

RelevanceDetector

April 25, 2021

```
[137]: import os
import sys
os.environ['TRANSFORMERS_CACHE'] = '/projectnb/cs505/mahirp/cache'
os.environ['PREFIX_PATH']="/projectnb/cs505/mahirp/python"
os.environ['TFHUB_CACHE_DIR']='/projectnb/cs505/mahirp/tfcache'
sys.path.append('/projectnb/cs505/mahirp/python/bin')
sys.path.append('/projectnb/cs505/mahirp/cache')
sys.path.append('/projectnb/cs505/mahirp/tfcache')
sys.path.append('/projectnb/cs505/mahirp/usr/bin')
```

```
[ ]: 'pip install --ignore-installed --prefix=$PREFIX_PATH transformers sklearn
```

```
[2]: from transformers import pipeline, set_seed
import re
import pandas as pd
import tensorflow_hub as hub
from datetime import datetime
from sklearn.metrics.pairwise import cosine_similarity
import random
import tensorflow as tf
import matplotlib.pyplot as plt
%matplotlib inline
tf.device('/GPU:0')
import numpy as np
import nltk
nltk.data.path.append("/projectnb/cs505/mahirp/nltk/")
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
loadFromFile=False

prepend=datetime.now().strftime("%m-%d-%Y-%H-%M-%S-")

stop_words = set(stopwords.words('english'))
```

```
[3]: useGlove=True
if useGlove:
    embeddings_index = {}
    f = open('glove.42B.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('Found %s word vectors.' % len(embeddings_index))

```

Found 1917494 word vectors.

```

[18]: prepend=datetime.now().strftime("%m-%d-%Y-%H-%M-%S-")
dataset = pd.read_excel('CS505-gunviolence/GVFC_AnnotatedHeadlines.xlsx',
    →index_col=0)
if not loadFromFile:
    true=dataset[dataset['V3relevance']==1]
    false=dataset[dataset['V3relevance']==0]
    trainDataset=true.sample(frac=0.7)
    testDataset=true.drop(trainDataset.index)
    ftrainDataset=false.sample(frac=0.7)
    ftestDataset=false.drop(ftrainDataset.index)
    trainDataset=trainDataset.append(ftrainDataset).sample(frac=1)
    testDataset=testDataset.append(ftestDataset).sample(frac=1)
    indices=np.array(trainDataset.index)
    np.save(open('train_indices.npy','wb'),indices)
else:
    print('loaded from storage')
    indices=np.load(open('train_indices.npy','rb'))
    trainDataset=dataset.iloc[indices]
    testDataset=dataset.drop(indices)
print(len(trainDataset))
print(len(testDataset))

```

910

390

```
[22]: len(testDataset)
```

[22]: 390

```

[4]: # embed = hub.load("https://tfhub.dev/google/universal-sentence-encoder-large/5")
embed = hub.load("https://tfhub.dev/google/universal-sentence-encoder/4")
generator = pipeline('text-generation', model='gpt2',device=0)
tokenizer=generator.tokenizer

```

INFO:absl:Using /projectnb/cs505/mahirp/tfcache to cache modules.

All model checkpoint layers were used when initializing TFGPT2LMHeadModel.

All the layers of TFGPT2LMHeadModel were initialized from the model checkpoint at gpt2.

