

# **Configuration Manual**

MSc Research Project MSCTOPAI

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#### National College of Ireland

#### **MSc Project Submission Sheet**



#### **School of Computing**

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Programme:	MSCAITOP	Year:	2024		
Module:	MSCAITOP				
Lecturer: Submission Due	Professor Sheresh Zahoor				
Date:	12/8/2024				
Project Title:	Co-pilot widget for assisting the public in processing US presidential political candidate tweets from Twitter in 2024 US elections candidate choice through sarcasm and stance detection				

#### Word Count: 608 Page Count: 72

I hereby certify that the information contained in this (my submission) is information pertaining to research I conducted for this project. All information other than my own contribution will be fully referenced and listed in the relevant bibliography section at the rear of the project.

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**Date:** 11/8/2024

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## Configuration Manual

Arthur Ryan Student ID: 23333138

## **1** Introduction

This document summarises the specification of the software and hardware for the project 'Co-pilot widget for assisting the public in processing US presidential political candidate tweets from Twitter in 2024 US elections candidate choice through sarcasm and stance detection'

## 2 Environment

Python 3.0

## **3** Libraries

The following libraries are required for running the project code notebooks.

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import os
!pip install gensim # Gensim is an open-source library for unsupervised
topic modeling and natural language processing
import nltk
nltk.download('punkt')
import pandas as pd
```

import numpy as np import matplotlib.pyplot as plt import seaborn as sns from wordcloud import WordCloud, STOPWORDS import nltk import re from nltk.corpus import stopwords #!pip install nltk import nltk nltk.download("stopwords") import seaborn as sns import gensim from gensim.utils import simple\_preprocess from gensim.parsing.preprocessing import STOPWORDS

```
import plotly.express as px
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import roc_auc_score
from sklearn.metrics import confusion_matrix
from sklearn.metrics import f1_score
from sklearn.metrics import precision score
```

For latency measurements:

import time

#### For Sarcasm and Stance:

import keras
#import tensorflow as tf

from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad\_sequences
from tensorflow.keras.layers import LSTM, GRU, Dense, Embedding,
Dropout, GlobalAveragePooling1D, Flatten, SpatialDropout1D,
Bidirectional
from keras.models import Sequential
from keras.metrics import F1Score

#### For Charting:

import io import xml.etree.ElementTree as ET import time

from matplotlib.lines import Line2D
from matplotlib.markers import MarkerStyle
from matplotlib.transforms import Affine2D

## **4** Software Specification

Google Colab: a web environment that uses Google Cloud Gmail account: to access Google Colab

## 5 Hardware Specification

Hardware: Hewlard Packard (HP) Z440 Memory: 32GB RAM Storage: 1TB SSD Processor: 4 CPU each of 1.6 Ghz

## 6 Code Section

## 6.0 Code sources

#### Fake News

https://www.kaggle.com/code/paramarthasengupta/fake-news-detector-edaprediction-99

#### Sarcasm

https://www.kaggle.com/datasets/deepnews/fakenews-reddit-comments/data

#### Stance

https://www.kaggle.com/datasets/arashnic/7-nlp-tasks-with-tweets

### 6.1 Truth-Fake News detection

#### Data Cleaning

- · use of gensim library to preprocess data
- screen out stop words
- screen out words less than length 2



[ ] 1 # Transforming the unmatching subjects to the same notation 2 df.subject=df.subject.replace({'politics':'PoliticsNews','politicsNews';'PoliticsNews'})

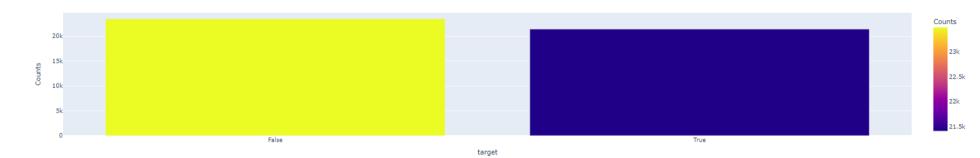
Additional data cleaning: transforming generic references to proper nouns

#### Exploratory Data Analysis (EDA)

#### Histogram for balance of true v fake news



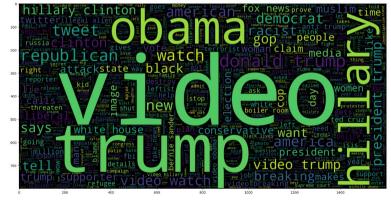
₹



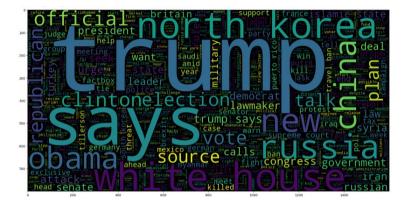
#### Historgam of subjects covered



#### Below is Word Cloud of True News words



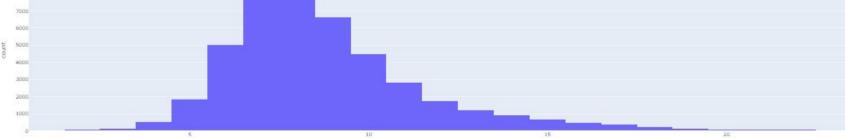
Below is Word Cloud of Fake News words



To understand dimensions of data being examined a graph of the title length was made

Measuring length of Titles





#### **Prediction with TF-iDF + Logistic Regression (only using title)**

#### Using only the Title to predict

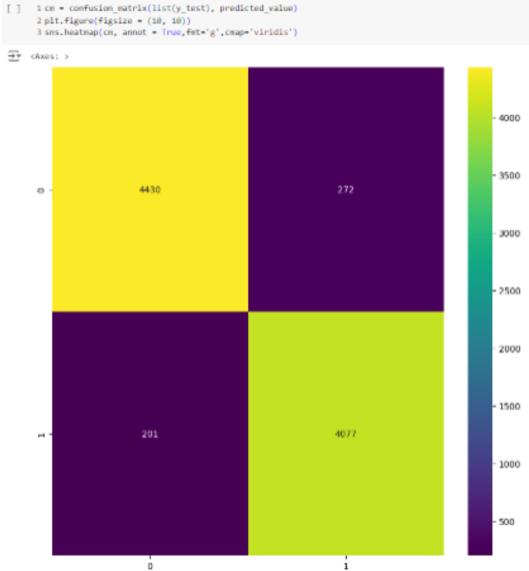
1st Prediction - based on Title(i.e., max 34 words)

[ ] 1 X\_train, X\_test, y\_train, y\_test = train\_test\_split(df.clean\_joined\_title, df.target, test\_size = 0.2,random\_state=2)
2 vec\_train = CountVectorizer().fit(X\_train)
3 X\_vec\_train = vec\_train.transform(X\_train)
4 X\_vec\_test = vec\_train.transform(X\_test)

[ ] 1 model = LogisticRegression(C=2) 2 model.fit(X\_vec\_train, y\_train) 3 predicted\_value = model.predict(X\_vec\_test) 4 accuracy\_value = roc\_auc\_score(y\_test, predicted\_value) 5 print("Accuracy score = ", accuracy\_value) 6 f1\_score\_value = f1\_score(y\_test, predicted\_value, average='macro') 7 print("F1\_score = ", f1\_score\_value) 8 precision\_score\_value = precision\_score(y\_test, predicted\_value, average='macro') 9 print("Precison score = ", precision\_score\_value)

→ Accuracy score = 0.9475838516986892 F1\_score = 0.9472458762778595 Precison score = 0.9470268669868982 /usr/local/lib/python3.10/dist-packages/sklearn/linear\_model/\_logistic.py:460: ConvergenceWarning:

## Ist Confusion Matrix

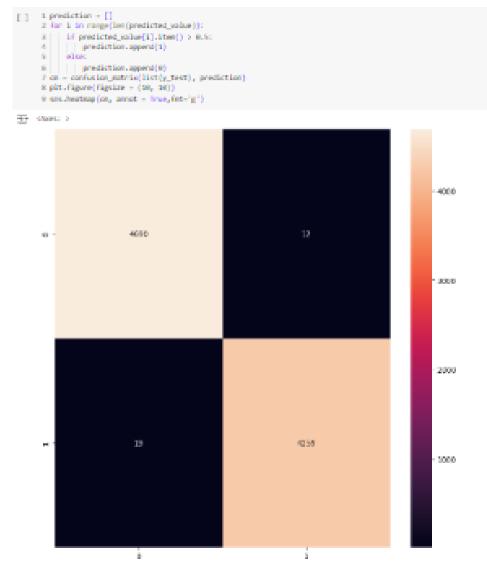


#### 2nd Prediction based on Text (i.e., more words than just title)

