PHASE-4: DEVELOPMENT Part-2

Project Title: Fake News Detection

Problem Definition:

The problem is to develop a fake news detection model using a Kaggle dataset. The goal is to distinguish between genuine and fake news articles based on their titles and text. This project involves using natural language processing (NLP) techniques to preprocess the text data, building a machine learning model for classification, and evaluating the model's performance.

Dataset Link:

https://www.kaggle.com/datasets/clmentbisaillon/fake-and-real-news-dataset

Introduction:

Creating a fake news detection model involves several steps, including text preprocessing, feature extraction, model training, and evaluation.

Project Design Steps:

1. Data Source:

Choose the fake news dataset available on Kaggle, containing articles titles and text, along with their labels (genuine or fake). This dataset should represent various topics and sources to ensure the model's generalizability.

2. Data Preprocessing:

Clean and preprocess the textual data to prepare it for analysis. This may involve tasks such as tokenization, lowercasing, removing stop words, and stemming/lemmatization.

3. Feature Extraction:

Convert the text data into numerical features that can be used as input for machine learning algorithms. Common techniques include:

- TF-IDF (Term Frequency-Inverse Document Frequency): Capturing the importance of words in a document relative to a corpus.
- Word Embeddings: Representing words as dense vectors using models like Word2Vec, GloVe, or BERT embeddings.
- N-grams: Capturing sequences of words to capture context

4. Model Selection:

Choose an appropriate machine learning or deep learning model for fake news detection. Common choices include logistic regression, Naive Bayes, Support Vector Machines (SVM), recurrent neural networks (RNNs), or transformers like BERT.

5. Model Training:

Train the selected model using the preprocessed data, using labeled examples to learn the patterns associated with real and fake news. The model learns to distinguish between real and fake news based on the provided features.

6. Evaluation:

Evaluate the model's performance using metrics like accuracy, precision, recall, F1-score, and ROC-AUC. It's essential to consider the class imbalance issue, as fake news might be a minority class

Installation steps and the Program:

```
import numpy as np
import pandas as pd
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
  for filename in filenames:
    print(os.path.join(dirname, filename))
!pip install genism
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
import seaborn as sns
```

import gensim

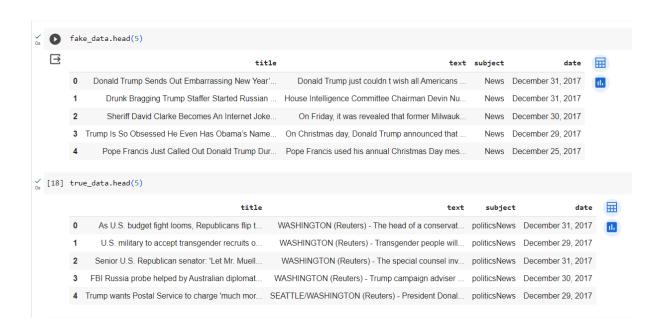
```
from gensim.utils import simple_preprocess
from gensim.parsing.preprocessing import STOPWORDS
import plotly.express as px
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import roc_auc_score
from sklearn.metrics import confusion_matrix
fake_data = pd.read_csv('/kaggle/input/fake-and-real-news-dataset/Fake.csv')
print("fake_data",fake_data.shape)
true_data= pd.read_csv('/kaggle/input/fake-and-real-news-dataset/True.csv')
print("true_data",true_data.shape)
fake_data.head(5)
true_data.head(5)
true_data['target'] = 1
fake data['target'] = 0
df = pd.concat([true_data, fake_data]).reset_index(drop = True)
df['original'] = df['title'] + ' ' + df['text']
df.head()
df.isnull().sum()
stop_words = stopwords.words('english')
stop_words.extend(['from', 'subject', 're', 'edu', 'use'])
def preprocess(text):
  result = []
  for token in gensim.utils.simple_preprocess(text):
    if token not in gensim.parsing.preprocessing.STOPWORDS and len(token) > 2 and token not in
stop_words:
      result.append(token)
  return result
df.subject=df.subject.replace({'politics':'PoliticsNews','politicsNews':'PoliticsNews'})
sub_tf_df=df.groupby('target').apply(lambda x:x['title'].count()).reset_index(name='Counts')
sub_tf_df.target.replace({0:'False',1:'True'},inplace=True)
fig = px.bar(sub_tf_df, x="target", y="Counts",
```