If your task is similar to the task the model of the checkpoint was trained on, you can already use TFGPT2LMHeadModel for predictions without further training.

```
[5]: def generateDataset(dataset,isTest=False):
    df = pd.DataFrame(columns=['original','headline','generated','relevant'])
    c=0
    gen_seq=3 if not isTest else 1
    for original,headline,caption,relevant in zip(
        dataset['whole_text'],dataset['news_title'],dataset['generated_caption_of_lead_image'],dataset['relevant']):
        set_seed(random.randint(1,1000))
        print("{} / {}".format(c,len(dataset)))
        c+=1
        text=generator(caption.lower(), max_length=min(100,len(original)), num_return_sequences=gen_seq)
        headLineText=generator(headline.lower(), max_length=min(100,len(original)), num_return_sequences=gen_seq)
        for t,h in zip(text,headLineText):
            df=df.append({'original':original,'headline':headLineText,'generated':t,'relevant':h},ignore_index = True)
    return df
```

```
[6]: regexMap = { r"https?://[a-zA-Z0-9./]*": "", r"@[\w]*": " ", r"#[\w]*": "\n", r"\n": " ", r"\\": " ", r"[^a-zA-Z\s]": "", r"[s]+": " "}

def cleanData(text):
    text=text.strip()
    for regx in regexMap.keys():
        text = re.sub(regx, regexMap[regx], text)
    return text
```

```
[9]: def getEmbeddings(text):
    global useGlove
    text=text.lower()
    if not useGlove:
        print('\rusing USE',end='')
        return embed([text]).numpy().astype(np.float)
    else:
        print('\rusing GLOVE',end='')
        embedding_matrix = np.zeros((1,512),dtype=np.float)
        tokens=word_tokenize(text)
        tokens=[word for word in tokens if word not in stop_words]
        cnt=0
        for i,word in enumerate(tokens):
            embedding_vector = embeddings_index.get(word)
```

```

        if embedding_vector is not None:
            cnt+=1
            embedding_vector = np.pad(embedding_vector, (0,212), 'constant')
            embedding_matrix = embedding_matrix +
→((embedding_vector-embedding_matrix)/cnt)
        return embedding_matrix

```

```

[ ]: prepend=datetime.now().strftime("%m-%d-%Y-%H-%M-%S-")
# trainDF = pd.read_csv('CS505-gunviolence/
→04-25-2021-05-40-38-raw_train_relevanceDatasetV3.csv')
trainDF= generateDataset(trainDataset)
trainDF['generated']=list(map(cleanData,trainDF['generated']))
trainDF['headline']=list(map(cleanData,trainDF['headline']))
trainDF['original']=list(map(cleanData,trainDF['original']))
trainDF.to_csv('CS505-gunviolence/'+prepend+'raw_train_relevanceDatasetV3.
→csv',index=False)
useGlove=False
prefix='_GLOVE_' if useGlove else '_USE_'
trainDF['generated']=list(map(getEmbeddings,trainDF['generated']))
print('done')
trainDF['headline']=list(map(getEmbeddings,trainDF['headline']))
print('done')
trainDF['original']=list(map(getEmbeddings,trainDF['original']))

trainDF.to_csv('CS505-gunviolence/'+prepend+prefix+'train_relevanceDatasetV3.
→csv',index=False)
trainDF = pd.read_csv('CS505-gunviolence/'+prepend+'raw_train_relevanceDatasetV3.
→csv')
useGlove=True
prefix='_GLOVE_' if useGlove else '_USE_'
trainDF['generated']=list(map(getEmbeddings,trainDF['generated']))
print('done')
trainDF['headline']=list(map(getEmbeddings,trainDF['headline']))
print('done')
trainDF['original']=list(map(getEmbeddings,trainDF['original']))

trainDF.to_csv('CS505-gunviolence/'+prepend+prefix+'train_relevanceDatasetV3.
→csv',index=False)

```

```

[ ]: # testDF = pd.read_csv('CS505-gunviolence/
→04-25-2021-05-40-38-raw_test_relevanceDatasetV3.csv')
testDF=generateDataset(testDataset,True)
testDF['generated']=list(map(cleanData,testDF['generated']))
testDF['headline']=list(map(cleanData,testDF['headline']))
testDF['original']=list(map(cleanData,testDF['original']))
testDF.to_csv('CS505-gunviolence/'+prepend+'raw_test_relevanceDatasetV3.
→csv',index=False)

