Group Project: Detecting AI generated text

Kaggle Competition Link: https://www.kaggle.com/competitions/llm-detect-ai-generated-text

Goal: Classify essays into Al-written text and Student-written text.

Tasks performed:

- 1. Analysing the Data:
 - Remove punctuations
 - Analysing the length of essays
 - Comparing Essay Lengths
 - Analysis of most common words by student and LLM
- 1. Pre-processing the Data:
 - · Removing Duplicate records
 - Creating a vocabulary index: Using the TextVectorization to index the vocabulary found in the dataset. Later, we use the same layer instance to vectorize the samples. Our layer will only consider the top 30,000 words, and will truncate or pad sequences to be actually 200 tokens long.
- 1. Embedding and Modelling:
 - Used pre-trained GLOVE Embeddings
 - Created the Embedding Matrix
 - Created tensorflow.keras.layers.Embedding layer and embedded the text sequences
- 1. Modelling

Model Summary:

- input_1 (InputLayer)
- embedding_1 (Embedding)

- bidirectional (Bidirectional LSTM Layer)
- dense (Dense)
- dense_1 (Dense)
- dropout (Dropout)
- dense 2 (Dense)

1. Evaluation:

- Validation Accuracy: 98.41%
- Kaggle Submission Accuracy: 80.9%

I. Analysing the Data

Importing the Dataset

```
In [6]: import pandas as pd
        import seaborn as sns
        from matplotlib import pyplot as plt
        import string
        import sys
        import gc
        from sklearn.model_selection import StratifiedKFold
        from sklearn.model selection import train test split
        from sklearn.metrics import roc auc score
        import numpy as np
        from tensorflow.keras.layers import TextVectorization
        import tensorflow as tf
        #from datasets import Dataset
        from tqdm.auto import tqdm
        from transformers import PreTrainedTokenizerFast
        from sklearn.linear model import SGDClassifier
In [2]: #Data given by Kaggle
```

train prompt file = "train prompts.csv"

df train prompt = pd.read csv(train prompt file)

```
train_essay_file = "train_essays.csv"
df_train_essay = pd.read_csv(train_essay_file)
test_essay_file = "test_essays.csv"
df_test_essay = pd.read_csv(test_essay_file)
sub_file = "sample_submission.csv"
df_sub = pd.read_csv(sub_file)
```

In [3]: df_train_essay.head()

Out[3]:		id	prompt_id	text	generated
	0	0059830c	0	Cars. Cars have been around since they became	0
	1	005db917	0	Transportation is a large necessity in most co	0
	2	008f63e3	0	"America's love affair with it's vehicles seem	0
	3	00940276	0	How often do you ride in a car? Do you drive a	0
	4	00c39458	0	Cars are a wonderful thing. They are perhaps o	0

In [4]: df_train_prompt.head()

Out[4]:		prompt_id prompt_name		instructions	source_text	
	0	0	Car-free cities	Write an explanatory essay to inform fellow ci	# In German Suburb, Life Goes On Without Cars	
	1	1	Does the electoral college work?	Write a letter to your state senator in which	# What Is the Electoral College? by the Office	

In [5]: df_train_essay.info()

```
<class 'pandas.core.frame.DataFrame'>
        RangeIndex: 1378 entries, 0 to 1377
        Data columns (total 4 columns):
             Column
                        Non-Null Count Dtype
             id
                        1378 non-null
                                        object
             prompt id 1378 non-null
                                        int64
                        1378 non-null
         2
             text
                                        object
             generated 1378 non-null
                                        int64
        dtypes: int64(2), object(2)
        memory usage: 43.2+ KB
In [6]: df_train_essay["generated"].value counts()
        generated
Out[6]:
             1375
        Name: count, dtype: int64
In [7]: ## Since there is class imbalanced problem, we will use external dataset
        train_essay_ex = "train_v2_drcat_02.csv"
        df_train_essay_ex = pd.read_csv(train_essay_ex)
        df train essay ex.rename(columns = {"label":"generated"}, inplace=True)
        ## Merging datasets
        df = pd.concat([df train essay ex[["text", "generated"]]],df train essay[["text", "generated"]]])
        df['generated'].value counts()
        generated
Out[7]:
             28746
             17500
        Name: count, dtype: int64
In [8]: df.head()
```