```
1 #prediction based on Text 100-200 words vs only Title 8 words max
    2
    3 X_train, X_test, y_train, y_test = train_test_split(df.clean_joined_text, df.target, test_size = 0.2,random_state=2)
    4 vec_train = CountVectorizer().fit(X_train)
    5 X_vec_train = vec_train.transform(X_train)
    6 X_vec_test = vec_train.transform(X_test)
    7 model = LogisticRegression(C=2.5)
    8 model.fit(X_vec_train, y_train)
    9 predicted_value = model.predict(X_vec_test)
    10 accuracy_value = roc_auc_score(y_test, predicted_value)
    11 print(accuracy_value)
    12 print("Accuracy score = ", accuracy_value)
    13 f1_score_value = f1_score(y_test, predicted_value, average='macro')
    14 print("F1_score = ", f1_score_value)
    15 precision_score_value = precision_score(y_test, predicted_value, average='macro')
    16 print("Precison score = ", precision_score_value)
```

0.996503283394869
 Accuracy score = 0.996503283394869
 F1\_score = 0.9965399136276033
 Precison score = 0.996577763310009

#### - 2nd The Confusion Matrix



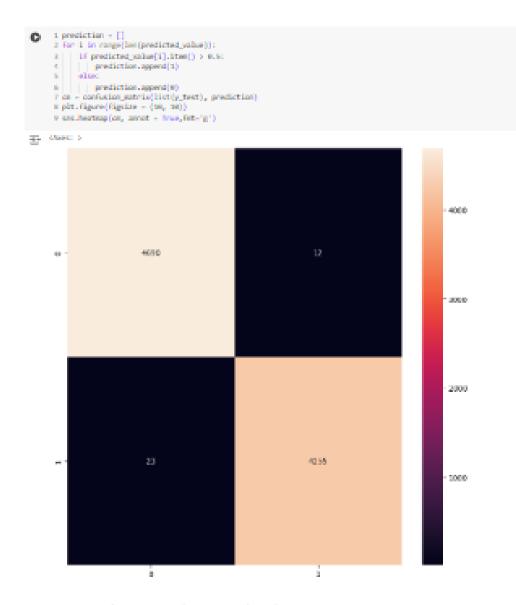
#### 3rd - Prediction - Title and Text combined

[ ] 1 df['clean\_final'] = df['original'].apply(preprocess) 2 df['clean\_joined\_final']=df['clean\_final'].apply(lambda x:" ".join(x))

#### 1 #prediction based on Text 100-200 words vs only Title 8 words max

2
3 X\_train, X\_test, y\_train, y\_test = train\_test\_split(df.clean\_joined\_final, df.target, test\_size = 0.2,random\_state=2)
4 vec\_train = CountVectorizer().fit(X\_train)
5 X\_vec\_train = vec\_train.transform(X\_train)
6 X\_vec\_test = vec\_train.transform(X\_test)
7 model = LogisticRegression(C=2.5)
8 model.fit(X\_vec\_train, y\_train)
9 predicted\_value = model.predict(X\_vec\_test)
10 accuracy\_value = noc\_alor\_score(y\_test, predicted\_value)
11 print(accuracy\_value)
12 print("Accuracy score = ", accuracy\_value)
13 f1\_score\_value = f1\_score(y\_test, predicted\_value, average='macro')
14 print("F1\_score = ", f1\_score(y\_test, predicted\_value, average='macro')
15 precision\_score\_value = precision\_score\_value)

0.996035775213476 Accuracy score = 0.996035775213476 F1\_score = 0.9960932826649207 Precison score = 0.9961538004445567



Above establishes baseline from historic tweet data

## Build out data i.e., record predictions to feature columns in 'memory' / csv file to be passed to next metric notebook

#### Build out dataset

```
[] 1 # Build out Dataset - make predictions and assign to target column in dataframe
      2
      3 import numpy as np
      4 np.set_printoptions(threshold=np.inf)
     6 #X_train, X_test, y_train, y_test = train_test_split(df.clean_joined_final, df.target, test_size = 0.2,random_state=0)
     7 #X_train, X_test, y_train, y_test = train_test_split(df.clean_joined_text, df.target, test_size = 0.2,random_state=0)
     9 #vec_train = CountVectorizer().fit(X_train)
    10 #X_vec_train = vec_train.transform(X_train)
    11 #X_vec_test = vec_train.transform(X_test)
    12
    13 # ->> put in new tweets here ->>X_vec_test = vec_train.transform(X_test)
    14
    15 #model = LogisticRegression(C=3)
    16 #model.fit(X_vec_train, y_train)
    17
    18 # Vectorising Text column (vs Title or Title and Text) to input to model previously traing on the static data inorder to make predictions on the new X data
    19 X_vec_train = vec_train.transform(df_new['clean_joined_text'])
    20 predicted_value = model.predict(X_vec_test) # it is these predicted values that are to have the graphics to be created from
    21 #print(predicted_value.reshape(-1))
    22
    23 df_new['predicted_true_fake'] = 0
     2.4
```

```
23 df_new['predicted_true_fake'] = 0
24
25 # for i in range(len(df_new)):
26 # if predicted_value[i] == 1:
27 # #df_new['target'].iloc[i] = 1
28 # df_new.loc[i, 'predicted_true_fake'] = 1
29 # # elif predicted_value.iloc[i] == 0:
30 # # df_new['Fake'] = 0
31
32 i = 0
33 for p in predicted_value:
34 if p == 1:
35 #df_new['target'].iloc[i] = 1
36 df_new.loc[i, 'predicted_true_fake'] = 1
37 i += 1
38 # elif predicted_value.iloc[i] == 0:
39 # df_new['Fake'] = 0
40
41
42 print(df_new['predicted_true_fake'])
43
44 #df_new.info()
45 #df_new = df_new[0,1,2,3,4]
46 #df2 = df1[['A', 'C']].copy()
47 df_new = df_new[['id', 'content', 'clean_text', 'clean_joined_text', 'predicted_true_fake']].copy()
48 #df.drop(df.columns[[5:]], axis=1, inplace=True)
49 df_new.info()
50
51 # ADD section for EDA again here
52
53 # NOT relevant as not in training mode but prediction # accuracy_value = roc_auc_score(y_test, predicted_value)
54 # NOT relevant as not in training mode but prediction # print(accuracy_value)
```

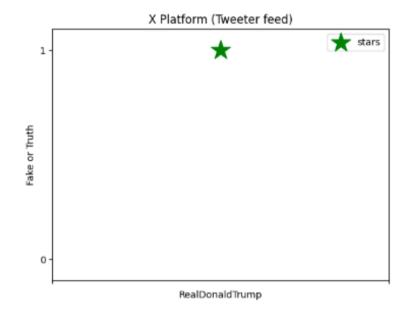


## Sample graphic to represent prediction of Fake v True Tweet

```
1 import matplotlib.pyplot as plt
С
     2 from matplotlib.ticker import StrMethodFormatter
     3
     4 # x-axis values
     5 x = [0] \# [1,2,3,4,5,6,7,8,9,10]
     6 # y-axis values
     7 y = df_new['predicted_true_fake'].iloc[0] # [2,4,5,7,6,8,9,11,12,12]
     8
     9
    10 if y == 0:
    11 # plotting points as a scatter plot
    12 plt.scatter(x, y, label= "rasberry", color= "red",
    13 marker= "X", s=450)
    14
    15 if y == 1:
    16 # plotting points as a scatter plot
    17 plt.scatter(x, y, label= "stars", color= "green",
    18 marker= "*", s=450)
    19
    20
    21 # x-axis label
    22 plt.xlabel('RealDonaldTrump')
    23 # frequency label
    24 plt.ylabel('Fake or Truth')
    25 # plot title
    26 plt.title('X Platform (Tweeter feed)')
    27 # showing legend
    28 plt.legend()
    29 # Setting the axis range
    30 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
    31 plt.ylim(-0.1, 1.1)
    32 plt.gca().set_xticklabels([])
    33
    34 #plt.set_yticklabels([str(round(float(label), 0)) for target in y])
```

```
26 plt.title('X Platform (Tweeter feed)')
27 # showing legend
28 plt.legend()
29 # Setting the axis range
30 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
31 plt.ylim(-0.1, 1.1)
32 plt.gca().set_xticklabels([])
33
34 #plt.set_yticklabels([str(round(float(label), 0)) for target in y])
35 plt.gca().yaxis.set_major_formatter(StrMethodFormatter('{x:,.0f}')) # No decimal
36 plt.xticks[[-1, 1])
37 plt.yticks[[0, 1])
38
39 # function to show the plot
40 plt.show()
```

Attempting to set identical low and high xlims makes transformation singular; automatically expanding.



Create memory for metrics i.e., save a file and add feature 1, 2(in next notebook), 3(in next notebook after notebook 2)

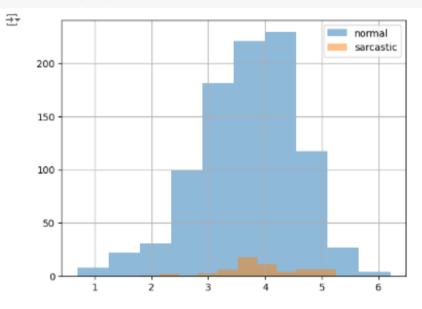
```
1 # save prediction values to csv file to be shared with other metrics to follow i.e., stance and mis/disinformation
     2
     3 df new.to csv('/content/drive/MyDrive/Colab Notebooks/MSCAITOP/data to graph.csv', encoding="utf-8")
     4
     5 # #df_new.save_csv("data_to_graph")
     6
     7 # import csv
     8
     9 # # field names
    10 # fields = ['Author', 'Text', 'clean_text', 'clean_joined_text', 'target'] # ['Name', 'Branch', 'Year', 'CGPA']
    11
    12 # # data rows of csv file
    13 # rows = df_new # [ ['Nikhil', 'COE', '2', '9.0'],
    14 # ['Sanchit', 'COE', '2', '9.1'],
    15 #
          # ['Aditya', 'IT', '2', '9.3'],
    16 #
          # ['Sagar', 'SE', '1', '9.5'],
    17 #
               # ['Prateek', 'MCE', '3', '7.8'],
                # ['Sahil', 'EP', '2', '9.1']]
    18 #
    19
    20 # # name of csv file
    21 # filename = "/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv"
    22
    23 # # writing to csv file
    24 # with open(filename, 'w') as csvfile:
    25 # # creating a csv writer object
    26 # csvwriter = csv.writer(csvfile)
    27
    28 # # writing the fields
    29 # csvwriter.writerow(fields)
    30
    31 # # writing the data rows
    32 #
            csvwriter.writerows(rows
    33
    34
    35 # df_new.info()
    36 # df_new.predicted_true_fake
```

## 6.2 Sarcasm

Exploratory data analysis (EDA)

```
[ ] 1 # histograph of sarcastic vs normal
2
```

```
3 train_df.loc[train_df['label'] == 0, 'comment'].str.len().apply(np.log1p).hist(label='normal', alpha=.5)
4 train_df.loc[train_df['label'] == 1, 'comment'].str.len().apply(np.log1p).hist(label='sarcastic', alpha=.5)
5 plt.legend();
```



#### Analysis of Data

#### All reddit comments of any size

1 sub\_df = train\_df.groupby('subreddit')['label'].agg([np.size, np.mean, np.sum, np.median, np.std, np.min, np.max])
2 sub\_df.sort\_values(by='sum', ascending=False).head(10)

	size	mean	sum	median	std	min	max
subreddit							
politics	68	0.102941	7	0.0	0.306141	0	1
The_Donald	50	0.100000	5	0.0	0.303048	0	1
leagueoflegends	14	0.285714	4	0.0	0.468807	0	1
wow	5	0.400000	2	0.0	0.547723	0	1
nfi	24	0.083333	2	0.0	0.282330	0	1
oddlysatisfying	3	0.888887	2	1.0	0.577350	0	1
nba	8	0.250000	2	0.0	0.482910	0	1
AskReddit	100	0.020000	2	0.0	0.140705	0	1
CFBOffTopic	1	1.000000	1	1.0	NaN	1	1
supergirlTV	1	1.000000	1	1.0	NaN	1	1

Create pipeline and train the TF-iDF + Logistic Regression model

#### Training the model

