer1 = Dense(units=128,activation='relu',kernel_initializer=tf.keras.i
         nitializers.glorot normal(seed=33))(concatenated)
         dropout1=Dropout(0.3)(dense layer1)
         dense_layer2 = Dense(units=64,activation='relu',kernel_initializer=tf.keras.in
         itializers.glorot normal(seed=33))(dropout1)
         dropout2=Dropout(0.3)(dense_layer2)
         dense layer3 = Dense(units=64,activation='relu',kernel initializer=tf.keras.in
         itializers.glorot_normal(seed=33))(dropout2)
         Out = Dense(units=2,activation='softmax',kernel_initializer=tf.keras.initializ
         ers.glorot_normal(seed=3),name='Output')(dense_layer3)
         model = Model(inputs=[words,subreddit_train_layer,is_nsfw_train_layer,time_of_
         day_train_layer,created_utc_train_layer,subscribers_train_layer],outputs=Out)
         #model = Model(inputs=[words, subreddit train layer, is nsfw train layer, time of
         _day_train_layer,created_utc_train_layer,subscribers_train_layer],outputs=Out)
         model.summary()
```

Model: "model"

Layer (type)	Output Shape		
words (InputLayer)	[(None, 20)]	0	
embedding (Embedding)	(None, 20, 300)	2712000	words[0][0]
created_utc_train_layer (InputL	[(None, 1)]	0	
time_of_day_train_layer (InputL	[(None, 1)]	0	
is_nsfw_train_layer (InputLayer	[(None, 1)]	0	
subreddit_train_layer (InputLay	[(None, 1)]	0	
lstm (LSTM) [0]	(None, 100)	160400	embedding[0]
subscribers_train_layer (InputL	[(None, 1)]	0	
dense (Dense) train_layer[0][0]	(None, 3)	6	created_utc_
flatten_3 (Flatten) train_layer[0][0]	(None, 1)	0	time_of_day_
flatten_2 (Flatten) n_layer[0][0]	(None, 1)	0	is_nsfw_trai
flatten_1 (Flatten) ain_layer[0][0]	(None, 1)	0	subreddit_tr
flatten (Flatten)	(None, 100)	0	lstm[0][0]
<pre>concatenate (Concatenate) train_layer[0][0]</pre>	(None, 107)	0	subscribers_ dense[0][0] flatten_3[0]
[0]			flatten_2[0]

	[a]	<u>-</u>	-		flatten_1[0]
	[0]				flatten[0]
	dense_2 (Dense) [0][0]	(None,	128)	13824	concatenate
	dropout (Dropout) [0]	(None,	128)	0	dense_2[0]
	dense_3 (Dense) [0]	(None,	64)	8256	dropout[0]
	dropout_1 (Dropout) [0]	(None,	64)	0	dense_3[0]
	dense_4 (Dense) [0]	(None,	64)	4160	dropout_1[0]
	Output (Dense) [0]	(None,	2)	130	dense_4[0]
	Total params: 2,898,776 Trainable params: 186,776 Non-trainable params: 2,712,000				
	4			_	>
In [16]:	<pre>def scheduler(epoch,lr): if((epoch+1)%3==0): lr=lr*0.95 return lr else: return lr</pre>				

```
In [17]: | filepath="model_save/weights-{epoch:02d}-{val_accuracy:.4f}.h5"
         checkpoint = ModelCheckpoint(filepath=filepath, monitor='val accuracy',mode='a
         uto')
         lrschedule = tf.keras.callbacks.LearningRateScheduler(scheduler,verbose=0.1)
         #stop the training if your validation accuracy is not increased in last 2 epoc
         hs.
         early_stop= EarlyStopping(monitor='val_accuracy', patience=2,verbose=1)
         #If your validation accuracy at that epoch is less than previous epoch accurac
         y, you have to decrese the
         #learning rate by 10%
         reduce_lr = ReduceLROnPlateau(monitor='val_accuracy', factor=0.9,
                                        patience=0, min_lr=0.001, verbose=1)
         model.compile(
           loss='categorical_crossentropy',
           optimizer=tf.keras.optimizers.Adam(learning_rate=0.00001, beta_1=0.9, beta_2
         =0.999, epsilon=1e-07, amsgrad=False),
           metrics=['accuracy',tf.keras.metrics.Precision(),tf.keras.metrics.Recall(),t
         fa.metrics.F1Score(num classes=2)]
```