In [9]:

```
O Phones\n\nModern humans today are always on th...

1 This essay will explain if drivers should or s...

2 Driving while the use of cellular devices\n\nT...

3 Phones & Driving\n\nDrivers should not be able...

4 Cell Phone Operation While Driving\n\nThe abil...

0
```

```
        Out[9]:
        id
        prompt_id
        text

        0
        00000aaaa
        2
        Aaa bbb ccc.

        1
        1111bbbb
        3
        Bbb ccc ddd.

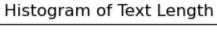
        2
        2222cccc
        4
        CCC ddd eee.
```

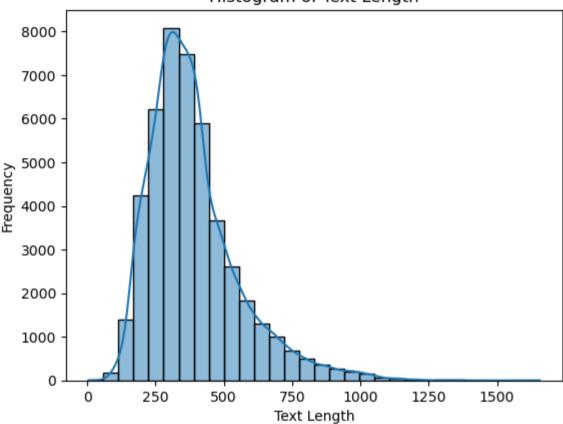
df_test_essay

Since the test data from kaggle is a dummy data, we will validate our results on using train test split in our dataframe

Analyzing the Texts

```
In [10]: #Length of essays
    df["text_len"] = df["text"].apply(lambda x : len(x.split()))
    sns.histplot(df['text_len'], bins=30, kde=True)
    plt.title('Histogram of Text Length')
    plt.xlabel('Text Length')
    plt.ylabel('Frequency')
    plt.show()
```





```
In [11]: df["text_len"].describe()
         count
                  46246.000000
Out[11]:
                    388.761428
         mean
                    167.367549
         std
                      4.000000
         min
         25%
                    276.000000
         50%
                    357.000000
         75%
                    461.000000
         max
                   1656.000000
         Name: text_len, dtype: float64
In [12]: #Count for unique words
         unique_words = set()
         for text in df['text']:
```

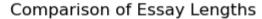
```
words = text.lower().split() # Convert to lowercase and split into words
unique_words.update(words)

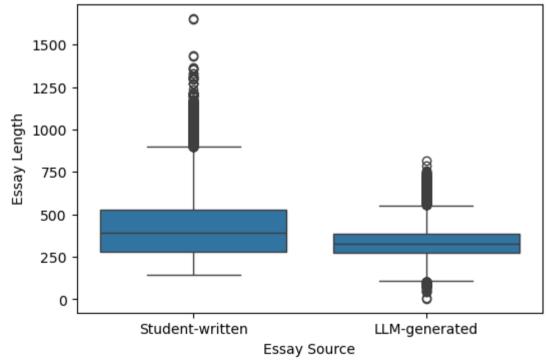
# Remove punctuation from each word
unique_words = {word.strip(string.punctuation) for word in unique_words}

# Now, unique_words set contains all unique words
total_unique_words = len(unique_words)
print("Total unique words:", total_unique_words)
```

Total unique words: 99072

```
In [13]: plt.figure(figsize=(6,4))
    sns.boxplot(x='generated', y='text_len', data=df)
    plt.title('Comparison of Essay Lengths')
    plt.xlabel('Essay Source')
    plt.ylabel('Essay Length')
    plt.xticks([0, 1], ['Student-written', 'LLM-generated'])
    plt.show()
```