```

```

useGlove=False
prefix='_GLOVE_' if useGlove else '_USE_'
testDF['generated']=list(map(getEmbeddings,testDF['generated']))
print('done')
testDF['headline']=list(map(getEmbeddings,testDF['headline']))
print('done')
testDF['original']=list(map(getEmbeddings,testDF['original']))
testDF.to_csv('CS505-gunviolence/'+prepend+prefix+'test_relevanceDatasetV3.
→csv',index=False)
testDF = pd.read_csv('CS505-gunviolence/'+prepend+'raw_test_relevanceDatasetV3.
→csv')
useGlove=True
prefix='_GLOVE_' if useGlove else '_USE_'
testDF['generated']=list(map(getEmbeddings,testDF['generated']))
print('done')
testDF['headline']=list(map(getEmbeddings,testDF['headline']))
print('done')
testDF['original']=list(map(getEmbeddings,testDF['original']))
testDF.to_csv('CS505-gunviolence/'+prepend+prefix+'test_relevanceDatasetV3.
→csv',index=False)

```

```

[79]: def getEnsembleModel(embeddingDims=512):
    inp=tf.keras.Input(shape=(3,embeddingDims),name='embeddings')
    inp2=tf.keras.Input(shape=(3,embeddingDims),name='glove_embeddings')
    x=tf.keras.layers.Dense(256,activation='relu',kernel_regularizer=tf.keras.
→regularizers.l2(l2=8e-6),name='features01')(inp)
    x=tf.keras.layers.Flatten()(x)
    x=tf.keras.layers.Dense(64,activation='relu',kernel_regularizer=tf.keras.
→regularizers.l2(l2=8e-5),name='features')(x)
    x=tf.keras.layers.Dense(2,activation='softmax')(x)

    y=tf.keras.layers.
→Dense(256,activation='relu',name='features02',kernel_regularizer=tf.keras.
→regularizers.l2(l2=8e-6))(inp2)
    y=tf.keras.layers.Flatten()(y)
    y=tf.keras.layers.Dense(64,activation='relu',kernel_regularizer=tf.keras.
→regularizers.l2(l2=8e-5),name='features2')(y)
    y=tf.keras.layers.Dense(2,activation='softmax')(y)
    z=tf.keras.layers.Average()([x, y])
    return tf.keras.Model([inp, inp2],z)

def getModel(embeddingDims=512,isGlove=False):
    if not isGlove:
        inp=tf.keras.Input(shape=(3,embeddingDims),name='embeddings')
    else:
        inp=tf.keras.Input(shape=(3,embeddingDims),name='glove_embeddings')

```

```

        x=tf.keras.layers.Dense(256,activation='relu',kernel_regularizer=tf.keras.
→regularizers.l2(l2=2e-6),name='features0')(inp)
        x=tf.keras.layers.Flatten()(x)
        x=tf.keras.layers.Dense(64,activation='relu',kernel_regularizer=tf.keras.
→regularizers.l2(l2=8e-5),name='features')(x)
        x=tf.keras.layers.Dense(2,activation='softmax')(x)
        return tf.keras.Model(inp,x)

def getAutoEncoder():
    inp = tf.keras.Input(shape=(512))
    x=tf.keras.layers.Flatten()(inp)
    x=tf.keras.layers.Dense(128)(x)
    x=tf.keras.layers.Dense(512)(x)
    return tf.keras.Model(inp,x)

```

```

[113]: model=getModel()
        gloveModel=getModel(isGlove=True)
        ensembleModel=getEnsembleModel()

```

```

[105]: model.summary()
        gloveModel.summary()
        ensembleModel.summary()

```

Model: "functional_55"

Layer (type)	Output Shape	Param #
embeddings (InputLayer)	[(None, 3, 512)]	0
features0 (Dense)	(None, 3, 256)	131328
flatten_36 (Flatten)	(None, 768)	0
features (Dense)	(None, 64)	49216
dense_36 (Dense)	(None, 2)	130
Total params: 180,674		
Trainable params: 180,674		
Non-trainable params: 0		

Model: "functional_57"

Layer (type)	Output Shape	Param #
glove_embeddings (InputLayer	[(None, 3, 512)]	0

features0 (Dense)	(None, 3, 256)	131328

flatten_37 (Flatten)	(None, 768)	0

features (Dense)	(None, 64)	49216

dense_37 (Dense)	(None, 2)	130
=====		

Total params: 180,674
Trainable params: 180,674
Non-trainable params: 0

Model: "functional_59"

