```
In [1]: import numpy as np # linear algebra
          import pandas as pd
In [2]: # dowload dataset
          fake = pd.read csv('ISOT Dataset/Fake.csv')
          true = pd.read csv("ISOT Dataset/True.csv")
         # Add flag to track fake and real
In [3]:
          fake['target'] = 1
          true['target'] = 0
In [4]: data = pd.concat([fake, true]).reset index(drop = True)
         data.shape
Out[4]: (44898, 5)
In [5]: from sklearn.utils import shuffle
         data = shuffle(data)
         data = data.reset index(drop=True)
In [6]: data.head()
Out[6]:
                                 title
                                                                  subject
                                                          text
                                                                                 date target
             Republican ex-Treasury chief
                                        WASHINGTON (Reuters) -
                                                                              June 25,
                                                               politicsNews
                                                                                          0
                    Paulson slams Tru...
                                        Henry Paulson, a Republ...
                                                                                2016
              SHOCK POLL In MUST WIN
                                         Apparently the Black Lives
                                                                           Jul 11, 2016
                                                                 left-news
                                                                                          1
               State Of FLORIDA: Hispa...
                                             Matter terror group...
                   MEDALS OF VALOR:
                                       It s great to have a president
          2
                President Trump Honored
                                                                   politics
                                                                           Jul 27, 2017
                                                                                          1
                                               who appreciates...
                              Agent...
              Newsweek Just Made Their
                                       Newsweek has never been a
                                                                          November 9,
          3
                                                                                          1
                                                                    News
                 BEST Cover Ever And ...
                                             publication to shy a ...
                                                                                2017
                  Trump says he believes
                                        WASHINGTON (Reuters) -
                                                                           October 16,
                                                               politicsNews
                                                                                          0
                 Cuba responsible for at...
                                        President Donald Trump ...
                                                                                2017
In [7]: ## Data Preparation
         data = data[data['text'].notna()]
          data = data[data['title'].notna()]
         data = data[data['subject'].notna()]
In [8]: import matplotlib.pyplot as plt
          import seaborn as sns
          import nltk
          nltk.download('stopwords')
          nltk.download('wordnet')
```

```
Inltk datal Downloading nackage stonwords to
Out[8]: True
In [9]: # Let's do some statistics of the text columns
         txt len = data.text.str.split().str.len()
         txt_len.describe()
Out[9]: count
                  44898.000000
         mean
                     405.282284
         std
                     351.265595
         min
                       0.000000
         25%
                     203.000000
         50%
                     362.000000
         75%
                     513.000000
         max
                    8135.000000
         Name: text, dtype: float64
In [10]: # Let's do some statistics of the title columns
         title len = data.title.str.split().str.len()
         title len.describe()
Out[10]: count
                  44898.000000
         mean
                      12.453472
         std
                       4.111476
         min
                       1.000000
         25%
                      10.000000
         50%
                      11.000000
         75%
                      14.000000
                      42.000000
         Name: title, dtype: float64
In [11]: # Class Distribution
         # 1: Unreliable
         # 2: Reliable
         sns.countplot(x='target', data= data)
Out[11]: <AxesSubplot:xlabel='target', ylabel='count'>
            20000
            15000
            10000
             5000
               0
                           Ó
                                               1
                                    target
In [12]: print(data.target.value_counts())
         print()
         print(round(data.target.value counts(normalize=True),2)*100)
```

```
23481
          1
          0
               21417
          Name: target, dtype: int64
               52.0
          1
          0
               48.0
In [13]: data.isnull().sum()
Out[13]: title
                      0
                      0
          text
          subject
                      0
                      0
          date
          target
                      0
          dtype: int64
In [14]: column n = ['date', 'title', 'subject', 'text', 'target']
          remove c = ['subject','date']
          categorical features = []
          target_col = ['target']
          text f = ['title', 'text']
In [15]: # cleaning
          import nltk
         from nltk.corpus import stopwords
         import re
          from nltk.stem.porter import PorterStemmer
         from collections import Counter
         ps = PorterStemmer()
         wnl = nltk.stem.WordNetLemmatizer()
          stop words = stopwords.words('english')
          stopwords dict = Counter(stop words)
         # remove unused columns
         def remove unused c(df, column n=remove c):
              df = df.drop(column_n, axis=1)
              return df
          # impute null values with none
          def null process(feature df):
              for col in text f:
                  feature_df.loc[feature df[col].isnull(),col] = "None"
              return feature df
         # clean data
          def clean dataset(df):
              # remove unused column
              df = remove unused c(df)
              #impute null value
              df = null process(df)
              return df
         # Cleaning text from unused characters
         def clean text(text):
              text = str(text).replace(r'http[\w:/\.]+', ' ') # removing urls
text = str(text).replace(r'[^\.\w\s]', ' ') # remove everything
```

```
text = str(text).replace('[^a-zA-Z]', ' ')
    text = str(text).replace(r'\s\s+', ' ')
    text = text.lower().strip()
    #text = ' '.join(text)
    return text

## Nltk Preprocessing include:
# Stop words, Stemming and Lemmetization
# For our project we use only Stop word removal
def nltk_preprocess(text):
    text = clean_text(text)
    wordlist = re.sub(r'[^\w\s]', '', text).split()
    text = ' '.join([wnl.lemmatize(word) for word in wordlist if word
    return text

In [16]:

df = clean_dataset(data)
df['text'] = df.text.apply(nltk_preprocess)
df['title'] = df.title.apply(nltk_preprocess)
```