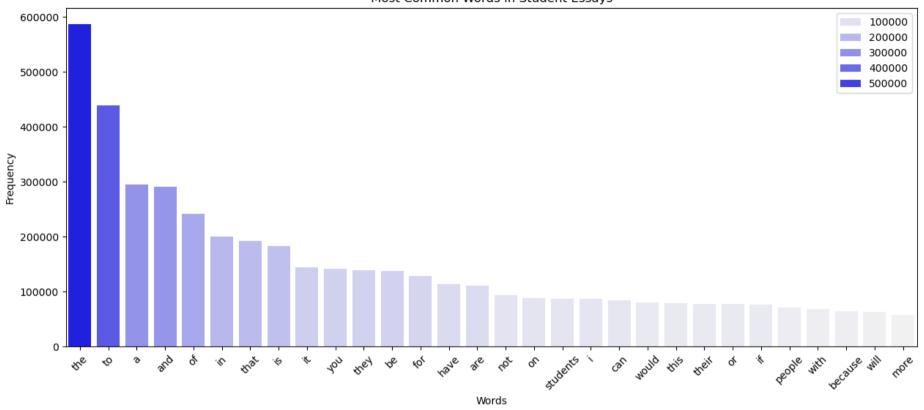




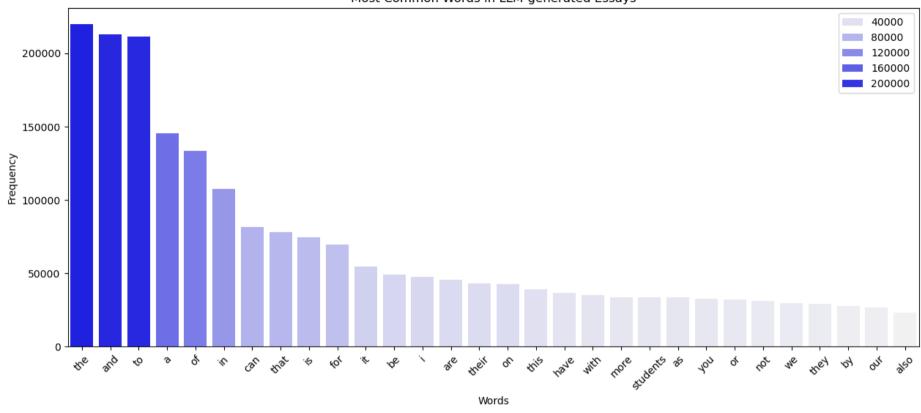
Analysis of most common words by student and LLM

```
In [14]: from collections import Counter
         import itertools
         def plot_most_common_words(text_series, num_words=30, title="Most Common Words"):
             all_text = ' '.join(text_series).lower()
             words = all text.split()
             word freg = Counter(words)
             common words = word freq.most common(num words)
             # Plot the most common words
             plt.figure(figsize=(15, 6))
             sns.barplot(x=[word for word, freq in common_words], y=[freq for word, freq in common_words], hue=[freq for word,
             plt.title(title)
             plt.xticks(rotation=45)
             plt.xlabel('Words')
             plt.ylabel('Frequency')
             plt.show()
         plot_most_common_words(df[df['generated'] == 0]['text'], title="Most Common Words in Student Essays")
         plot most common words(df[df['generated'] == 1]['text'], title="Most Common Words in LLM-generated Essays")
```