```
1 # build bigrams, put a limit on maximal number of features
2 # and minimal word frequency
3 tf_idf = TfidfVectorizer(ngram_range=(1, 2), max_features=50000, min_df=2)
4 # multinomial logistic regression a.k.a softmax classifier
5 logit = LogisticRegression(C=1, n_jobs=4, solver='lbfgs',
6 | | | | | | | | | | random_state=42, verbose=1)
7 # sklearn's pipeline
8 tfidf_logit_pipeline = Pipeline([('tf_idf', tf_idf),
9 | | | | | | | | | | | | | ('logit', logit)])
```

- 1 %%time
  2 tfidf\_logit\_pipeline.fit(train\_texts, y\_train)
- [Parallel(n\_jobs=4)]: Using backend LokyBackend with 4 concurrent workers. CPU times: user 115 ms, sys: 113 ms, total: 228 ms

Wall time: 2.34 s

 Pipeline

 TfidfVectorizer

 LogisticRegression

Visualisations Word Clouds



Prediction score from TF-iDF + Logistic Regression model

#### Reference base line Saracasm accuracy

[ ] 1 accuracy\_value = accuracy\_score(y\_valid, valid\_pred)
2 print("Accuracy score = ", accuracy\_value)
3 f1\_score\_value = f1\_score(y\_valid, valid\_pred, average='macro')
4 print("F1\_score = ", f1\_score\_value)
5 precision\_score\_value = precision\_score(y\_valid, valid\_pred, average='macro')
6 print("Precison score = ", precision\_score\_value)

```
Accuracy score = 0.932
F1_score = 0.48240165631469983
Precison score = 0.466
```

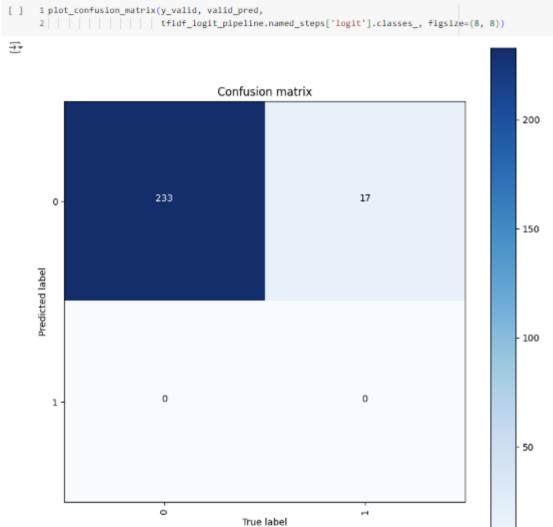
#### Point TF-iDF + Logistic Regression model trained model at new data

Predict Sarcasm with Logistic regression based on new X(twitter) texts

#### Confusion Matrix

```
[ ] 1 def plot_confusion_matrix(actual, predicted, classes,
     2
                                normalize=False,
     3
                                title='Confusion matrix', figsize=(7,7),
     4
                                cmap=plt.cm.Blues, path_to_save_fig=None):
           .....
     5
           This function prints and plots the confusion matrix.
     6
     7
           Normalization can be applied by setting 'normalize=True'.
           .....
     8
           import itertools
     9
    10
           from sklearn.metrics import confusion matrix
    11
           cm = confusion_matrix(actual, predicted).T
    12
           if normalize:
    13
          cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
    14
    15
           plt.figure(figsize=figsize)
           plt.imshow(cm, interpolation='nearest', cmap=cmap)
    16
    17
           plt.title(title)
    18
           plt.colorbar()
    19
           tick_marks = np.arange(len(classes))
    20
           plt.xticks(tick marks, classes, rotation=90)
    21
           plt.yticks(tick_marks, classes)
    22
    23
           fmt = '.2f' if normalize else 'd'
    24
           thresh = cm.max() / 2.
    25
           for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
    26
             plt.text(j, i, format(cm[i, j], fmt),
    27
                       horizontalalignment="center",
    28
                       color="white" if cm[i, j] > thresh else "black")
    29
    30
           plt.tight_layout()
    31
           plt.ylabel('Predicted label')
    32
           plt.xlabel('True label')
    33
    34
           if path_to_save_fig:
    35
           plt.savefig(path_to_save_fig, dpi=300, bbox_inches='tight')
```

Sarcasm - Confusion Matrix

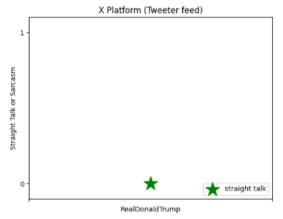


÷

Sample graphic to represent prediction of Sarcasm or Not/Straight Talk

```
[ ] 1 # TO BE POINTED AT THE NEW DATA #
     2 # TO BE POINTED AT THE NEW DATA #
     3 # TO BE POINTED AT THE NEW DATA # - originally form Truth Fake module
     4
     5 import matplotlib.pyplot as plt
     6 from matplotlib.ticker import StrMethodFormatter
     7
     8 # x-axis values
     9 x = [0] # [1,2,3,4,5,6,7,8,9,10]
    10 # y-axis values
    11 y = df_new_texts['predicted_sarcasm'].iloc[0] # [2,4,5,7,6,8,9,11,12,12]
    12
    13
    14 if y == 1:
    15 # plotting points as a scatter plot
    16 plt.scatter(x, y, label= "rasberry talk", color= "red",
    17 marker= "X", s=450)
    18 plt.legend(loc='upper right')
    19
    20 if y == 0:
    21 # plotting points as a scatter plot
    22 plt.scatter(x, y, label= "straight talk", color= "green",
    23 marker= "*", s=450)
    24 plt.legend(loc='lower right')
    25
    26
    27 # x-axis label
    28 plt.xlabel('RealDonaldTrump')
    29 # frequency label
    30 plt.ylabel('Straight Talk or Sarcasm')
31 # plot title
```





Improving the prediction performance: Build and use a LSTM model

## Model Improvement above tfidf+logit regression 70% accurate -> trying here LSTM to

get higher accuracy

 $\sim$ 

[]	1 # LSTM
	2
	3 # Text pre-processing
	4 # import tensorflow as tf
	5 # from tensorflow.keras.preprocessing.text import Tokenizer
	6 # from tensorflow.keras.preprocessing.sequence import pad_sequences
	7 # from tensorflow.keras.callbacks import EarlyStopping
	8 # # Modeling
	9 # from tensorflow.keras.models import Sequential
	10 # from tensorflow.keras.layers import LSTM, GRU, Dense, Embedding, Dropout, GlobalAveragePooling1D, Flatten, SpatialDropout1D, Bidirectional
	11
	12
	13 # #!pip install keras
	14 # import keras.preprocessing.text Tokenizer
	15 # from keras.preprocessing.sequence import pad_sequences
	16 # from keras.layers import Embedding, LSTM, Dense
	17 # from keras.models import Sequential
	18
	19 import keras
	20 #import tensorflow as tf
	21 from tensorflow.keras.preprocessing.text import Tokenizer
	22 from tensorflow.keras.preprocessing.sequence import pad_sequences
	23 from tensorflow.keras.layers import LSTM, GRU, Dense, Embedding, Dropout, GlobalAveragePooling1D, Flatten, SpatialDropout1D, Bidirectional
	24 from keras.models import Sequential
	25
	26
	27 # train_texts, valid_texts, y_train, y_valid = \
	<pre>28 # train test_split(train_df['comment'], train_df['label'], random_state=42)</pre>

```
26
    27 # train_texts, valid_texts, y_train, y_valid = \
    28 #
                train test split(train df['comment'], train df['label'], random state=42)
    29
    30 # The input text, example could be list of sentences
    31 texts = train_texts[:] # [...]
    32
    33 # The labels corresponding to the input text
    34 labels = y_train[:] # train_df['label'] # train_texts['Author'] # [...]
    35
    36 # Hyperparameters
    37 max_words = 10000 # max number of words to use in the vocabulary
    38 max len = 200 # max length of each text (in terms of number of words)
    39 embedding_dim = 200 # dimension of word embeddings
    40 lstm units = 64 # number of units in the LSTM layer
    41 num classes = len(set(labels)) # 2 # len(y train) # y train.unique() # 2 # len(set(labels)) # number of classes
    42
    43 # Tokenize the texts and create a vocabulary
    44 tokenizer = Tokenizer(num_words=max_words)
    45 tokenizer.fit on texts(texts.values.astype('U'))
    46 sequences = tokenizer.texts_to_sequences(texts.values.astype('U'))
    47
    48 # Pad the sequences so they all have the same length
    49 x = pad_sequences(sequences, maxlen=max_len)
    50
    51 # Create one-hot encoded labels
    52 y = keras.utils.to_categorical(labels,
    53 num_classes)
    54
    55 # Build the model
    56 model = Sequential()
    57 model.add(Embedding(max_words, embedding_dim, input_length=max_len))
    58 model.add(LSTM(1stm units))
    59 model.add(Dense(num_classes, activation='softmax')) # 'relu'' ; "rmsprop"
    60
    61 # optimizer='adam', loss='mse'
    62
    63
    64 # Compile the model
    65 model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy']) # 'binary_crossentropy'
```

# The LSTM model is trained over 10 epochs (shorten quantity of data to be reproducible in a code demonstration i.e., full dataset takes 10 mintues just to do this section)



//usr/local/lib/python3.10/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument 'input\_length' is deprecated. Just remove it. warnings.warn(

Epoch 1/10 24/24 --------- 9s 225ms/step - accuracy: 0.8997 - loss: 0.4441 Epoch 2/10 24/24 -----12s 307ms/step - accuracy: 0.9435 - loss: 0.2157 Epoch 3/10 24/24 --------- 10s 303ms/step - accuracy: 0.9414 - loss: 0.2068 Epoch 4/10 24/24 --------- 7s 292ms/step - accuracy: 0.9411 - loss: 0.1798 Epoch 5/10 24/24 ---------- 5s 225ms/step - accuracy: 0.9645 - loss: 0.0946 Epoch 6/10 24/24 ----------- 10s 440ms/step - accuracy: 0.9825 - loss: 0.0444 Epoch 7/10 24/24 ----------- 13s 533ms/step - accuracy: 0.9960 - loss: 0.0230 Epoch 8/10 24/24 ----------- 14s 262ms/step - accuracy: 0.9976 - loss: 0.0091 Epoch 9/10 24/24 ----Epoch 10/10 24/24 ---------- 7s 309ms/step - accuracy: 1.0000 - loss: 0.0036

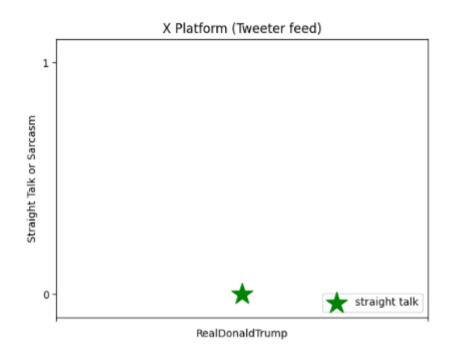
<keras.src.callbacks.history.History at 0x7bffe6bc5e10>

Sample graphic to represent prediction of Sarcasm or Not/Straight Talk