Layer (type)	Output Shape	Param #	Connected to
=====			
embeddings (InputLayer)	[(None, 3, 512)]	0	

glove_embeddings (InputLayer)	[(None, 3, 512)]	0	

features01 (Dense) embeddings[0][0]	(None, 3, 256)	131328	

features02 (Dense) glove_embeddings[0][0]	(None, 3, 256)	131328	

flatten_38 (Flatten) features01[0][0]	(None, 768)	0	

flatten_39 (Flatten) features02[0][0]	(None, 768)	0	

features (Dense) flatten_38[0][0]	(None, 64)	49216	

features2 (Dense) flatten_39[0][0]	(None, 64)	49216	

dense_38 (Dense)	(None, 2)	130	features[0][0]

dense_39 (Dense)	(None, 2)	130	features2[0][0]

average_9 (Average)	(None, 2)	0	dense_38[0][0] dense_39[0][0]
=====			
=====			
Total params: 361,348			
Trainable params: 361,348			
Non-trainable params: 0			


```
[19]: # df=pd.read_csv('CS505-gunviolence/'+prepend+'train_relevanceDatasetV3.csv')
# df = pd.read_csv('CS505-gunviolence/
→04-19-2021-04-43-45-train_relevanceDatasetV3.csv')
df = pd.read_csv('CS505-gunviolence/
→04-25-2021-04-41-28-_USE_train_relevanceDatasetV3.csv')
df2 = pd.read_csv('CS505-gunviolence/
→04-25-2021-04-41-28-_GLOVE_train_relevanceDatasetV3.csv')
labels=df.pop('relevant')
```

```
[20]: regexMap = { r"\n": " ", r"[\s]+": " ", r"\[,": "[" }
```

```
[119]: def generateNPDataset(dataframe):
features=[]
for index,data in dataframe.iterrows():
c1=np.array(eval(cleanData(data['headline']))),dtype=np.float)
c2=np.array(eval(cleanData(data['generated']))),dtype=np.float)
c3=np.array(eval(cleanData(data['original']))),dtype=np.float)
# features.append(c1.T@((c2+c3)/2))
features.append(np.vstack((c1,c2,c3)))
features=np.array(features,dtype=np.float)
return features
```

```
[22]: features=generateNPDataset(df)
```

```
[23]: glove_features=generateNPDataset(df2)
```

```
[24]: print(features.shape)
print(glove_features.shape)
```

```
(2730, 3, 512)
(2730, 3, 512)
```



```
[49]: dataset = tf.data.Dataset.from_tensor_slices(({ 'embeddings':
    ↳ features, 'glove_embeddings': glove_features}, tf.keras.utils.
    ↳ to_categorical(labels)))
```

```
[50]: l=len(dataset)
dataset=dataset.shuffle(l+1).batch(16)
l=len(dataset)
train_set=dataset.take(int(l*0.75))
test_set=dataset.skip(int(l*0.75))
val_set=test_set.take(int(l*0.75*0.10))
test_set=test_set.skip(int(l*0.75*0.10))
print(len(test_set))
print(len(train_set))
print(len(val_set))
```

```
31
128
12
```

```
[ ]: model.compile(optimizer=tf.keras.optimizers.Adam(lr=5e-5),
    loss=tf.keras.losses.CategoricalCrossentropy(from_logits=True),
    metrics=['accuracy'])
gloveModel.compile(optimizer=tf.keras.optimizers.Adam(lr=5e-5),
    loss=tf.keras.losses.CategoricalCrossentropy(from_logits=True),
    metrics=['accuracy'])
ensembleModel.compile(optimizer=tf.keras.optimizers.Adam(lr=5e-5),
    loss=tf.keras.losses.CategoricalCrossentropy(from_logits=True),
    metrics=['accuracy'])
history=model.fit(train_set,validation_data=val_set,epochs=8,shuffle=True)
history=ensembleModel.
    ↳ fit(train_set,validation_data=val_set,epochs=10,shuffle=True)
history=gloveModel.fit(train_set,validation_data=val_set,epochs=7,shuffle=True)
```

```
[107]: model.evaluate(test_set)
gloveModel.evaluate(test_set)
ensembleModel.evaluate(test_set)
```

```
31/31 [=====] - 0s 1ms/step - loss: 0.6212 - accuracy:
0.7286
31/31 [=====] - 0s 1ms/step - loss: 0.6629 - accuracy:
0.6429
31/31 [=====] - 0s 1ms/step - loss: 0.6507 - accuracy:
0.6837
```

```
[107]: [0.6507479548454285, 0.6836734414100647]
```

```
[118]: model.save(prepend+'final_USE.h5')
gloveModel.save(prepend+'final_GLOVE.h5')
```

```
ensembleModel.save(prepend+'final_ENSEMBLE.h5')
```

```
[29]: results={}
sampleCount=0
testDF = pd.read_csv('CS505-gunviolence/
→04-25-2021-04-41-28-_USE_test_relevanceDatasetV3.csv')
testDF2= pd.read_csv('CS505-gunviolence/
→04-25-2021-04-41-28-_GLOVE_test_relevanceDatasetV3.csv')
tfeatures=[]
tlabels=tf.keras.utils.to_categorical(testDF.pop('relevant'))
tfeatures=generateNPDataset(testDF)
tfeatures_glove=generateNPDataset(testDF2)
```

```
[30]: print(tfeatures.shape)
test_dataset = tf.data.Dataset.from_tensor_slices(({'embeddings':
→tfeatures, 'glove_embeddings':tfeatures_glove}, tlabels)).batch(32)
```

