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In [ ]: # BERT with Keras
        import pandas as pd
        df balanced = pd.read csv("dataframe edit.tsv", sep = '\t')
        df.head()
        from sklearn.model_selection import train_test_split
        X_train, X_test, y_train, y_test = train_test_split(df_balanced['text'],df_b
                                                             stratify=df balanced['hy
         !pip install tensorflow text
        import tensorflow as tf
        import tensorflow hub as hub
        import tensorflow_text as text
        bert_preprocess = hub.KerasLayer("https://tfhub.dev/tensorflow/bert_en_uncas
        bert encoder = hub.KerasLayer("https://tfhub.dev/tensorflow/bert_en_uncased
         # DistilBERT
        #https://tfhub.dev/jeongukjae/distilbert en uncased L-6 H-768 A-12/1
        # Bert layers
        text input = tf.keras.layers.Input(shape=(), dtype=tf.string, name='text')
        preprocessed_text = bert_preprocess(text_input)
        outputs = bert encoder(preprocessed text)
        # Neural network layers
        1 = tf.keras.layers.Dropout(0.1, name="dropout")(outputs['pooled_output'])
        1 = tf.keras.layers.Dense(1, activation='sigmoid', name="output")(1)
        # Use inputs and outputs to construct a final model
        model = tf.keras.Model(inputs=[text input], outputs = [1])
        METRICS = [
              tf.keras.metrics.BinaryAccuracy(name='accuracy'),
              tf.keras.metrics.Precision(name='precision'),
              tf.keras.metrics.Recall(name='recall')
        1
        model.compile(optimizer='adam',
                       loss='binary crossentropy',
                       metrics=METRICS)
        model.fit(X_train, y_train, epochs=5)
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model.evaluate(X_test, y_test)

y_predicted = model.predict(X_test)
y_predicted = y_predicted.flatten()

import numpy as np

y_predicted = np.where(y_predicted > 0.5, 1, 0)
y_predicted

from sklearn.metrics import confusion_matrix, classification_report

cm = confusion_matrix(y_test, y_predicted)
cm

from matplotlib import pyplot as plt
import seaborn as sn
sn.heatmap(cm, annot=True, fmt='d')
plt.xlabel('Predicted')
plt.ylabel('Truth')
print(classification_report(y_test, y_predicted))
```

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In [ ]: ## **GPT-3 Zero shot**
        !pip install openai
        import os
        import openai
        import pandas as pd
        ds = pd.read_csv("dataframe_edit.tsv", sep = '\t')
        ds_65 = ds_{iail(65)}
        y_true = ds_65["hyperpartisan"].tolist()
        y_true
        ds_65["text"].head()
        len(ds 20)
        import os
        import openai
        OPENAI_API_KEY = "sk-8zJhEQJen18Qsq6WReemfj4Z3"
        openai.api_key = OPENAI_API_KEY
        start_sequence = "\nA:"
```

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restart sequence = "\n\nQ: "
alist= []
for i in ds_65["text"]:
 response = openai.Completion.create(
   model="text-davinci-003",
   prompt="Text: "+ i+ """\nQ: Does the above text contain hyperpartisan el
   Answer Only Yes or No.\nA:""",
   temperature=0,
   max tokens=100,
   top p=1,
   frequency penalty=0,
   presence penalty=0,
   stop=["\n"]
 alist.append(1 if response["choices"][0]["text"]==" Yes" else 0)
response["choices"][0]["text"]
type(alist[0])
# Classical ML
import sklearn.metrics as metrics
print(metrics.confusion matrix(y true, alist))
# Print the precision and recall, among other metrics
print(metrics.classification report(y true, alist, digits=3))
prompt="Text: "+ i+ """\nAdditional Information - Hyperpartisan argument is
prejudiced, or unreasoning allegiance to one party, faction, cause, or perso
information, does the above text contain hyperpartisan elements to it. Answe
prompt="Text: "+ i+ "\nQ: Does the above text contain hyperpartisan elements
import random
random list = [random.randint(0, 1) for in range(65)]
len(random list)
import sklearn.metrics as metrics
print(metrics.confusion matrix(y true,random list))
# Print the precision and recall, among other metrics
print(metrics.classification_report(y_true, random_list, digits=3))
```

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In [ ]: import pandas as pd
import numpy as np
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```
import collections
import csv
import os
import re
from nltk.tokenize import word tokenize
from sklearn.preprocessing import LabelEncoder
from collections import defaultdict
from nltk.corpus import wordnet as wn
from sklearn.feature extraction.text import TfidfVectorizer
from sklearn import model_selection, naive_bayes, svm
from sklearn.model selection import cross val score
from sklearn.metrics import accuracy_score, f1_score, precision_score, recal
from nltk import pos tag
from nltk.corpus import stopwords
from nltk.stem import WordNetLemmatizer
Corpus = pd.read csv("dataframe edit.tsv", sep = "\t", names = ['text', 'labe
Corpus.head()
df.groupby("label").count()
import nltk
nltk.download('wordnet')
# change it to str
Corpus.text = Corpus.text.astype(str)
Corpus['text'] = Corpus['text'].str.lower()
Corpus = Corpus.dropna()
Corpus.head()
Corpus.info()
#tokenizing our hyperpartisan text column here
Corpus['text'] = Corpus['text'].apply(nltk.word_tokenize)
# Tagging to understand if the word is a noun, verb, adverb etc
tag_map = defaultdict(lambda : wn.NOUN)
tag map['J'] = wn.ADJ
tag map['V'] = wn.VERB
tag_map['R'] = wn.ADV
import nltk
nltk.download('averaged perceptron tagger')
for index,entry in enumerate(Corpus['text']):
   Final_words = []
   word_Lemmatized = WordNetLemmatizer()
    for word, tag in pos_tag(entry):
```

# check for Stop words and consider only alphabets

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if word not in stopwords.words('english') and word.isalpha():
                    word Final = word Lemmatized.lemmatize(word, tag map[tag[0]])
                    Final_words.append(word_Final)
            # The final processed set of words for each iteration will be stored in
            Corpus.loc[index,'text_final'] = str(Final_words)
        Corpus.head()
        #Train, test split
        Train X, Test X, Train Y, Test Y = model selection.train test split(Corpus[
                                                                              Corpus[
        #Encoding our labels
        Encoder = LabelEncoder()
        Train Y = Encoder.fit transform(Train Y)
        Test Y = Encoder.fit transform(Test Y)
        Tfidf_vect = TfidfVectorizer()
        Tfidf_vect.fit(Corpus['text_final'])
        Train X Tfidf = Tfidf vect.transform(Train X)
        Test_X_Tfidf = Tfidf_vect.transform(Test_X)
        print(len(Tfidf vect.vocabulary ))
        # fit the NB classifier
        Naive = naive bayes.MultinomialNB()
        naive model = Naive.fit(Train X Tfidf,Train Y)
        predictions NB = Naive.predict(Test X Tfidf)
        print("Naive Bayes Accuracy Score -> ",accuracy score(predictions NB, Test Y
        f1_score(predictions_NB, Test_Y)
        #SVM classifier
        SVM = svm.SVC(C=2.0, kernel='poly',degree=2, gamma='scale')
        SVM.fit(Train_X_Tfidf,Train_Y)
        predictions SVM = SVM.predict(Test X Tfidf)
        print("SVM Accuracy Score -> ",accuracy_score(predictions_SVM, Test_Y)*100)
        f1 score(predictions SVM, Test Y)
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