```
In [21]: y_train =tf.keras.utils.to_categorical(train_data['dank_level'].values,2)
y_test =tf.keras.utils.to_categorical(test_data['dank_level'].values,2)
y_test.shape
```

Out[21]: (1719, 2)

3 - val_f1_score: 0.3839

Epoch 1/20

Epoch 2/20

```
Epoch 00002: LearningRateScheduler reducing learning rate to 9.99999974737875
2e-06.
racy: 0.5022 - precision: 0.5022 - recall: 0.5022 - f1_score: 0.4708 - val_lo
ss: 0.6923 - val_accuracy: 0.5265 - val_precision: 0.5265 - val_recall: 0.526
5 - val_f1_score: 0.4133
Epoch 3/20
Epoch 00003: LearningRateScheduler reducing learning rate to 9.49999976000981
3e-06.
racy: 0.5063 - precision: 0.5063 - recall: 0.5063 - f1 score: 0.4829 - val lo
ss: 0.6918 - val_accuracy: 0.5393 - val_precision: 0.5393 - val_recall: 0.539
3 - val f1 score: 0.4486
Epoch 4/20
Epoch 00004: LearningRateScheduler reducing learning rate to 9.49999957811087
4e-06.
racy: 0.4996 - precision: 0.4996 - recall: 0.4996 - f1 score: 0.4812 - val lo
ss: 0.6915 - val_accuracy: 0.5398 - val_precision: 0.5398 - val_recall: 0.539
8 - val f1 score: 0.4527
Epoch 5/20
Epoch 00005: LearningRateScheduler reducing learning rate to 9.49999957811087
4e-06.
racy: 0.5110 - precision: 0.5110 - recall: 0.5110 - f1 score: 0.5002 - val lo
ss: 0.6914 - val_accuracy: 0.5468 - val_precision: 0.5468 - val_recall: 0.546
8 - val_f1_score: 0.4760
Epoch 6/20
Epoch 00006: LearningRateScheduler reducing learning rate to 9.02499959920533
e-06.
racy: 0.5034 - precision: 0.5034 - recall: 0.5034 - f1_score: 0.4930 - val_lo
ss: 0.6911 - val_accuracy: 0.5439 - val_precision: 0.5439 - val_recall: 0.543
9 - val f1 score: 0.4750
Epoch 7/20
Epoch 00007: LearningRateScheduler reducing learning rate to 9.02499959920533
e-06.
racy: 0.5013 - precision: 0.5013 - recall: 0.5013 - f1_score: 0.4937 - val_lo
ss: 0.6909 - val_accuracy: 0.5474 - val_precision: 0.5474 - val_recall: 0.547
4 - val_f1_score: 0.4769
Epoch 8/20
```

Epoch 00001: LearningRateScheduler reducing learning rate to 9.99999974737875

racy: 0.4996 - precision: 0.4996 - recall: 0.4996 - f1_score: 0.4580 - val_lo ss: 0.6930 - val_accuracy: 0.5183 - val_precision: 0.5183 - val_recall: 0.518

```
Epoch 00008: LearningRateScheduler reducing learning rate to 9.02499959920533
e-06.
racy: 0.5010 - precision: 0.5010 - recall: 0.5010 - f1 score: 0.4932 - val lo
ss: 0.6907 - val_accuracy: 0.5521 - val_precision: 0.5521 - val_recall: 0.552
1 - val f1 score: 0.5214
Epoch 9/20
Epoch 00009: LearningRateScheduler reducing learning rate to 8.57374961924506
4e-06.
racy: 0.5075 - precision: 0.5075 - recall: 0.5075 - f1_score: 0.4987 - val_lo
ss: 0.6906 - val_accuracy: 0.5590 - val_precision: 0.5590 - val_recall: 0.559
0 - val_f1_score: 0.5403
Epoch 10/20
Epoch 00010: LearningRateScheduler reducing learning rate to 8.57374925544718
racy: 0.5075 - precision: 0.5075 - recall: 0.5075 - f1_score: 0.5015 - val_lo
ss: 0.6905 - val_accuracy: 0.5457 - val_precision: 0.5457 - val_recall: 0.545
7 - val f1 score: 0.5340
Epoch 11/20
Epoch 00011: LearningRateScheduler reducing learning rate to 8.57374925544718
3e-06.
racy: 0.4990 - precision: 0.4990 - recall: 0.4990 - f1 score: 0.4958 - val lo
ss: 0.6905 - val accuracy: 0.5480 - val precision: 0.5480 - val recall: 0.548
0 - val f1 score: 0.5398
Epoch 12/20
Epoch 00012: LearningRateScheduler reducing learning rate to 8.14506179267482
4e-06.
racy: 0.5095 - precision: 0.5095 - recall: 0.5095 - f1_score: 0.5067 - val_lo
ss: 0.6903 - val accuracy: 0.5445 - val precision: 0.5445 - val recall: 0.544
5 - val_f1_score: 0.5414
Epoch 13/20
Epoch 00013: LearningRateScheduler reducing learning rate to 8.14506165625061
8e-06.
racy: 0.5072 - precision: 0.5072 - recall: 0.5072 - f1_score: 0.5043 - val_lo
ss: 0.6903 - val_accuracy: 0.5439 - val_precision: 0.5439 - val_recall: 0.543
9 - val f1 score: 0.5408
Epoch 14/20
Epoch 00014: LearningRateScheduler reducing learning rate to 8.14506165625061
8e-06.
racy: 0.5189 - precision: 0.5189 - recall: 0.5189 - f1 score: 0.5154 - val lo
ss: 0.6901 - val_accuracy: 0.5445 - val_precision: 0.5445 - val_recall: 0.544
5 - val_f1_score: 0.5415
Epoch 15/20
```