(390, 3, 512)

```
[122]: # test_input=np.array(test_input,dtype=np.float)
# test_label=np.array(test_label)
model.evaluate(test_dataset)
gloveModel.evaluate(test_dataset)
ensembleModel.evaluate(test_dataset)
```

13/13 [=====] - 0s 1ms/step - loss: 0.6641 - accuracy: 0.6462

13/13 [=====] - 0s 1ms/step - loss: 0.6707 - accuracy: 0.6051

13/13 [=====] - 0s 2ms/step - loss: 0.6662 - accuracy: 0.6385

[122]: [0.666168212890625, 0.6384615302085876]

```
[120]: from sklearn.manifold import TSNE
def getIntermediate(model, layers):
    x=None
    for layer in layers:
        if x is None:
            x=model.get_layer(layer).output
        else:
            x=model.get_layer(layer)(x)
    return tf.keras.models.Model(inputs=model.input, outputs=x)

def getTSNEData(model, listOfLayers):
    intermediate_layer_model = getIntermediate(model, listOfLayers)
    predict = intermediate_layer_model.predict(test_dataset)
```

```
tsne = TSNE(n_components=2, verbose=1,
→random_state=42,n_jobs=16,method='exact')
return tsne.fit_transform(predict)
```

```
[128]: import matplotlib.pyplot as plt
%matplotlib inline
def plotTSNE(model,listOfLayers=['features']):
    z=getTSNEData(model,listOfLayers)
    classes = [1,0]
    label=['Relevant','Not Relevant']
    tsnedataset={}
    for c in classes:
        tsnedataset[c]=np.where(tlabels[:,0]==c)
    colors=['#e76f51','#2a9d8f','#1d3557','#a8dad9']
    for c in classes:
        plt.
→scatter(z[tsnedataset[c],0],z[tsnedataset[c],1],c=colors[c],label=label[c])
        plt.legend(loc='best')
        plt.title(model.name)
        plt.show()
```

```
[132]: model._name='USE_Based'
plotTSNE(model)
```

/share/pkg.7/tensorflow/2.3.1/install/lib/SCC/./python3.7/site-packages/tensorflow/python/keras/engine/functional.py:543: UserWarning: Input dict contained keys ['glove_embeddings'] which did not match any model input. They will be ignored by the model.

```
[n for n in tensors.keys() if n not in ref_input_names])
```