```
[ ] 1 # TO BE POINTED AT THE NEW DATA #
     2 # TO BE POINTED AT THE NEW DATA #
     3 # TO BE POINTED AT THE NEW DATA # - originally form Truth Fake module
     4
     5 import matplotlib.pyplot as plt
     6 from matplotlib.ticker import StrMethodFormatter
     7
      0
     9 plt.subplot(1, 2, 1) # row 1, column 2, count 1
     10 # x-axis values
     11 x = [0] # [1,2,3,4,5,6,7,8,9,10]
     12 # y-axis values
     13 y = df_new_texts['predicted_true_fake'].iloc[0] # [2,4,5,7,6,8,9,11,12,12]
     14
     15
     16 if y == 0:
     17 # plotting points as a scatter plot
     18 plt.scatter(x, y, label= "rasberry", color= "red",
     19 marker= "X", s=450)
     20
     21 if y == 1:
     22 # plotting points as a scatter plot
     23 plt.scatter(x, y, label= "stars", color= "green",
     24 marker= "*", s=450)
     25
     26
     27 # x-axis label
     28 plt.xlabel('RealDonaldTrump')
     29 # frequency label
     30 plt.ylabel('Fake or Truth')
     31 # plot title
     32 plt.title('X Platform (Twitter feed)')
     33 # showing legend
     34 plt.legend()
     35 # Setting the axis range
     36 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
     37 plt.ylim(-0.1, 1.1)
     38 plt.gca().set_xticklabels([])
    39
40 #plt.set yticklabels([str(round(float(label), 0)) for target in y])
```

```
3/ plt.ylim(-0.1, 1.1)
    38 plt.gca().set_xticklabels([])
    39
    40 #plt.set yticklabels([str(round(float(label), 0)) for target in y])
    41 plt.gca().yaxis.set_major_formatter(StrMethodFormatter('{x:,.0f}')) # No decimal
    42 plt.xticks([-1, 1])
    43 plt.yticks([0, 1])
    44
    45
    46 plt.subplot(1, 2, 2) # row 1, column 2, count 1
    47 # x-axis values
    48 x = [0] # [1,2,3,4,5,6,7,8,9,10]
    49 # y-axis values
    50 #y = df_new_texts['target'].iloc[0] # [2,4,5,7,6,8,9,11,12,12]
    51 #y = int(prediction[0]) # [2,4,5,7,6,8,9,11,12,12]
    52 y = prediction[0] # [2,4,5,7,6,8,9,11,12,12]
    53
    54 if y >= 0.99:
    55 # plotting points as a scatter plot
    56 plt.scatter(x, y, label= "rasberry talk", color= "red",
    57 marker= "X", s=450)
    58 plt.legend(loc='upper right')
    59
    60 if y == 0:
    61 # plotting points as a scatter plot
    62 plt.scatter(x, y, label= "straight talk", color= "green",
    63 marker= "*", s=450)
    64 plt.legend(loc='lower right')
    65
    66 # x-axis label
    67 plt.xlabel('RealDonaldTrump')
    68 # frequency label
    69 plt.ylabel('Straight Talk or Sarcasm')
    70 # plot title
    71 plt.title('X Platform (Twitter feed)')
    72 # showing legend
    73 #plt.legend()
    74 #plt.legend(['Legend'], loc='upper left')
    75 # Setting the axis range
    76 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
   77 nlt.vlim(-0.1. 1.1)
```

```
75 # Setting the axis range
76 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
77 plt.ylim(-0.1, 1.1)
78 plt.gca().set_xticklabels([])
79
80 #plt.set_yticklabels([str(round(float(label), 0)) for target in y])
81 plt.gca().yaxis.set_major_formatter(StrMethodFormatter('{x:,.0f}')) # No decimal
82 plt.xticks([-1, 1])
83 plt.yticks([0, 1])
84
85 # function to show the plot
86 #plt.show()
87
88 # function to show the plot
89 plt.show()
```





# EDA / Exploratory data analysis

```
[] 1 # Needs adjusting for three categories of labels 0,1,2 / true-neutral, misinformation(mistakes), disinformation(lies)
     2
     3 train_df.loc[train_df['label'] == 2, 'tweet'].str.len().apply(np.log1p).hist(label='disinformation', alpha=.5)
     4 train_df.loc[train_df['label'] == 1, 'tweet'].str.len().apply(np.log1p).hist(label='misinformation', alpha=.5)
     5 train_df.loc[train_df['label'] == 0, 'tweet'].str.len().apply(np.logip).hist(label='normal', alpha=.5)
     6 plt.legend();
÷
            disinformation
     120 - misinformation
            normal
      100
       80
       60
       40
       20
        0
                              4.0
                                      4.2
                      3.8
              3.6
                                              4.4
                                                      4.6
                                                              4.8
                                                                      5.0
```

**Build TF-iDF model (base model)** 

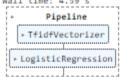
### **Build pipeline**

```
1 %%time
2 tfidf_logit_pipeline.fit(train_texts, y_train)
```

```
    [Parallel(n_jobs=4)]: Using backend LokyBackend with 4 concurrent workers.

    CPU times: user 122 ms, sys: 79 ms, total: 201 ms

    Wall time: 4.59 s
```



### **Base model: TF-iDF scores (accuracy + f1 + precision + recall)**

```
1 accuracy_value = accuracy_score(y_valid, valid_pred)
2 print("Accuracy score = ", accuracy_value)
3 f1_score_value = f1_score(y_valid, valid_pred, average='macro')
4 print("F1_score = ", f1_score_value)
5 precision_score_value = precision_score(y_valid, valid_pred, average='macro')
6 print("Precison score = ", precision_score_value)
7 recall_score_value = recall_score(y_valid, valid_pred, average='macro')
8 print("Recall score = ", recall_score_value)
9
```

Accuracy score = 0.6064516129032258 F1\_score = 0.37997052627772204 Precison score = 0.5466183574879228

# Y New Section - Prediction on Pulled from X / Twitter tweets

Build out the memory / file from previous two notebooks that added the Truth-Fake, then Sarcasm features/mettrics

```
[] 1 # NEW SECTION
2
3 df_new_texts['predicted_misinformation'] = 0
4 df_new_texts['predicted_disinformation'] = 0
5
6 for i in range(len(predicted_value)):
7 | if predicted_value[i] == 1:
8 | df_new_texts.loc[i, 'predicted_misinformation'] = 1
9 | elif predicted_value[i] == 2:
10 | df_new_texts.loc[i, 'predicted_disinformation'] = 1
11
12 df_new_texts = df_new_texts[['id', 'content', 'clean_text', 'clean_joined_text', 'predicted_true_fake', 'predicted_sarcasm', 'predicted_misinformation', 'predicted_disinformation']]
13
14 #print(df_new_texts['predicted_disinformation'])
```

~ ~

# Model

```
O
    1 def plot_confusion_matrix(actual, predicted, classes,
     2
                                normalize=False,
     3
                                title='Confusion matrix', figsize=(7,7),
     4
                                cmap=plt.cm.Blues, path_to_save_fig=None):
           .....
     5
     6
          This function prints and plots the confusion matrix.
           Normalization can be applied by setting 'normalize=True'.
     7
           .....
     8
     9
          import itertools
    10
           from sklearn.metrics import confusion_matrix
           cm = confusion matrix(actual, predicted).T
    11
           if normalize:
    12
    13
           cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
    14
    15
           plt.figure(figsize=figsize)
    16
           plt.imshow(cm, interpolation='nearest', cmap=cmap)
    17
           plt.title(title)
    18
          plt.colorbar()
          tick_marks = np.arange(len(classes))
    19
    20
           plt.xticks(tick_marks, classes, rotation=90)
    21
           plt.yticks(tick_marks, classes)
    22
    23
           fmt = '.2f' if normalize else 'd'
    24
           thresh = cm.max() / 2.
    25
           for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
    26
              plt.text(j, i, format(cm[i, j], fmt),
    27
                       horizontalalignment="center",
                       color="white" if cm[i, j] > thresh else "black")
    28
    29
    30
           plt.tight_layout()
           plt.ylabel('Predicted label')
    31
    32
          plt.xlabel('True label')
```

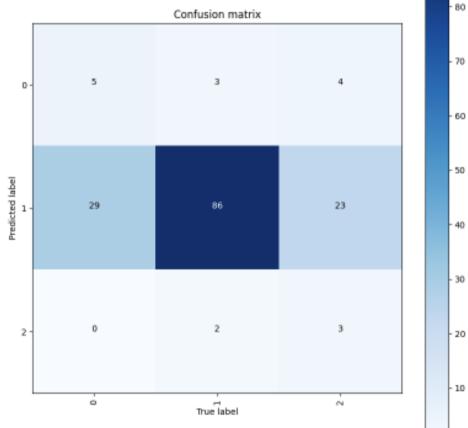
32	<pre>plt.xlabel('True label')</pre>
33	
34	<pre>if path_to_save_fig:</pre>
35	<pre>plt.savefig(path_to_save_fig, dpi=300, bbox_inches='tight')</pre>

### Read in memory file of previous notebooks features, Truth-Fake, Sarcasm then write the new feature Stance to the memory file