```
Epoch 00015: LearningRateScheduler reducing learning rate to 7.73780857343808
7e-06.
racy: 0.5242 - precision: 0.5242 - recall: 0.5242 - f1_score: 0.5207 - val_lo
ss: 0.6901 - val_accuracy: 0.5404 - val_precision: 0.5404 - val_recall: 0.540
4 - val_f1_score: 0.5390
Epoch 16/20
Epoch 00016: LearningRateScheduler reducing learning rate to 7.73780811869073
7e-06.
racy: 0.5192 - precision: 0.5192 - recall: 0.5192 - f1_score: 0.5176 - val_lo
ss: 0.6900 - val_accuracy: 0.5398 - val_precision: 0.5398 - val_recall: 0.539
8 - val_f1_score: 0.5385
Epoch 17/20
Epoch 00017: LearningRateScheduler reducing learning rate to 7.73780811869073
7e-06.
racy: 0.5257 - precision: 0.5257 - recall: 0.5257 - f1 score: 0.5240 - val lo
ss: 0.6899 - val_accuracy: 0.5381 - val_precision: 0.5381 - val_recall: 0.538
1 - val_f1_score: 0.5366
Epoch 18/20
Epoch 00018: LearningRateScheduler reducing learning rate to 7.3509177127562e
racy: 0.5101 - precision: 0.5101 - recall: 0.5101 - f1 score: 0.5087 - val lo
ss: 0.6899 - val accuracy: 0.5393 - val precision: 0.5393 - val recall: 0.539
3 - val f1 score: 0.5378
Epoch 19/20
Epoch 00019: LearningRateScheduler reducing learning rate to 7.3509177127562e
racy: 0.5148 - precision: 0.5148 - recall: 0.5148 - f1 score: 0.5130 - val lo
ss: 0.6898 - val_accuracy: 0.5398 - val_precision: 0.5398 - val_recall: 0.539
8 - val f1 score: 0.5384
Epoch 20/20
Epoch 00020: LearningRateScheduler reducing learning rate to 7.3509177127562e
racy: 0.5392 - precision: 0.5392 - recall: 0.5392 - f1 score: 0.5374 - val lo
ss: 0.6897 - val_accuracy: 0.5393 - val_precision: 0.5393 - val_recall: 0.539
3 - val_f1_score: 0.5377
```

WARNING:tensorflow:No training configuration found in the save file, so the m odel was *not* compiled. Compile it manually.

```
In [29]: accuracy=accuracy_score(y_test,test_prediction)
    print("Test accuracy_score",accuracy)
    f1_test_score=f1_score(y_test,test_prediction)
    print("Test F1_score",f1_test_score)
    print("Test confusion matrix")
    cnf_matrix2=confusion_matrix(y_test,test_prediction)
    p = sns.heatmap(pd.DataFrame(cnf_matrix2), annot=True, cmap="YlGnBu" ,fmt='g')
    plt.title('Test confusion matrix', y=1.1)
    plt.ylabel('Actual label')
    plt.xlabel('Predicted label')
```

Test accuracy_score 0.5590459569517161
Test F1_score 0.44752186588921283
Test confusion matrix

Out[29]: Text(0.5, 15.0, 'Predicted label')