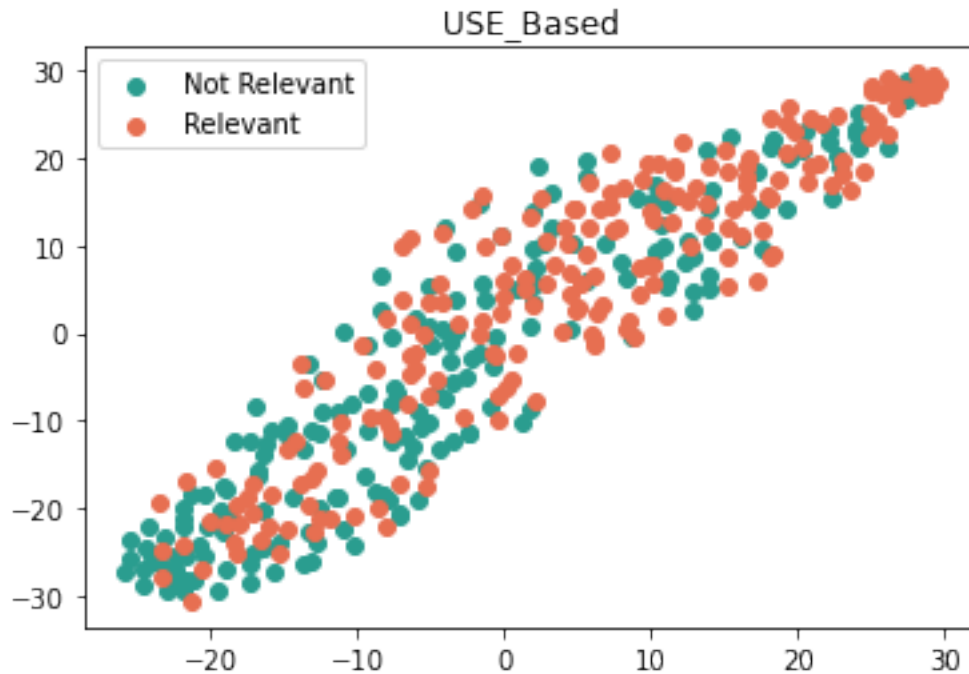
[t-SNE] Computing pairwise distances...

[t-SNE] Computed conditional probabilities for sample 390 / 390

[t-SNE] Mean sigma: 0.141848

[t-SNE] KL divergence after 250 iterations with early exaggeration: 64.503215

[t-SNE] KL divergence after 1000 iterations: 0.807066



```
[133]: gloveModel._name='GLOVE_Based'
plotTSNE(gloveModel)
```

```
/share/pkg.7/tensorflow/2.3.1/install/lib/SCC/./python3.7/site-
packages/tensorflow/python/keras/engine/functional.py:543: UserWarning: Input
dict contained keys ['embeddings'] which did not match any model input. They
will be ignored by the model.
```

```
[n for n in tensors.keys() if n not in ref_input_names])
```

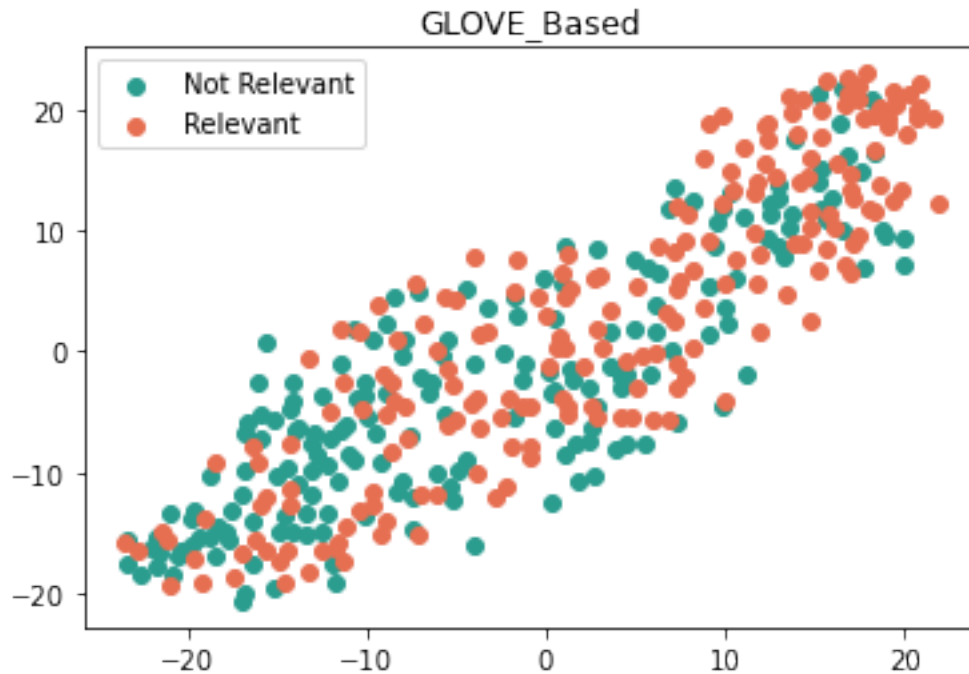
```
[t-SNE] Computing pairwise distances...
```

```
[t-SNE] Computed conditional probabilities for sample 390 / 390
```

```
[t-SNE] Mean sigma: 0.179231
```