```
0
   1 import csv
     2
     3 # Step 1: Read the existing CSV file and store its content
     4 csv_file_path = 'data_to_graph.csv'
     5 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'r') as file:
     6 reader = csv.reader(file)
     7 data = list(reader)
     8
     9 # Step 2: Define the values for the new column
    10 new_values = valid_pred # df_new_texts['predicted_sarcasm'] # ['New York', 'Los Angeles', 'Chicago', 'San Francisco']
    11
    12 # Step 3: Add the new column header to the first row of the data
    13 data[0].append('predicted_stance')
    14
    15 # Step 4: Add the new column values to the remaining rows of the data
    16 for i in range(1, 154): # len(data)):
    17 data[i].append(new_values[i - 1])
    18
    19 # Step 5: Write the updated data back to the CSV file
    20 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'w', newline='') as file:
    21 writer = csv.writer(file)
    22 writer.writerows(data)
```

#### Confusion matrix





41

Lο

Side-by-Side graphic: three metrics (True v Fake news + Sarcasm + Stance) visualised in graph version 1 i.e., the side-by-side graph

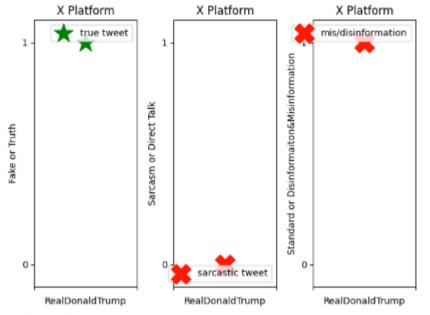
Sample graphic to represent 3 metrics i) True vs Fake ; ii) Straight vs Sarcasm ; iii) Misinformation vs Disinformation

```
1 # TO BE POINTED AT THE NEW DATA #
     2 # TO BE POINTED AT THE NEW DATA #
     3 # TO BE POINTED AT THE NEW DATA # - originally form Truth Fake module
     4
     5 import time
     6 import matplotlib.pyplot as plt
     7 from matplotlib.ticker import StrMethodFormatter
     8
    9 # measure User ; Sys ; and Wall time for evaluation
    10 #%%time
    11 start time = time.process time()
    12 # Code to measure
    13
    14 plt.subplot(1, 3, 1) # row 1, column 2, count 1
    15 # x-axis values
    16 \times = [0] \# [1,2,3,4,5,6,7,8,9,10]
    17 # y-axis values
    18
    19 df_new_texts = pd.read_csv('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv')
    20 y = df_new_texts['predicted_true_fake'].iloc[0] # [2,4,5,7,6,8,9,11,12,12]
    21
    22 if y == 0:
    23 # plotting points as a scatter plot
    24 plt.scatter(x, y, label= "fake tweet", color= "red",
    25 marker= "X", s=450)
    26 plt.legend(loc='lower left')
    27 if y >= 0.99:
    28 # plotting points as a scatter plot
    29 plt.scatter(x, y, label= "true tweet", color= "green",
    30 marker= "*", s=450)
    31 plt.legend(loc='upper right')
    32
    33 # x-axis label
    34 plt.xlabel('RealDonaldTrump')
    35 # frequency label
    36 plt.ylabel('Fake or Truth')
```

```
35 # frequency label
36 plt.ylabel('Fake or Truth')
37 # plot title
38 plt.title('X Platform')
39 # showing legend
40 #plt.legend()
41 # Setting the axis range
42 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
43 plt.ylim(-0.1, 1.1)
44 plt.gca().set_xticklabels([])
45
46 #plt.set_yticklabels([str(round(float(label), 0)) for target in y])
47 plt.gca().yaxis.set_major_formatter(StrMethodFormatter('{x:,.0f}')) # No decimal
48 plt.xticks([-1, 1])
49 plt.yticks([0, 1])
50
51
52 plt.subplot(1, 3, 2) # row 1, column 2, count 1
53 # x-axis values
54 x = [0] \# [1,2,3,4,5,6,7,8,9,10]
55 # y-axis values
56 #y = df_new_texts['target'].iloc[0] # [2,4,5,7,6,8,9,11,12,12]
57 #y = int(prediction[0]) # [2,4,5,7,6,8,9,11,12,12]
58 #y = prediction[0] # [2,4,5,7,6,8,9,11,12,12]
59 y = df_new_texts['predicted_sarcasm'].iloc[0]
60
61
62 if y >= 0.99:
63 # plotting points as a scatter plot
64 plt.scatter(x, y, label= "direct tweet", color= "green",
65 marker= "*", s=450)
66 plt.legend(loc='upper right')
67 if y == 0:
68 # plotting points as a scatter plot
69 plt.scatter(x, y, label= "sarcastic tweet", color= "red",
70
              marker= "X", s=450)
71 plt.legend(loc='lower right')
72
73 # x-axis label
74 plt.xlabel('RealDonaldTrump')
```

```
73 # x-axis label
    74 plt.xlabel('RealDonaldTrump')
    75 # frequency label
   76 plt.ylabel('Sarcasm or Direct Talk')
   77 # plot title
   78 plt.title('X Platform')
    79 # showing legend
    80 #plt.legend()
    81 #plt.legend(['Legend'], loc='upper left')
    82 # Setting the axis range
    83 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
    84 plt.ylim(-0.1, 1.1)
    85 plt.gca().set_xticklabels([])
    86
    87 #plt.set_yticklabels([str(round(float(label), 0)) for target in y])
    88 plt.gca().yaxis.set_major_formatter(StrMethodFormatter('{x:,.0f}')) # No decimal
    89 plt.xticks([-1, 1])
    90 plt.yticks([0, 1])
   91
   92 # function to show the plot
   93 #plt.show()
    94
   95
   96 plt.subplot(1, 3, 3) # row 1, column 2, count 1
   97 # x-axis values
   98 x = [0] # [1,2,3,4,5,6,7,8,9,10]
   99 # y-axis values
   100
   101 #y = predicted_value.iloc[0] # valid_pred # df_new_texts['predicted_true_fake'].iloc[0] # [2,4,5,7,6,8,9,11,12,12]
   102 y = df_new_texts['predicted_stance'].iloc[0]
   103
   104 if y == 0:
   105 # plotting points as a scatter plot
   106 plt.scatter(x, y, label= "neutral tweet", color= "green",
   107 marker= "star", s=450)
   108 plt.legend(loc='lower left')
   109 if y >= 0.99:
110 # plotting points as a scatter plot
```

```
.
109 if y >= 0.99:
110 # plotting points as a scatter plot
111 plt.scatter(x, y, label= "mis/disinformation", color= "red",
112
     marker= "X", s=450)
113 plt.legend(loc='upper right')
114
115 # x-axis label
116 plt.xlabel('RealDonaldTrump')
117 # frequency label
118 plt.ylabel('Standard or Disinformaiton&Misinformation')
119 # plot title
120 plt.title('X Platform')
121 # showing legend
122 #plt.legend()
123 # Setting the axis range
124 plt.xlim(0,0) # (-0.5, 0.5) # X-axis range from 0 to 5
125 plt.ylim(-0.1, 1.1)
126 plt.gca().set_xticklabels([])
127
128 #plt.set_yticklabels([str(round(float(label), 0)) for target in y])
129 plt.gca().yaxis.set_major_formatter(StrMethodFormatter('{x:,.0f}')) # No decimal
130 plt.xticks([-1, 1])
131 plt.yticks([0, 1])
132
133
134 # function to show the plot
135 plt.tight_layout()
136 plt.show()
137
138 end_time = time.process_time()
139 print(f"CPU time used: {end_time - start_time} seconds")
```



CPU time used: 0.7297617659999993 seconds

# New model (LSTM) to see can an improvement be made to the base model of Tf-iDF + Logistic Regression

- ->>NEW LSTM in Stance notebook<<- Model Improvement\*\* above tfidf+logit
- regression 70% accurate -> trying here LSTM to get higher accuracy bold text



```
25 from keras.metrics import F1Score # (average=None, threshold=None, name="f1_score", dtype=None)
26
27
28 # The input text, example could be list of sentences
29 texts = train_texts[:] # [...]
30
31 # The labels corresponding to the input text
32 labels = y train[:] # train_df['label'] # train_texts['Author'] # [...]
33
34 # Hyperparameters
35 max words = 10000 # max number of words to use in the vocabulary
36 max len = 200 # max length of each text (in terms of number of words)
37 embedding dim = 200 # dimension of word embeddings
38 lstm units = 64 # number of units in the LSTM layer
39 num_classes = len(set(labels)) # 2 # len(y_train) # y_train.unique() # 2 # len(set(labels)) # number of classes
40
41 # Tokenize the texts and create a vocabulary
42 tokenizer = Tokenizer(num_words=max_words)
43 tokenizer.fit_on_texts(texts.values.astype('U'))
44 sequences = tokenizer.texts_to_sequences(texts.values.astype('U'))
45
46 # Pad the sequences so they all have the same length
47 x = pad_sequences(sequences, maxlen=max_len)
48
49 # Create one-hot encoded labels
50 y = keras.utils.to_categorical(labels,
51 num_classes)
52
53 # Build the model
54 model = Sequential()
55 model.add(Embedding(max_words, embedding_dim, input_length=max_len))
56 model.add(LSTM(1stm units))
57 model.add(Dense(num_classes, activation='softmax')) # 'relu''; "rmsprop"
58
59 # optimizer='adam', loss='mse'
57 model.add(Dense(num classes, activation='softmax')) # 'relu''; "rmsprop"
58
59 # optimizer='adam', loss='mse'
60
61
62 # Compile the model
63 model.compile(loss='categorical_crossentropy', optimizer='adam', metrics=['accuracy']) # 'binary_crossentropy'
64
65 # train_texts, valid_texts, y_train, y_valid
66
67 # Train the model
68 model.fit(x, y, batch_size=32, epochs=10)
```

# Epoch training (data used is shortened to make the code run within the time limit of the demonstation)

Epoch 1/10
15/15 11s 421ms/step - accuracy: 0.5086 - loss: 1.0644
Epoch 2/10
15/15 10s 395ms/step - accuracy: 0.5684 - loss: 0.9416
Epoch 3/10
15/15 4s 262ms/step - accuracy: 0.5929 - loss: 0.8251
Epoch 4/10
15/15 4s 259ms/step - accuracy: 0.7658 - loss: 0.5960
Epoch 5/10
15/15 5s 367ms/step - accuracy: 0.8924 - loss: 0.3260
Epoch 6/10
15/15 5s 294ms/step - accuracy: 0.9796 - loss: 0.1402 Epoch 7/10
15/15 4s 197ms/step - accuracy: 0.9893 - loss: 0.0688
Epoch 8/10
15/15 5s 202ms/step - accuracy: 0.9940 - loss: 0.0424
Epoch 9/10
15/15 6s 273ms/step - accuracy: 0.9994 - loss: 0.0186
Epoch 10/10
15/15 3s 197ms/step - accuracy: 0.9959 - loss: 0.0161
<keras.src.callbacks.history.history 0x79d66adf25c0="" at=""></keras.src.callbacks.history.history>

### Point the new model (LSTM) at the new data