Most Common Words in Student Essays



Most Common Words in LLM-generated Essays



In [15]: **df**

Out[15]:

	text	generated	text_len
0	Phones\n\nModern humans today are always on th	0	379
1	This essay will explain if drivers should or s	0	366
2	Driving while the use of cellular devices\n\nT	0	178
3	Phones & Driving\n\nDrivers should not be able	0	212
4	Cell Phone Operation While Driving\n\nThe abil	0	332
•••			•••
1373	There has been a fuss about the Elector Colleg	0	430
1374	Limiting car usage has many advantages. Such a	0	397
1375	There's a new trend that has been developing f	0	749
1376	As we all know cars are a big part of our soci	0	525
1377	Cars have been around since the 1800's and hav	0	447

46246 rows × 3 columns

II. Data Pre-processing

1) Removing Duplicate records

```
In [16]: df.drop_duplicates(subset=['text'],inplace=True)
    df.reset_index(drop=True, inplace=True)
    df.drop('text_len',inplace=True ,axis=1 )
```

2) Train test split

```
In [17]: ## train_test_Split
X_train, X_test, y_train, y_test = train_test_split(df['text'],df['generated'], test_size=0.33, random_state=42)
print("X_train shape: ",X_train.shape)
print("y_train shape: ",y_train.shape)
```

```
print("X_test shape: ",X_test.shape)
print("y_test shape: ",y_test.shape)

X_train shape: (30061,)
y_train shape: (30061,)
X_test shape: (14807,)
y_test shape: (14807,)
```

3) Creating a vocabulary index

```
In [22]: #Using the TextVectorization to index the vocabulary found in the dataset.
#Later, we'll use the same layer instance to vectorize the samples.
#Our layer will only consider the top 30,000 words, and will truncate or pad sequences
#to be actually 200 tokens long.

vectorizer = TextVectorization(max_tokens=30000, output_sequence_length=200)
text_ds = tf.data.Dataset.from_tensor_slices(X_train).batch(128)
vectorizer.adapt(text_ds)

2024-01-19 16:04:53.275354: I tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114] Plugin optimizer for device_type GPU is enabled.

In [23]: vectorizer.get_vocabulary()[:10]

Out[23]: ['', '[UNK]', 'the', 'to', 'and', 'a', 'of', 'in', 'that', 'is']

In [24]: voc = vectorizer.get_vocabulary()
word_index = dict(zip(voc, range(len(voc))))
```

III. Embedding & Modelling

Load pre-trained Glove word embeddings

```
import os
path_to_glove_file = os.path.join(
    os.path.expanduser("~"), "/Users/dibyanshu/Documents/Jupyter Codes/NLP/Movie Classifier/glove.6B/glove.6B.200d.tx1
)
embeddings_index = {}
with open(path_to_glove_file) as f:
```

```
for line in f:
    word, coefs = line.split(maxsplit=1)
    coefs = np.fromstring(coefs, "f", sep=" ")
    embeddings_index[word] = coefs

print("Found %s word vectors." % len(embeddings_index))
```

Found 400000 word vectors.

Creating Embedding Matrix

```
In [26]: num\_tokens = len(voc) + 2
         embedding dim = 200
         hits = 0
         misses = 0
         # Prepare embedding matrix
         embedding_matrix = np.zeros((num_tokens, embedding_dim))
         for word, i in word index.items():
             embedding vector = embeddings index.get(word)
             if embedding vector is not None:
                 # Words not found in embedding index will be all-zeros.
                 # This includes the representation for "padding" and "00V"
                 embedding matrix[i] = embedding vector
                 hits += 1
             else:
                 misses += 1
         print("Converted %d words (%d misses)" % (hits, misses))
```

Converted 20426 words (9574 misses)

Creating Embedding Layer

```
In [28]: from tensorflow.keras.layers import Embedding
import keras
embedding_layer = Embedding(
    num_tokens,
    embedding_dim,
    embeddings_initializer=keras.initializers.Constant(embedding_matrix),
    trainable=True,
)
```