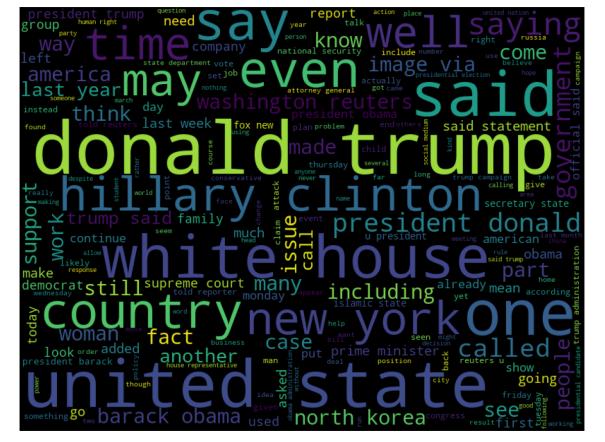
In [17]: df.head()

Out[17]:

target	text	title	
0	washington reuters henry paulson republican u	republican extreasury chief paulson slam trump	0
1	apparently black life matter terror group mana	shock poll must win state florida hispanic tur	1
1	great president appreciates special agent poli	medal valor president trump honored agent offi	2
1	newsweek never publication shy away controvers	newsweek made best cover ever people freaking	3
0	washington reuters president donald trump said	trump say belief cuba responsible attack hurt	4

```
In [18]: from wordcloud import WordCloud, STOPWORDS

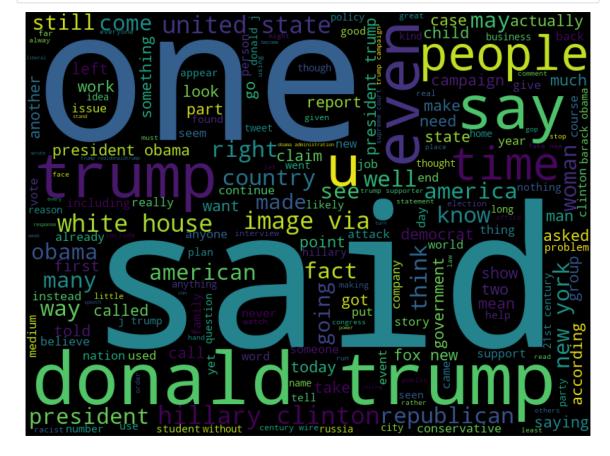
# initialize the word cloud
wordcloud = WordCloud(background_color='black', width=800, height=600
# generate the word cloud
text_cloud = wordcloud.generate(" ".join(df['text']))
# plotting the word cloud
plt.figure(figsize=(20,30))
plt.imshow(text_cloud)
plt.axis('off')
plt.show()
```