```
Using NEW data -> predict using NEW model LSTM previously trained on reference
ground truth labelled stance data
```

```
[ ] 1# Using NEW data -> predict using NEW model LSTM prviously trained on reference ground truth labelled stance data
     2 # write Stance prediction results to shared dataframe
     3
     4 import numpy as np
     5
     6 #X = df_new_texts['Text']
     7
     8 # Tokenize the texts and create a vocabulary
     9 tokenizer = Tokenizer(num_words=max_words)
    10 #tokenizer.fit on texts(texts.values.astype('U'))
    11 tokenizer.fit_on_texts(df_new_texts['content'].values.astype('U'))
    12 # sequences = tokenizer.texts_to_sequences(texts.values.astype('U'))
    13 sequences = tokenizer.texts_to_sequences(df_new_texts['content'].values.astype('U'))
    14
    15 # Pad the sequences so they all have the same length
    16 X = pad_sequences(sequences, maxlen=max_len)
    17
    18 prediction = []
    19 prediction = model.predict(X)
    20
    21 # Creating multi-dimension array
    22 #array1 = [1, 2, 4, [5, [6, 7]]]
    23
    24 # Object Data type is accept all data-type
    25 Data_type = int
    26
    27 # Now we fix the error
    28 #np_array = numpy.array(prediction, dtype=Data_type)
    29 # TEMP commented out
    30 #prediction = np.array(prediction, dtype=Data_type)
    31 #prediction = prediction.flatten('C')
    32
    33
    34 df_new_texts['predicted_stance_neutral'] = 0
    35 df_new_texts['predicted_stance_misinformation'] = 0
    36 df_new_texts['predicted_stance_disinformation'] = 0
    37 df_new_texts['predicted_stance_all_types'] = 0
    38
    20
```



76 df\_new\_texts.info()
77 df\_new\_texts.describe()
78 print(len(prediction))

80 len(X)

79 #print(len( df new texts['predicted stance']))

```
51
```

# Training new model and data output

content	id	
	965925223949357056	0
Thank you to @ foxandfriends for the great tim		0
"There is no serious person out there who woul	965928352614965248	1
The President Obama quote just before elec	965930611272712192	2
Republicans are now leading the Generic Poll,	965932714141650946	3
Matt Schlapp and CPAC are getting ready for an	965935035328155649	4
Joe Biden was a TOTAL FAILURE in Government. H	1273405198698975232	9721
Will be interviewed on @ seanhannity tonight a	1273408026968457216	9722
pic.twitter.com/31m1spbU8X	1273442195161387008	9723
pic.twitter.com/vpCE5MadUz	1273442469066276864	9724
pic.twitter.com/VL1c0BHW41	1273442528411385858	9725

### Write third (Stance) metric to the 'memory' / built up csv file

# IMPORTANT - Only run ONCE

- · write feaure column for neutral stance
- · write feaure column for misinformation
- · write feaure column for disinformation
- · write feaure column for all stance types

```
1 # IMPORTANT - Only run ONCE
     2 # save shared data frame with Saracasm prediction resulls added to the shared file
     3
     4 import csv
     5
     6 # # Step 1: Read the existing CSV file and store its content
     7 # csv_file_path = 'data_to_graph.csv'
     8 # with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'r') as file:
            reader = csv.reader(file)
     94
    10 #
            data = list(reader)
    11
    12 # # Step 2: Define the values for the new "City" column
    13 # new_values = df_new_texts['predicted_sarcasm'] # ['New York', 'Los Angeles', 'Chicago', 'San Francisco']
    14
    15 # # Step 3: Add the new "City" column header to the first row of the data
    16 # data[0].append('predicted_sarcasm')
    17
    18 # # Step 4: Add the new "City" column values to the remaining rows of the data
    19 # for i in range(1, len(data)):
    20 # data[i].append(new_values[i - 1])
    21
```

```
20 #
        data[i].append(new_values[i - 1])
21
22 # # Step 5: Write the updated data back to the CSV file
23 # with open('/content/drive/MyDrive/Colab Notebooks/MSCAITOP/data to graph.csv', 'w', newline='') as file:
24 #
        writer = csv.writer(file)
25 #
        writer.writerows(data)
26
27
28 # Step 1: Read the existing CSV file and store its content
29 csv file path = 'data to graph.csv'
30 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'r') as file:
31 reader = csv.reader(file)
32 data = list(reader)
33 #Step 2 , 3, 4 repeated
34 new_values = df_new_texts['predicted_stance_neutral']
35 data[0].append('predicted_stance_neutral')
36 #data = data[['Author', 'Text', 'clean_text', 'clean_joined_text', 'predicted_true_fake', 'predicted_sarcasm']].copy()
37 for i in range(1, len(data)):
38 data[i].append(new values[i - 1])
39 # Step 5: Write the updated data back to the CSV file
40 with open('/content/drive/MyDrive/Colab Notebooks/MSCAITOP/data to graph.csv', 'w', newline='') as file:
41 writer = csv.writer(file)
42 writer.writerows(data)
43
44 # Step 1: Read the existing CSV file and store its content
45 csv_file_path = 'data_to_graph.csv'
46 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'r') as file:
47 #file = file[['Author', 'Text', 'clean text', 'clean joined text', 'predicted true fake', 'predicted sarcasm', 'predicted stance neutral']].copy()
48
     reader = csv.reader(file)
49 data = list(reader)
50 #Step 2, 3, 4 repeated
51 new_values = df_new_texts['predicted_stance_misinformation']
52 data[0].append('predicted_stance_misinformation')
53 for i in range(1, len(data)):
54 data[i].append(new_values[i - 1])
55 # Step 5: Write the updated data back to the CSV file
56 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'w', newline='') as file:
57 writer = csv.writer(file)
58 writer.writerows(data)
```

```
58 writer.writerows(data)
59
60 # Step 1: Read the existing CSV file and store its content
61 csv_file_path = 'data_to_graph.csv'
62 with open('/content/drive/MyDrive/Colab Notebooks/MSCAITOP/data to graph.csv', 'r') as file:
63 #file = file[['Author', 'Text', 'clean_text', 'clean_joined_text', 'predicted_true_fake', 'predicted_sarcasm', 'predicted_stance_neutral', 'predicted_stance_misinformation']].copy()
64 reader = csv.reader(file)
65 data = list(reader)
66 #Step 2, 3, 4 repeated
67 new values = df new texts['predicted stance disinformation']
68 data[0].append('predicted_stance_disinformation')
69 for i in range(1, len(data)):
70 data[i].append(new_values[i - 1])
71 # Step 5: Write the updated data back to the CSV file
72 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'w', newline='') as file:
73 writer = csv.writer(file)
74 writer.writerows(data)
75
76
77 # Step 1: Read the existing CSV file and store its content
78 csv_file_path = 'data_to_graph.csv'
79 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'r') as file:
80 #file = file[['Author', 'Text', 'clean text', 'predicted true fake', 'predicted sarcasm', 'predicted stance neutral', 'predicted stance misinformation', 'predicted stance disinformation']].copy()
81 reader = csv.reader(file)
82 data = list(reader)
83 #Step 2, 3, 4 repeated
84 new_values = df_new_texts['predicted_stance_all_types']
85 data[0].append('predicted_stance_all_types')
86 for i in range(1, len(data)):
87 data[i].append(new_values[i - 1])
88 # Step 5: Write the updated data back to the CSV file
89 with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'w', newline='') as file:
90 writer = csv.writer(file)
91 writer.writerows(data)
92
93 # # Step 5: Write the updated data back to the CSV file
94 # with open('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv', 'w', newline='') as file:
95 # writer = csv.writer(file)
96 # writer.writerows(data)
 12.21.44
               WITCH - CAVINITCH (ITTE)
 96 4
               writer.writerows(data)
  97
  QR.
 99 print('predicted sarcasm = ', df new texts['predicted sarcasm'].sum())
100 print('predicted stance neutral = ', df new texts['predicted stance neutral'].sum())
101 print('predicted stance misinformation = ', df new texts['predicted stance misinformation'].sum())
102 print('predicted stance_disinformation = ', df new texts['predicted_stance_disinformation'].sum())
103 print('predicted stance all types = ', df new texts['predicted stance all types'].sum())
104 #print(df_new_texts['predicted_stance_all_types'].value_counts)
```

Output new version 2 of the visualisation i.e., the combined chart (of the three metrics True v Fake news, Sarcasm, Stance)

# ~ NEW combined chart

### Plot version 2 chart - the combined features chart

```
[ ] 1 # NEW combined chart
     2
     3 # importing package
     4 import matplotlib.pyplot as plt
     5 import numpy as np
     6 import pandas as pd
     7
     8 from matplotlib.ticker import StrMethodFormatter
     9
    10 import io
    11 import xml.etree.ElementTree as ET
    12 import time
    13
    14 from matplotlib.lines import Line2D
    15 from matplotlib.markers import MarkerStyle
    16 from matplotlib.transforms import Affine2D
    17
    18 from google.colab import drive
    19 drive.mount('/content/drive/')
    20
    21 # x-axis values
    22 x = [0] # [1,2,3,4,5,6,7,8,9,10]
    23 # y-axis values
    24
    25 # measure User ; Sys ; and Wall time for evaluation
    26
    27 start_time = time.process_time()
    28 # Code to measure
    29
    30 df_new_texts = pd.read_csv('/content/drive/MyDrive/Colab_Notebooks/MSCAITOP/data_to_graph.csv')
    31
    32
    33 for i in range(10):
    34 y1 = df_new_texts['predicted_true_fake'].iloc[i] # [2,4,5,7,6,8,9,11,12,12]
    35 y2 = df_new_texts['predicted_sarcasm'].iloc[i]
    36 #y3 = df_new_texts['predicted_stance_neutral'].iloc[i]
    37 y3 = df_new_texts['predicted_stance_all_types'].iloc[i]
    38
    39 # 'predicted_stance_misinformation'
    40 # 'predicted_stance_disinformation'
```