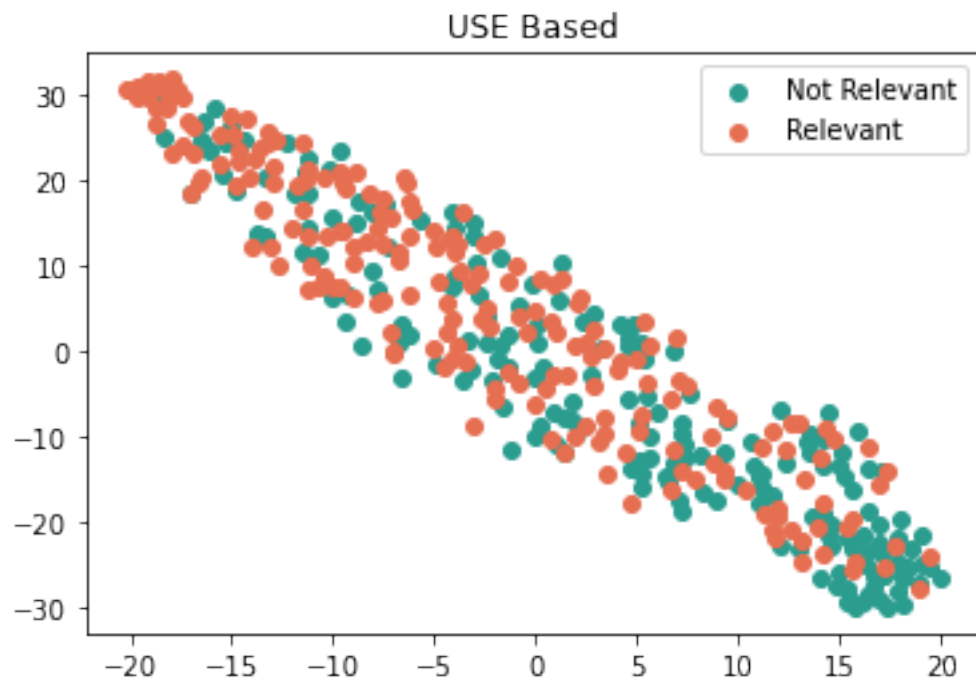
```
[t-SNE] KL divergence after 250 iterations with early exaggeration: 71.086230
```

```
[t-SNE] KL divergence after 1000 iterations: 0.913512
```



```
[134]: ensembleModel._name='USE Based'  
plotTSNE(ensembleModel)  
ensembleModel._name='GLOVE Based'  
plotTSNE(ensembleModel,['features2'])
```

```
[t-SNE] Computing pairwise distances...  
[t-SNE] Computed conditional probabilities for sample 390 / 390  
[t-SNE] Mean sigma: 0.160090  
[t-SNE] KL divergence after 250 iterations with early exaggeration: 64.408855  
[t-SNE] KL divergence after 1000 iterations: 0.698121
```



```
[t-SNE] Computing pairwise distances...  
[t-SNE] Computed conditional probabilities for sample 390 / 390  
[t-SNE] Mean sigma: 0.195384  
[t-SNE] KL divergence after 250 iterations with early exaggeration: 66.365957  
[t-SNE] KL divergence after 1000 iterations: 0.926434
```