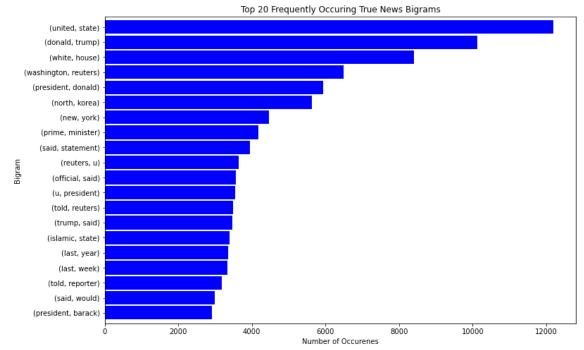


```
In [19]: # reliable news (0)
    reliable_news = " ".join(df[df['target']==0]['text'])
    wc = wordcloud.generate(reliable_news)
    plt.figure(figsize=(20,30))
    plt.imshow(wc)
    plt.axis('off')
    plt.show()
```

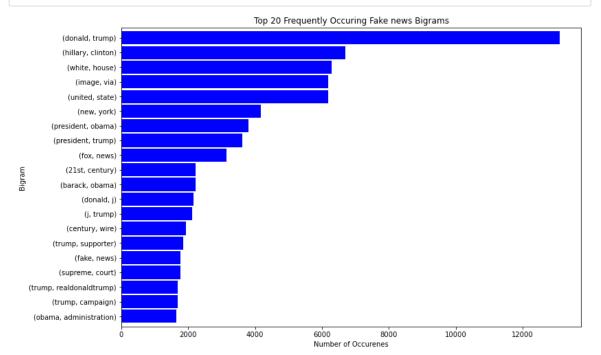
```
said statement president already much donate made clinton expected congress true already possible works told reuters support with the said trump and already problem of the said trump of the sa
```

```
In [20]: # unreliable news (1)
    unreliable_news = ' '.join(df[df['target']==1]['text'])
    wc= wordcloud.generate(unreliable_news)
    plt.figure(figsize=(20,30))
    plt.imshow(wc)
    plt.axis('off')
    plt.show()
```

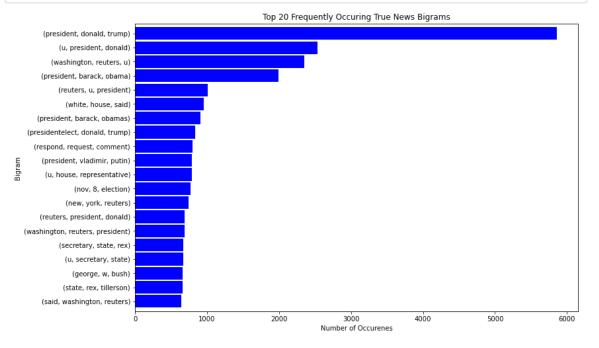




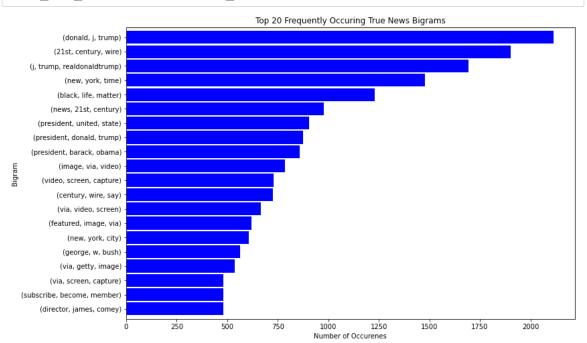
In [22]: plot_top_ngrams(unreliable_news, 'Top 20 Frequently Occuring Fake new



In [23]: # Trigram
plot_top_ngrams(reliable_news, "Top 20 Frequently Occuring True News