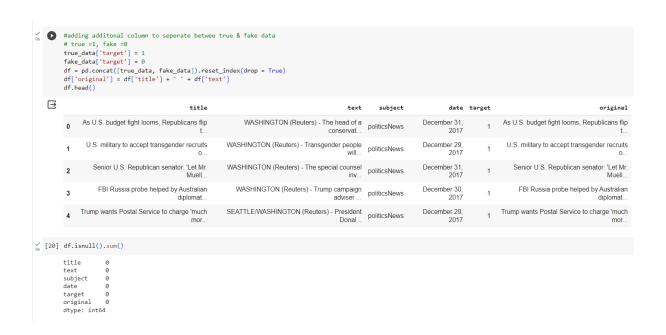
```
color='Counts', barmode='group',
       height=350)
fig.show()
sub_check=df.groupby('subject').apply(lambda x:x['title'].count()).reset_index(name='Counts')
fig=px.bar(sub_check,x='subject',y='Counts',color='Counts',title='Count of News Articles by Subject')
fig.show()
df['clean_title'] = df['title'].apply(preprocess)
df['clean_title'][0]
df['clean_joined_title']=df['clean_title'].apply(lambda x:" ".join(x))
plt.figure(figsize = (20,20))
wc = WordCloud(max_words = 2000, width = 1600, height = 800, stopwords =
stop_words).generate(" ".join(df[df.target == 1].clean_joined_title))
plt.imshow(wc, interpolation = 'bilinear')
maxlen = -1
for doc in df.clean_joined_title:
  tokens = nltk.word_tokenize(doc)
  if(maxlen<len(tokens)):</pre>
    maxlen = len(tokens)
print("The maximum number of words in a title is =", maxlen)
fig = px.histogram(x = [len(nltk.word tokenize(x)) for x in df.clean joined title], nbins = 50)
fig.show()
X train, X test, y train, y test = train test split(df.clean joined title, df.target, test size =
0.2, random state=2)
vec_train = CountVectorizer().fit(X_train)
X vec train = vec train.transform(X train)
X_vec_test = vec_train.transform(X_test)
model = LogisticRegression(C=2)
model.fit(X vec train, y train)
predicted_value = model.predict(X_vec_test)
accuracy_value = roc_auc_score(y_test, predicted_value)
print(accuracy value)
cm = confusion_matrix(list(y_test), predicted_value)
plt.figure(figsize = (7, 7))
```

```
sns.heatmap(cm, annot = True,fmt='g',cmap='viridis')
df['clean_text'] = df['text'].apply(preprocess)
df['clean_joined_text']=df['clean_text'].apply(lambda x:" ".join(x))
plt.figure(figsize = (20,20))
wc = WordCloud(max words = 2000, width = 1600, height = 800, stopwords =
stop_words).generate(" ".join(df[df.target == 1].clean_joined_text))
plt.imshow(wc, interpolation = 'bilinear')
maxlen = -1
for doc in df.clean_joined_text:
  tokens = nltk.word_tokenize(doc)
  if(maxlen<len(tokens)):</pre>
    maxlen = len(tokens)
print("The maximum number of words in a News Content is =", maxlen)
fig = px.histogram(x = [len(nltk.word_tokenize(x)) for x in df.clean_joined_text], nbins = 50)
fig.show()
X_train, X_test, y_train, y_test = train_test_split(df.clean_joined_text, df.target, test_size =
0.2,random_state=2)
vec_train = CountVectorizer().fit(X_train)
X_vec_train = vec_train.transform(X_train)
X_vec_test = vec_train.transform(X_test)
model = LogisticRegression(C=2.5)
model.fit(X_vec_train, y_train)
predicted_value = model.predict(X_vec_test)
accuracy_value = roc_auc_score(y_test, predicted_value)
print(accuracy value)
prediction = []
for i in range(len(predicted_value)):
  if predicted value[i].item() > 0.5:
    prediction.append(1)
  else:
    prediction.append(0)
cm = confusion_matrix(list(y_test), prediction)
plt.figure(figsize = (6, 6))
```