```
44.1
42 #y4 = df new texts['predicted stance misinformation'].iloc[0]
43 #y5 = df new texts['predicted stance disinformation'].iloc[0]
44
45 # create data
46 # x = [1,2,3,4,5]
47 # y = [3,3,3,3,3]
48
49 # text_style = dict(horizontalalignment='right', verticalalignment='center',
50 #
                       fontsize=12, fontfamily='monospace')
51 # marker_style = dict(linestyle=':', color='0.8', markersize=10,
52 #
                        markerfacecolor="tab:blue", markeredgecolor="tab:blue")
53
54 # y1 = [0]
55 # y2 = [0]
56 # v3 = [0]
57 # all good / all zeros / true news 0 / straight talk 0/ neutral stance 0
58 if y1 == [0] and y2 == [0] and y3 == [0]:
59 #print('Test = if block runs')
60
     # plot lines
61 x1 = [3,0]
62
     y1 = [3,0]
     #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
63
64
     plt.plot(x1, y1, label = "True News", color='green') # , **marker_style, **text_style) # linestyle="-", color='red'
65
     plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
66
     plt.text(-3, -7, 'True News')
67
     x2 = [3,0]
68
     y_2 = [5, 0]
     #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
69
      plt.plot(x2, y2, label = "Direct Talk", color='green') # , **marker style, **text_style) # linestyle="-", color='red'
70
71
     plt.plot(-3, -5, marker="0", markersize=20, markeredgecolor="green", markerfacecolor="green")
72
     plt.text(-3, -7, 'Direct Talk')
73
     x3 = [3,0]
74
     y3 = [7,0]
75
     #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
76
     plt.plot(x3, y3, label = "Stance Neutral", color='green') # , **marker style, **text_style) # linestyle="-", color='red'
```

77 plt.plot(3, 7, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")

81 # y1 = [1] 82 # y2 = [1] 83 # y3 = [1] 84 # all up / all ones / all fake 1 / sarcasm 1/ all misinformation 1 85 if y1 == [1] and y2 == [1] and y3 == [1]: 86 #print('Test = if block runs') 87 # plot lines 88 x1 = [-3, 0]89 y1 = [-3, 0]90 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red'</pre> 91 plt.plot(x1, y1, label = "Fake News", color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 92 plt.plot(-3, -3, marker="0", markersize=20, markeredgecolor="red", markerfacecolor="red") 93 plt.text(-3, -3, 'Fake News') 94 x2 = [-3,0]95 v2 = [-5,0]#plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red'</pre> 96 plt.plot(x2, y2, label = "Sarcasm", color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red' 97 98 plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red") 99 plt.text(-3, -5, 'Sarcasm') 100 x3 = [-3,0]101 y3 = [-7,0]#plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red'</pre> 102 plt.plot(x3, y3, label = "Misinformation", color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 103 104 plt.plot(-3, -7, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red") 105 plt.text(-3, -7, 'Misinformation') 106 107 # y1 = [1] 108 # y2 = [0] 109 # y3 = [1] 110 # all good / all zeros / true news 0 / straight talk 0/ neutral stance 0 111 if y1 == [1] and y2 == [0] and y3 == [1]: 112 #print('Test = if block runs')

```
111 if y1 == [1] and y2 == [0] and y3 == [1]:
112 #print('Test = if block runs')
113
       # plot lines
114
       x1 = [-3, 0]
115
       v1 = [-3,0]
116
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
117
       plt.plot(x1, y1, label = "Fake News", color='red') # , **marker style, **text style) # linestyle="-", color='red'
118
       plt.plot(-3, -3, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
       plt.text(-3, -3, 'Fake News')
119
120
       x2 = [3,0]
121
       y_2 = [5,0]
122
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
       plt.plot(x2, y2, label = "Direct Talk", color='green') # , **marker style, **text style) # linestyle="-", color='red'
123
124
       plt.plot(3, 5, marker="0", markersize=20, markeredgecolor="green", markerfacecolor="green")
125
       plt.text(3, 5, 'Direct Talk')
126
       x3 = [-3,0]
127
       y3 = [-7,0]
128
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
129
       plt.plot(x3, y3, label = "Stance Misinformation", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
130
       plt.plot(-3, -7, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
131
       plt.text(-3, -7, 'Misinformation')
132
133 # y1 = [0]
134 # y2 = [0]
135 # y3 = [1]
136 # Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1
137 if y1 == [0] and y2 == [0] and y3 == [1]:
138 x1 = [3,0]
139
       y1 = [3,0]
140
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
141
       plt.plot(x1, y1, label = "True News", color='green') # , **marker_style, **text_style) # linestyle="-", color='red'
142
       plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
143
       plt.text(3, 3, 'True News')
144
       x2 = [3,0]
145
       y2 = [5,0]
146
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
147
       plt.plot(x2, y2, label = "Direct Talk", color='green') # , **marker_style, **text_style) # linestyle="-", color='red'
148
       plt.plot(3, 5, marker="0", markersize=20, markeredgecolor="green", markerfacecolor="green")
149
       plt.text(3, 5, 'Direct Talk')
```

151 y3 = [-7,0] 152 #plt.plot(x1, v1, label = "line 1", marker='<', markersize=20, color='red') # . \*\*marker style, \*\*text style) # linestyle="-", color='red' 153 plt.plot(x3, y3, label = "Misinformation", color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 154 plt.plot(-3, -7, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red") 155 plt.text(-3, -7, 'Misinformation') 156 157 plt.legend(loc='lower right') 158 from matplotlib import pyplot as plt plt.savefig('/content/drive/My Drive/Colab Notebooks/MSCAITOP/MSCAITOP graphic v2a.png', bbox inches='tight') 159 160 161 162 # y1 = [0] 163 # y2 = [1] 164 # y3 = [1] 165 # passed testing 166 # Mixed / all fake 0/sarcasm 1/all misinformation&disinformation 1 167 if y1 == [0] and y2 == [1] and y3 == [1]: 168 x1 = [3,0] 169 y1 = [3,0]#plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red'</pre> 170 171 plt.plot(x1, y1, label = "True News", color='green') # , \*\*marker style, \*\*text style) # linestyle="-", color='red' 172 plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green") 173 plt.text(3, 3, 'True News') 174 x2 = [-3,0] 175 y2 = [-5,0]#plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red'</pre> 176 177 plt.plot(x2, y2, label = "Sarcasm Talk", color='red') # , \*\*marker style, \*\*text\_style) # linestyle="-", color='red' 178 plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red") 179 plt.text(-3, -5, 'Sarcasm Talk') 180 x3 = [-3,0]181 y3 = [-7,0]182 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red'</pre> 183 plt.plot(x3, y3, label = "Misinformation", color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 184 plt.plot(-3, -7, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red") 185 plt.text(-3, -7, 'Misinformation')

184	<pre>plt.plot(-3, -7, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")</pre>
185	<pre>plt.text(-3, -7, 'Misinformation')</pre>
186	
187	# y1 = [0]
188	# y2 = [1]
189	# y3 = [0]
190	# Mixed / all fake 0/sarcasm 1/all misinformation&disinformation 0
191	if $y_1 == [0]$ and $y_2 == [1]$ and $y_3 == [0]$ :
192	x1 = [3, 0]
193	$y_1 = [3, 0]$
194	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
195	plt.plot(x1, y1, label = "True News", color='green') # , **marker_style, **text_style) # linestyle="-", color='red'
196	plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
197	plt.text(3, 3, 'True News')
198	$x^2 = [-3, 0]$
199	$y_2 = [-5, 0]$
200	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
201	plt.plot(x2, y2, label = "Sarcasm Talk", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
202	plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
203	plt.text(-3, -5, 'Sarcasm Talk')
204	$x_3 = [3, 0]$
205	$y_3 = [7, 0]$
206	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
207	<pre>plt.plot(x3, y3, label = "Stance_Neutral", color='green') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
208	plt.plot(3, 7, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
209	<pre>plt.text(3, 7, 'Stance_Neutral')</pre>
210	
211	# y1 = [1]
212	# y2 = [1]
213	# y3 = [0]
214	# Mixed / all fake 1/sarcasm 1/all misinformation&disinformation 0
215	if y1 == [1] and y2 == [1] and y3 == [0]:
216	x1 = [-3, 0]
217	y1 = [-3, 0]
218	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
219	<pre>plt.plot(x1, y1, label = "Fake News", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
220	<pre>plt.plot(-3, -3, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")</pre>
221	plt.text(-3, -3, 'Fake News')
222	$x^2 = [-3, 0]$

```
220
     plt.plot(-3, -3, marker="0", markersize=20, markeredgecolor="red", markerfacecolor="red")
221
     plt.text(-3, -3, 'Fake News')
222
      x2 = [-3, 0]
223 v2 = [-5,0]
224
     #plt.plot(x1, v1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'
       plt.plot(x2, y2, label = "Sarcasm Talk", color='red') # , **marker style, **text style) # linestyle="-", color='red'
225
       plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
226
227
       plt.text(-3, -5, 'Saracasm Talk')
228 x3 = [3,0]
229 y3 = [7,0]
     #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
230
     plt.plot(x3, y3, label = "Stance Neutral", color='green') # , **marker style, **text style) # linestyle="-", color='red'
231
      plt.plot(3, 7, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
232
233 plt.text(3, 7, 'Stance_Neutral')
234
235 # y1 = [1]
236 # y2 = [1]
237 # y3 = [0]
238 # Mixed / all fake 1/sarcasm 1/all misinformation&disinformation 0
239 if y1 == [1] and y2 == [1] and y3 == [0]:
240 x1 = [-3,0]
241 y1 = [-3,0]
242 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
243 plt.plot(x1, y1, label = "Fake News", color='red') # , **marker style, **text style) # linestyle="-", color='red'
244 plt.plot(-3, -3, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
245
     plt.text(-3, -3, 'Fake News')
      x2 = [-3, 0]
246
247
      y_2 = [-5, 0]
      #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
248
249
       plt.plot(x2, y2, label = "Sarcasm Talk", color='red') # , **marker style, **text style) # linestyle="-", color='red'
250
       plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
251
       plt.text(-3, -5, 'Sarcasm Talk')
252
      x3 = [3,0]
253
     y_3 = [7,0]
254
      #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
       plt.plot(x3, y3, label = "Stance Neurtral", color='green') # , **marker style, **text style) # linestyle="-", color='red'
255
256
       plt.plot(3, 7, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
257
     plt.text(3, 7, 'Stance Neutral')
250
```