In [24]: plot_top_ngrams(unreliable_news, "Top 20 Frequently Occuring True New



In [25]: !pip install transformers

```
Requirement already satisfied: transformers in /home/administrator/
anaconda3/lib/python3.9/site-packages (4.18.0)
Requirement already satisfied: sacremoses in /home/administrator/an
aconda3/lib/python3.9/site-packages (from transformers) (0.0.49)
Requirement already satisfied: regex!=2019.12.17 in /home/administr
ator/anaconda3/lib/python3.9/site-packages (from transformers) (202
1.8.3)
Requirement already satisfied: packaging>=20.0 in /home/administrat
or/anaconda3/lib/python3.9/site-packages (from transformers) (21.0)
Requirement already satisfied: tqdm>=4.27 in /home/administrator/an
aconda3/lib/python3.9/site-packages (from transformers) (4.62.3)
Requirement already satisfied: requests in /home/administrator/anac
onda3/lib/python3.9/site-packages (from transformers) (2.26.0)
Requirement already satisfied: tokenizers!=0.11.3,<0.13,>=0.11.1 in
/home/administrator/anaconda3/lib/python3.9/site-packages (from tra
nsformers) (0.12.1)
Requirement already satisfied: numpy>=1.17 in /home/administrator/a
naconda3/lib/python3.9/site-packages (from transformers) (1.20.3)
Requirement already satisfied: pyyaml>=5.1 in /home/administrator/a
naconda3/lib/python3.9/site-packages (from transformers) (6.0)
Requirement already satisfied: huggingface-hub<1.0,>=0.1.0 in /home
/administrator/anaconda3/lib/python3.9/site-packages (from transfor
mers) (0.5.1)
Requirement already satisfied: filelock in /home/administrator/anac
onda3/lib/python3.9/site-packages (from transformers) (3.3.1)
Requirement already satisfied: typing-extensions>=3.7.4.3 in /home/
administrator/anaconda3/lib/python3.9/site-packages (from huggingfa
ce-hub<1.0,>=0.1.0->transformers) (3.10.0.2)
Requirement already satisfied: pyparsing>=2.0.2 in /home/administra
tor/anaconda3/lib/python3.9/site-packages (from packaging>=20.0->tr
ansformers) (3.0.4)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in /home/admin
istrator/anaconda3/lib/python3.9/site-packages (from requests->tran
sformers) (1.26.7)
Requirement already satisfied: charset-normalizer~=2.0.0 in /home/a
dministrator/anaconda3/lib/python3.9/site-packages (from requests->
transformers) (2.0.4)
Requirement already satisfied: idna<4,>=2.5 in /home/administrator/
anaconda3/lib/python3.9/site-packages (from requests->transformers)
(3.2)
                      . . . . .
                                  . . . .
```

```
import torch
In [26]:
         from transformers.file utils import is tf available, is torch available
         from transformers import BertTokenizerFast, BertForSequenceClassification
         from transformers import Trainer, TrainingArguments
         from sklearn.model selection import train test split
         import random
```

```
In [27]: import tensorflow as tf
         with tf.device('GPU:1'):
             def set seed(seed: int):
                 Helper function for reproducible behavior to set the seed in
                 installed).
                 Args:
                      seed (:obj:`int`): The seed to set.
                 random.seed(seed)
```

```
np.random.seed(seed)
if is_torch_available():
    torch.manual_seed(seed)
    torch.cuda.manual_seed_all(seed)
    # ^^ safe to call this function even if cuda is not avail
if is_tf_available():
    import tensorflow as tf

    tf.random.set_seed(seed)

set_seed(123)
2022-05-16 14:38:42.403950: I tensorflow/core/platform/cpu_feature_
```

guard.cc:151] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to use the following CPU instructio ns in performance-critical operations: AVX2 AVX512F FMA To enable them in other operations, rebuild TensorFlow with the app ropriate compiler flags.