Import the data:

```
import numpy as np # linear algebra
       import pandas as pd
                                                                       + Code + Text
[7] import os
       for dirname, _, filenames in os.walk('<u>/kaggle/input</u>'):
           for filename in filenames:
             print(os.path.join(dirname, filename))
[9] !pip install gensim # Gensim is an open-source library for unsupervised topic modeling and natural language processing
       import nltk
       nltk.download('punkt')
       Requirement already satisfied: gensim in /usr/local/lib/python3.10/dist-packages (4.3.2)
Requirement already satisfied: numpy>=1.18.5 in /usr/local/lib/python3.10/dist-packages (from gensim) (1.23.5)
       Requirement already satisfied: scipy>=1.7.0 in /usr/local/lib/python3.10/dist-packages (from gensim) (1.11.3)
       Requirement already satisfied: smart-open>=1.8.1 in /usr/local/lib/python3.10/dist-packages (from gensim) (6.4.0)
       [nltk_data] Downloading package punkt to /root/nltk_data...
[nltk_data] Unzipping tokenizers/punkt.zip.
  + Code + Text
    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
          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
/<sub>1s</sub> [16] # Importing data
          fake_data = pd.read_csv('/content/Fake.csv')
          print("fake_data", fake_data.shape)
          true_data = pd.read_csv('/content/True.csv')
          print("true_data", true_data.shape)
          fake_data (23481, 4)
          true data (21417, 4)
```





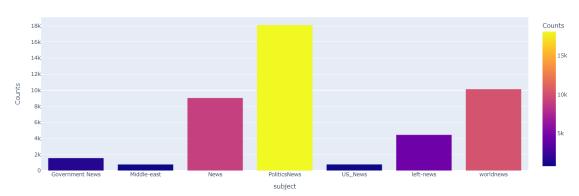
Data cleanup:

```
os import nltk
           nltk.download('stopwords')
      \begin{tabular}{ll} \hline \bf = & [nltk\_data] & Downloading package stopwords to /root/nltk\_data... \\ [nltk\_data] & Unzipping corpora/stopwords.zip. \\ \hline \end{tabular} 
           True
✓ [23] import nltk
           nltk.download('stopwords')
           from nltk.corpus import stopwords
           {\color{red}\mathsf{import}}\ {\color{gray}\mathsf{gensim}}
           from gensim.utils import simple_preprocess
           stop_words = stopwords.words('english')
stop_words.extend(['from', 'subject', 're', 'edu', 'use'])
           def preprocess(text):
                result = []
                for token in gensim.utils.simple_preprocess(text):
                      if token not in gensim.parsing.preprocessing.STOPWORDS and len(token) > 2 and token not in stop_words:
                          result.append(token)
                return result
           [nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Package stopwords is already up-to-date!

// [24] # Transforming the unmatching subjects to the same notation
df.subject=df.subject.replace({'politics':'PoliticsNews','politicsNews':'PoliticsNews'})
fig.show()
                20k
               15k
                                                                                                                                                                             22.5k
               10k
                 5k
                                                                                        target
```