IV. Modelling

```
In [29]: from tensorflow.keras import layers
    from tensorflow.keras.layers import Dense
    from tensorflow.keras.layers import Dense
    from tensorflow.keras.layers import Dropout
    from tensorflow.keras.layers import Bidirectional

# Creating model
    int_sequences_input = keras.Input(shape=(None,), dtype="int64")
    embedded_sequences = embedding_layer(int_sequences_input)
    x = layers.Bidirectional(LSTM(100))(embedded_sequences)
    x = layers.Dense(64, activation="relu")(x)
    x = layers.Dense(64, activation="relu")(x)
    x = layers.Dense(64, activation="relu")(x)
    nodel = keras.Model(int_sequences_input, preds)
    print(model.summary())
```

Model: "model"

Layer (type)	Output Shape	Param #				
input_1 (InputLayer)	[(None, None)]	0				
<pre>embedding_1 (Embedding)</pre>	(None, None, 200)	6000400				
<pre>bidirectional (Bidirection al)</pre>	(None, 200)	240800				
dense (Dense)	(None, 64)	12864				
dense_1 (Dense)	(None, 64)	4160				
dropout (Dropout)	(None, 64)	0				
dense_2 (Dense)	(None, 1)	65				
Total params: 6258289 (23.87 MB) Trainable params: 6258289 (23.87 MB) Non-trainable params: 0 (0.00 Byte)						

None

```
In [30]: x_train = vectorizer(np.array([[s] for s in X_train])).numpy()
    x_val = vectorizer(np.array([[s] for s in X_test])).numpy()

y_train = np.array(y_train)
    y_val = np.array(y_test)
```

Fitting the Model

```
In [31]: model.compile(
    loss="binary_crossentropy", optimizer="rmsprop", metrics=["acc"]
)
model.fit(x_train, y_train, batch_size=128, epochs=5, validation_data=(x_val, y_val))
Epoch 1/5
```

```
2024-01-19 16:12:24.682040: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:12:24.864243: I tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:12:24.875307: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device_type GPU is enabled.
      2024-01-19 16:12:25.141623: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:12:25.156087: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:12:48.864895: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:12:48.930037: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:12:48.938149: I tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      0.9689
      Epoch 2/5
      0.9848
      Epoch 3/5
      0.9868
      Epoch 4/5
      0.9901
      Epoch 5/5
      <keras.src.callbacks.History at 0x29b3920a0>
Out[31]:
In [32]: y_pred= model.predict(x_val)
      2024-01-19 16:15:56.444228: I tensorflow/core/grappler/optimizers/custom_graph_optimizer_registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:15:56.499720: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
      2024-01-19 16:15:56.516145: I tensorflow/core/grappler/optimizers/custom graph optimizer registry.cc:114] Plugin opti
      mizer for device type GPU is enabled.
```

V. Evaluation: Preparing Output For Kaggle

```
df_test_essay
In [35]:
Out[35]:
                 id prompt_id
                                    text
         0 0000aaaa
                           2 Aaa bbb ccc.
         1 1111bbbb
                              Bbb ccc ddd.
                           4 CCC ddd eee.
         2 2222cccc
In [36]: x_test = vectorizer(np.array([[s] for s in df_test_essay['text']])).numpy()
In [37]: y_pred= model.predict(x_test)
         1/1 [======= ] - 0s 86ms/step
In [38]: y_pred
         array([[0.97673917],
Out[38]:
                [0.5001904],
                [0.5001904 ]], dtype=float32)
In [39]: df_sub.drop('generated',axis=1,inplace= True)
In [40]: df_sub['generated']=y_pred
```

```
In [41]: df_sub.to_csv("Submission.csv")
In [42]: y_pred = [9,3,4]
         y_pred= np.array(y_pred)
In [43]: type(y_pred)
         numpy.ndarray
Out[43]:
 In [5]: y_pred=pd.Series(y_pred.reshape(-1))
 In [4]: df_sub
 Out[4]:
                  id generated
         0 0000aaaa
                           0.1
         1 1111bbbb
                           0.9
         2 2222ccc
                          0.4
 In [ ]:
```