2022-05-16 14:38:43.367928: I tensorflow/core/common_runtime/gpu/gp u_device.cc:1532] Created device /job:localhost/replica:0/task:0/de vice:GPU:0 with 14637 MB memory: -> device: 0, name: Tesla V100-PC IE-16GB, pci bus id: 0000:3b:00.0, compute capability: 7.0

2022-05-16 14:38:43.368594: I tensorflow/core/common_runtime/gpu/gp u_device.cc:1532] Created device /job:localhost/replica:0/task:0/de vice:GPU:1 with 14637 MB memory: -> device: 1, name: Tesla V100-PC IE-16GB, pci bus id: 0000:d8:00.0, compute capability: 7.0

```
In [28]: with tf.device('GPU:1'):
    model_name = "bert-base-uncased"
    max_length= 512
```

In [29]: with tf.device('GPU:1'):
 tokenizer = BertTokenizerFast.from_pretrained(model_name, do_lower)

In [30]: data.head()

Out[30]:

	title	text	subject	date	target
0	Republican ex-Treasury chief Paulson slams Tru	WASHINGTON (Reuters) - Henry Paulson, a Republ	politicsNews	June 25, 2016	0
1	SHOCK POLL In MUST WIN State Of FLORIDA: Hispa	Apparently the Black Lives Matter terror group	left-news	Jul 11, 2016	1
2	MEDALS OF VALOR: President Trump Honored Agent	It s great to have a president who appreciates	politics	Jul 27, 2017	1
3	Newsweek Just Made Their BEST Cover Ever And	Newsweek has never been a publication to shy a	News	November 9, 2017	1
4	Trump says he believes Cuba responsible for at	WASHINGTON (Reuters) - President Donald Trump	politicsNews	October 16, 2017	0

```
In [31]: with tf.device('GPU:1'):
    ## Data Preparation
    data = data[data['text'].notna()]
```

```
In [33]: with tf.device('GPU:1'):
    def prepare_data(df, test_size=0.2, include_title=True, include_a
```

```
texts = []
                 labels = []
                 for i in range(len(df)):
                     text = df['text'].iloc[i]
                     label = df['target'].iloc[i]
                     if text and label in [0,1]:
                         texts.append(text)
                         labels.append(label)
                 return train test split(texts, labels, test size=test size)
             train texts, valid texts, train labels, valid labels = prepare da
In [34]: print(len(train_texts), len(train_labels))
         print(len(valid texts), len(valid labels))
         35918 35918
         8980 8980
In [35]: with tf.device('GPU:1'):
             # tokenizing the dataset
             train encodings = tokenizer(train texts, truncation=True, padding
             valid encodings = tokenizer(valid texts, truncation=True, padding
In [36]: with tf.device('GPU:1'):
             # converting the encoding into a PyTorch datset
             class NewsGroupsDataset(torch.utils.data.Dataset):
                 def init (self, encodings, labels):
                     self.encodings = encodings
                     self.labels = labels
                 def getitem (self, idx):
                     item = {k: torch.tensor(v[idx]) for k, v in self.encoding
                     item['labels'] = torch.tensor([self.labels[idx]])
                     return item
                 def len (self):
                     return len(self.labels)
             # convert tokenize data into torch dataset
             train dataset = NewsGroupsDataset(train encodings, train labels)
             valid dataset = NewsGroupsDataset(valid encodings, valid labels)
In [37]: |with tf.device('GPU:1'):
             model = BertForSequenceClassification.from pretrained(model name)
```

```
Some weights of the model checkpoint at bert-base-uncased were not
         used when initializing BertForSequenceClassification: ['cls.seq rel
         ationship.bias', 'cls.predictions.transform.dense.bias', 'cls.predi
         ctions.decoder.weight', 'cls.predictions.transform.LayerNorm.bias',
         'cls.seq_relationship.weight', 'cls.predictions.transform.dense.weight', 'cls.predictions.bias', 'cls.predictions.transform.LayerNorm.
         weight']
         - This IS expected if you are initializing BertForSequenceClassific
         ation from the charkmaint of a model trained on another tack or wit
In [38]: with tf.device('GPU:1'):
             from sklearn.metrics import precision recall fscore support
             from sklearn.metrics import accuracy score
             def computer metrics(pred):
                  labels = pred.label ids
                  preds = pred.predictions.argmax(-1)
                  precision, recall, f1, = precision recall fscore support(la
                  acc = accuracy score(labels, preds)
                  return {
                      'accuracy': acc,
                      'f1': f1,
                      'precision': precision,
                      'recall': recall
                  }
In [39]: with tf.device('GPU:1'):