Count of News Articles by Subject



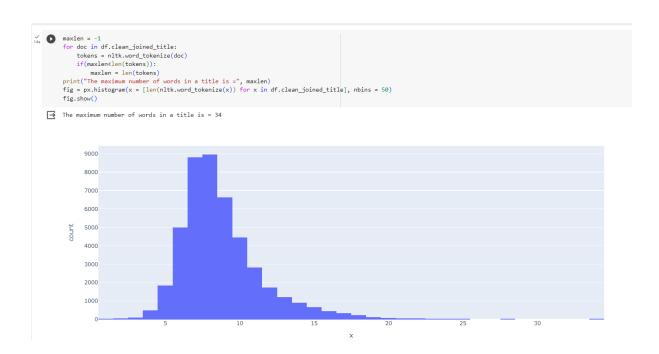
```
df['clean_title'] = df['title'].apply(preprocess)
df['clean_title'][0]

ightharpoonup [28] df['clean_joined_title']=df['clean_title'].apply(lambda x:" ".join(x))
```

```
plt.figure(figsize = (20,20))
wc = WordCloud(max_words = 2000 , width = 1600 , height = 800 , stopwords = stop_words).generate(" ".join(df[df.target == 1].clean_joined_title))
plt.imshow(wc, interpolation = 'bilinear')

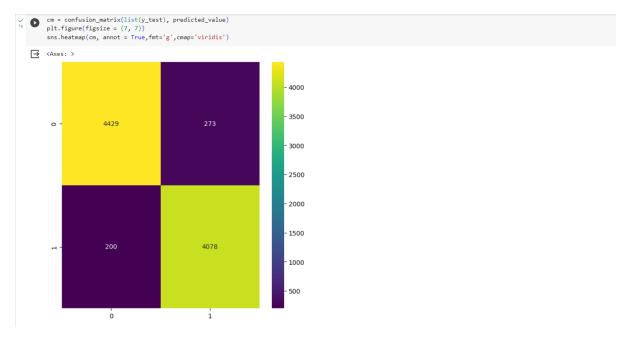
→ <matplotlib.image.AxesImage at 0x7adbcf47df30>

         100
                                                                                                   tell
                                                                                                                                                                     winslamic
                                                                                                                            many
                                                                               seek
         200
                                                                                                  party
                                                          nΦ
                                                                                                                                                        saudi
                                                                                deal
                                                                          democrat
                                                                                                                                                        russian
                                                                                                                               britain
               O
                                                                                                                                                               nator
                                                                                                                                                                fight
```



Creating Prediction Model:

Create the confusion matrix:

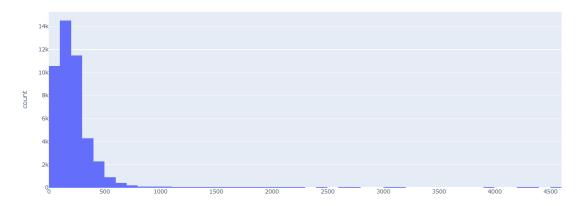


Checking the content of news:

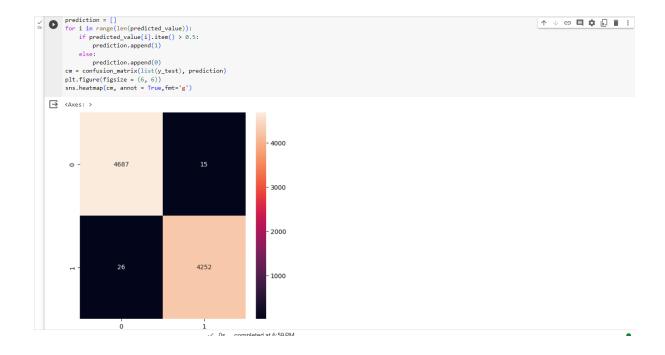
```
Saled Line to the control of the con
```

```
maxlen = -1
for doc in df.clean_joined_text:
    tokens = nltk.word_tokenize(doc)
    if(maxlenxlen(tokens)):
        maxlen = len(tokens)
print("The maximum number of words in a News Content is =", maxlen)
fig = px.histogram(x = [len(nltk.word_tokenize(x)) for x in df.clean_joined_text], nbins = 50)
fig.show()
```

☐ The maximum number of words in a News Content is = 4573



Predicting the Model:



Conclusion:

In this phase, our model's training and evaluation part is developed.