```
256
       plt.plot(3, 7, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
257 plt.text(3, 7, 'Stance Neutral')
258
259 # y1 = [1]
260 # y2 = [0]
261 # y3 = [0]
262 #passed testing
263 # Mixed / all fake 1/sarcasm 0/all misinformation&disinformation 0
264 if y1 == [1] and y2 == [0] and y3 == [0]:
265 x1 = [-3,0]
266 y1 = [-3,0]
267 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'
268
       plt.plot(x1, y1, label = "Fake News", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
269
       plt.plot(-3, -3, marker="0", markersize=20, markeredgecolor="red", markerfacecolor="red")
       plt.text(-3, -3, 'Fake News')
270
271 x2 = [3,0]
272
       y2 = [5,0]
273
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
274
       plt.plot(x2, y2, label = "Direct Talk", color='green') # , **marker style, **text style) # linestyle="-", color='red'
275
       plt.plot(3, 5, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
       plt.text(3, 5, 'Direct Talk')
276
277
       x3 = [3,0]
278
       y3 = [7,0]
279
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
280
       plt.plot(x3, y3, label = "Stance Neurtral", color='green') # , **marker style, **text style) # linestyle="-", color='red'
281
       plt.plot(3, 7, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
282
       plt.text(3, 7, 'Stance_Neutral')
283
284
285
286 plt.xlim(-4, 6)
287 plt.ylim(-8, 9)
288 plt.gca().set_xticklabels([])
289 plt.gca().set_yticklabels([])
290
291 # y1 = [0]
292 # y2 = [0]
293 # y3 = [1]
294 # Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1
```

295 if y1 == [0] and y2 == [0] and y3 == [1]:

and parigra().see\_areasances([]) 289 plt.gca().set yticklabels([]) 290 291 # v1 = [0] 292 # y2 = [0] 293 # v3 = [1] 294 # Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1 295 if y1 == [0] and y2 == [0] and y3 == [1]: 296 x1 = [3.0] 297 y1 = [3,0] 298 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 299 plt.plot(x1, y1, label = "True News", color='green') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 300 plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green") 301 plt.text(3, 3, 'True News') 302 x2 = [3,0]303  $y_2 = [5,0]$ 304 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red'</pre> 305 plt.plot(x2, y2, label = "Direct Talk", color='green') # , \*\*marker style, \*\*text\_style) # linestyle="-", color='red' 306 plt.plot(3, 5, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green") 307 plt.text(3, 5, 'Direct Talk') 308 x3 = [-3,0] 309 y3 = [-7,0] 310 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red'</pre> 311 plt.plot(x3, y3, label = "Stance Misinformation", color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red' 312 plt.plot(-3, -7, marker="0", markersize=20, markeredgecolor="red", markerfacecolor="red") 313 plt.text(-3, -7, 'Stance Misinformation') 314 315 316 # \*\*\* NEW for 0,1,2 single feature for stance column \*\*\* 317 # y1 = [0] 318 # y2 = [0] 319 # y3 = [2]320 # Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1 321 if y1 == [0] and y2 == [0] and y3 == [2]: 322 x1 = [3,0] 323 y1 = [3,0] 324 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker style, \*\*text style) # linestyle="-", color='red'</pre> 325 plt.plot(x1, y1, label = "True News", color='green') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 326 plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green") 327 plt.text(3, 3, 'True News') 328 x2 = [3,0] 329 y2 = [5,0] 330 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red'</pre> 331 plt.plot(x2, y2, label = "Direct Talk", color='green') # , \*\*marker\_style, \*\*text\_style) # linestyle="-", color='red' 332 plt.plot(3, 5, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green") 333 plt.text(3, 5, 'Direct Talk')

```
332
       plt.plot(3, 5, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
333
       plt.text(3, 5, 'Direct Talk')
334 x3 = [-3,0]
335
       v3 = [-9.0]
336
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') ₩ , **marker style, **text style) ₩ linestyle="-", color='red'</pre>
       plt.plot(x3, y3, label = "Stance Disinformation", color='red') # , **marker style, **text style) # linestyle="-", color='red'
337
338
       plt.plot(-3, -9, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
339
       plt.text(-3, -9, 'Stance Disinformation')
340
341 # y1 = [0]
342 # y2 = [1]
343 # y3 = [2]
344 # Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1
345 if y1 == [0] and y2 == [1] and y3 == [2]:
346 x1 = [3,0]
347 y1 = [3.0]
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
348
       plt.plot(x1, y1, label = "True News", color='green') # , **marker_style, **text_style) # linestyle="-", color='red'
349
350
       plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
351
       plt.text(3, 3, 'True News')
352 x2 = [-3,0]
353
       v2 = [-5,0]
354
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
355
       plt.plot(x2, y2, label = "Sarcasm Talk", color='red') # , **marker style, **text style) # linestyle="-", color='red'
       plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
356
357
       plt.text(-3, -5, 'Sarcasm Talk')
358
       x3 = [-3,0]
359
       v3 = [-9.0]
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
360
       plt.plot(x3, y3, label = "Stance Disinformation", color='red') # , **marker style, **text style) # linestyle="-", color='red'
361
362
       plt.plot(-3, -9, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
       plt.text(-3, -9, 'Stance Disinformation')
363
364
365 # y1 = [1]
366 # v2 = [1]
367 # y3 = [2]
368 # Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1
369 if y1 == [1] and y2 == [1] and y3 == [2]:
370 x1 = [-3,0]
```

370	x1 = [-3, 0]
371	$y_1 = (-3, 0)$
372	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
373	plt.plot(x1, y1, label = "Fake News", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
374	plt.plot(-3, -3, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
375	plt.text(-3, -3, 'Fake News')
376	$x^2 = (-3, 0)$
377	$y_2 = [-5, 0]$
378	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
379	<pre>plt.plot(x2, y2, label = "Sarcasm Talk", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
380	plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
381	plt.text(-3, -5, 'Sarcasm Talk')
382	$x_3 = [-3, 0]$
383	y3 = [-9,0]
384	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
385	<pre>plt.plot(x3, y3, label = "Stance Disinformation", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
386	<pre>plt.plot(-3, -9, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")</pre>
387	<pre>plt.text(-3, -9, 'Stance Disinformation')</pre>
388	
389	# y1 = [1]
390	# y2 = [0]
391	# y3 = [2]
392	# Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1
393	if y1 == [1] and y2 == [0] and y3 == [2]:
394	x1 = [-3,0]
395	y1 = [-3,0]
396	<pre>#plt.plot(x1, y1, label = "line 1", marker='&lt;', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
397	<pre>plt.plot(x1, y1, label = "Fake News", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
398	
399	
400	x2 = [3, 0]
401	y2 = [5,0]
402	
403	
404	
405	<pre>plt.text(3, 5, 'Direct Talk')</pre>
406	x3 = [-3,0]

```
405
       plt.text(3, 5, 'Direct Talk')
406 x3 = [-3,0]
407 V3 = [-9.0]
408 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'
       plt.plot(x3, y3, label = "Stance Disinformation", color='red') # , **marker style, **text style) # linestyle="-", color='red'
409
410
       plt.plot(-3, -9, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
411 plt.text(-3, -9, 'Stance Disinformation')
412
413 # y1 = [0]
414 # y2 = [1]
415 # v3 = [2]
416 # Mixed / all fake 0/sarcasm 0/all misinformation&disinformation 1
417 if y1 == [0] and y2 == [1] and y3 == [2]:
418 x1 = [3,0]
419
       v1 = [3.0]
420 #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
421
       plt.plot(x1, y1, label = "True News", color='green') # , **marker style, **text style) # linestyle="-", color='red'
422
       plt.plot(3, 3, marker="o", markersize=20, markeredgecolor="green", markerfacecolor="green")
423
       plt.text(3, 3, 'True News')
424 x2 = [-3,0]
425
       y2 = [-5, 0]
426
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker style, **text style) # linestyle="-", color='red'</pre>
427
       plt.plot(x2, y2, label = "Sarcasm Talk", color='red') # , **marker style, **text style) # linestyle="-", color='red'
428
       plt.plot(-3, -5, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
429
       plt.text(-3, -5, 'Sarcasm Talk')
430 x3 = [-3,0]
431
       y3 = [-9,0]
432
       #plt.plot(x1, y1, label = "line 1", marker='<', markersize=20, color='red') # , **marker_style, **text_style) # linestyle="-", color='red'</pre>
433
       plt.plot(x3, y3, label = "Stance Disinformation", color='red') # , **marker_style, **text_style) # linestyle="-", color='red'
434
       plt.plot(-3, -9, marker="o", markersize=20, markeredgecolor="red", markerfacecolor="red")
435 plt.text(-3, -9, 'Stance Disinformation')
436
437 plt.legend(loc='lower right')
438
439 from matplotlib import pyplot as plt
448 plt.savefig('/content/drive/My_Drive/Colab_Notebooks/MSCAITOP_MSCAITOP_graphic_v2b.png', bbox_inches='tight')
441
442 plt.show()
443
444 # 11, = ax.plot([0.1, 0.5, 0.9], [0.1, 0.9, 0.5], "bo-",
```

# References

- 1 https://www.kaggle.com/code/paramarthasengupta/fake-news-detector-eda-prediction-99
- 2 https://www.kaggle.com/datasets/deepnews/fakenews-reddit-comments/data
- 3 https://www.kaggle.com/datasets/arashnic/7-nlp-tasks-with-tweets