             training args = TrainingArguments(
                  output dir='./results',
                                                    # output directory
                                                    # total number of training &
                  num_train_epochs=1,
                  per device train batch size=10, # batch size per device duri
                  per device eval batch size=20, # batch size for evaluation
                                                   # number of warmup steps for
                  warmup steps=100,
                  logging_dir='./logs',
                                                    # directory for storing logs
                  load best model at end=True, # load the best model when i
                  # but you can specify `metric_for_best_model` argument to cha
                  logging steps=200,
                                                   # log & save weights each lo
                  save steps=200,
                  evaluation strategy="steps", # evaluate each `logging ste
             )
In [40]: with tf.device('GPU:1'):
             trainer = Trainer(
                      model = model,
                      args = training_args,
                      train dataset=train dataset,
                      eval dataset=valid dataset,
                      compute metrics=computer metrics,
                  )
In [41]: with tf.device('GPU:1'):
             trainer.train()
```

```
/home/administrator/anaconda3/lib/python3.9/site-packages/transform
         ers/optimization.py:306: FutureWarning: This implementation of Adam
         W is deprecated and will be removed in a future version. Use the Py
         Torch implementation torch.optim.AdamW instead, or set `no deprecat
         ion warning=True` to disable this warning
           warnings.warn(
         ***** Running training *****
           Num examples = 35918
           Num Epochs = 1
           Instantaneous batch size per device = 10
           Total train batch size (w. parallel, distributed & accumulation)
In [46]: with tf.device('GPU:1'):
             # evaluate the current model after training
             trainer.evaluate()
         ***** Running Evaluation *****
           Num examples = 8980
           Batch size = 20
         /home/administrator/anaconda3/lib/python3.9/site-packages/torch/nn/
         parallel/ functions.py:68: UserWarning: Was asked to gather along d
         imension \overline{0}, but all input tensors were scalars; will instead unsque
         eze and return a vector.
           warnings.warn('Was asked to gather along dimension 0, but all '
         Attempted to log scalar metric eval loss:
         9.067665814654902e-05
         Attempted to log scalar metric eval accuracy:
         1.0
         Attempted to log scalar metric eval f1:
         Attempted to log scalar metric eval precision:
         Attempted to log scalar metric eval recall:
         1.0
         Attempted to log scalar metric eval runtime:
         56.0933
         Attempted to log scalar metric eval samples per second:
         160.09
         Attempted to log scalar metric eval steps per second:
         Attempted to log scalar metric epoch:
         1.0
In [43]: with tf.device('GPU:1'):
         # saving the fine tuned model & tokenizer
             model path = "fake-news-bert-base-uncased"
             model.save pretrained(model path)
             tokenizer.save pretrained(model path)
         Configuration saved in fake-news-bert-base-uncased/config.json
         Model weights saved in fake-news-bert-base-uncased/pytorch model.bi
         tokenizer config file saved in fake-news-bert-base-uncased/tokenize
         r config.json
         Special tokens file saved in fake-news-bert-base-uncased/special to
         kens map.json
In [44]: def get prediction(text, convert to label=False):
```

```
# prepare our text into tokenized sequence
             inputs = tokenizer(text, padding=True, truncation=True, max lengt
             # perform inference to our model
             outputs = model(**inputs)
             # get output probabilities by doing softmax
             probs = outputs[0].softmax(1)
             # executing argmax function to get the candidate label
             d = {
                 0: "reliable",
                 1: "fake"
             if convert to label:
                 return d[int(probs.argmax())]
             else:
                 return int(probs.argmax())
In [45]: real news = """
          Donald Trump Sends Out Embarrassing New Year's Eve Message; This is
         get prediction(real news, convert to label=True)
Out[45]: 'fake'
 In [ ]:
 In [ ]:
In [ ]:
 In [ ]:
 In [ ]:
 In [ ]:
```