

## Designing a NLP model on Sarcasm detection.

```
In [1]: import numpy as np
import pandas as pd
import tensorflow as tf
import seaborn as sns
import re
import re,string,unicodedata
from nltk.corpus import stopwords

from keras.preprocessing.text import Tokenizer
from keras.preprocessing.sequence import pad_sequences
from keras.models import Sequential

import warnings
warnings.filterwarnings("ignore")
```

```
In [2]: df = pd.read_json("Sarcasm_Headlines_Dataset_v2.json", lines=True)
df.head()
```

```
Out[2]:
```

	is_sarcastic	headline	article_link
0	1	thirtysomething scientists unveil doomsday clo...	<a href="https://www.theonion.com/thirtysomething-scienc...">https://www.theonion.com/thirtysomething-scienc...</a>
1	0	dem rep. totally nails why congress is falling...	<a href="https://www.huffingtonpost.com/entry/donna-edw...">https://www.huffingtonpost.com/entry/donna-edw...</a>
2	0	eat your veggies: 9 deliciously different recipes	<a href="https://www.huffingtonpost.com/entry/eat-your-...">https://www.huffingtonpost.com/entry/eat-your-...</a>
3	1	inclement weather prevents liar from getting t...	<a href="https://local.theonion.com/inclement-weather-p...">https://local.theonion.com/inclement-weather-p...</a>
4	1	mother comes pretty close to using word 'strea...	<a href="https://www.theonion.com/mother-comes-pretty-c...">https://www.theonion.com/mother-comes-pretty-c...</a>

```
In [3]: df.head()
```

```
Out[3]:
```

	is_sarcastic	headline	article_link
0	1	thirtysomething scientists unveil doomsday clo...	<a href="https://www.theonion.com/thirtysomething-scienc...">https://www.theonion.com/thirtysomething-scienc...</a>
1	0	dem rep. totally nails why congress is falling...	<a href="https://www.huffingtonpost.com/entry/donna-edw...">https://www.huffingtonpost.com/entry/donna-edw...</a>
2	0	eat your veggies: 9 deliciously different recipes	<a href="https://www.huffingtonpost.com/entry/eat-your-...">https://www.huffingtonpost.com/entry/eat-your-...</a>
3	1	inclement weather prevents liar from getting t...	<a href="https://local.theonion.com/inclement-weather-p...">https://local.theonion.com/inclement-weather-p...</a>
4	1	mother comes pretty close to using word 'strea...	<a href="https://www.theonion.com/mother-comes-pretty-c...">https://www.theonion.com/mother-comes-pretty-c...</a>

```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 28619 entries, 0 to 28618
Data columns (total 3 columns):
#   Column          Non-Null Count  Dtype
---  -
0   is_sarcastic    28619 non-null  int64
1   headline        28619 non-null  object
2   article_link    28619 non-null  object
dtypes: int64(1), object(2)
memory usage: 670.9+ KB
```

```
In [5]: df.shape
```

```
Out[5]: (28619, 3)
```

### checking for null values in train data

```
In [6]: df.isnull().sum()
```

```
Out[6]: is_sarcastic    0
headline        0
article_link     0
dtype: int64
```

```
In [7]: df.describe(include='object')
```

Out[7]:

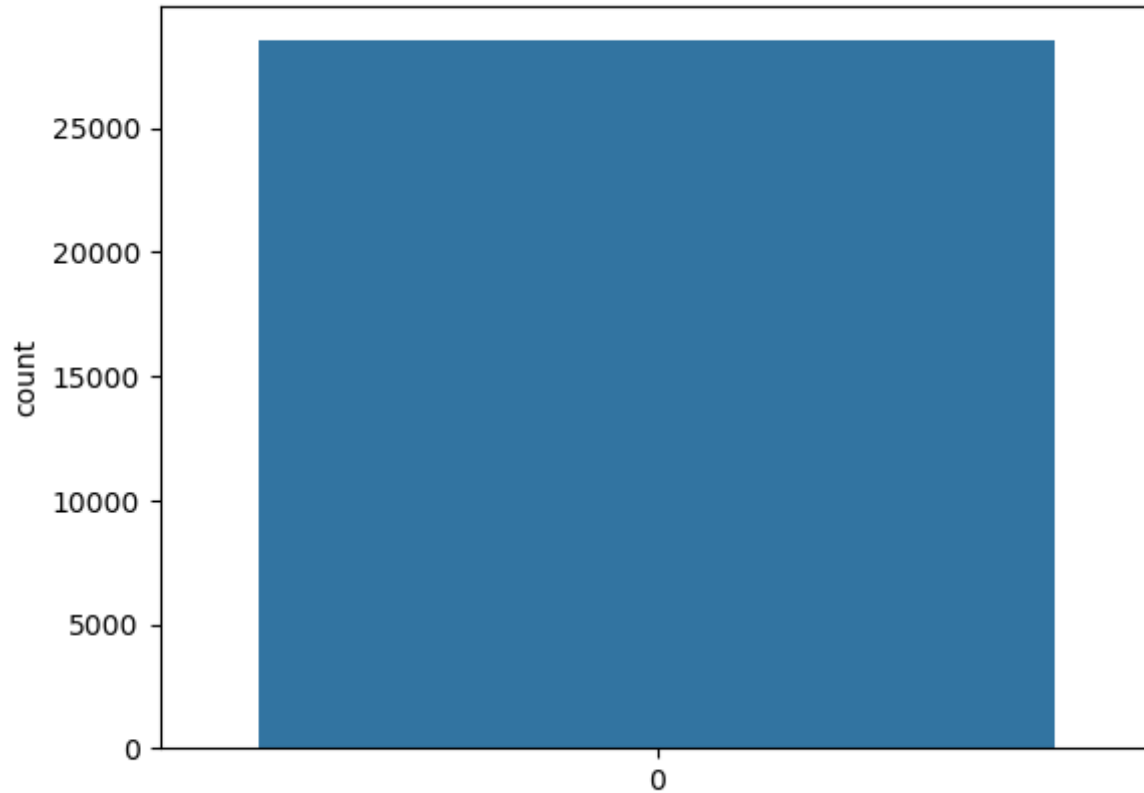
	headline	article_link
<b>count</b>	28619	28619
<b>unique</b>	28503	28617
<b>top</b>	'no way to prevent this,' says only nation whe...	<a href="https://politics.theonion.com/nation-not-sure-...">https://politics.theonion.com/nation-not-sure-...</a>
<b>freq</b>	12	2

checking for duplicate values

In [8]: `df['headline'].duplicated().sum()`

Out[8]: 116

In [9]: `df = df.drop(df[df['headline'].duplicated()].index, axis=0)`In [10]: `sns.countplot(df['is_sarcastic']);`



```
In [11]: import nltk
nltk.download('stopwords')
stop = set(stopwords.words('english'))
punctuation = list(string.punctuation)
stop.update(punctuation)
```

```
[nltk_data] Downloading package stopwords to
[nltk_data] C:\Users\santh\AppData\Roaming\nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

Removing the stopwords from text

```
In [12]: def split_into_words(text):
# split into words by white space
words = text.split()
return words
```

```
def to_lower_case(words):  
    # convert to lower case  
    words = [word.lower() for word in words]  
    return words  
  
def remove_punctuation(words):  
    # prepare regex for char filtering  
    re_punc = re.compile('[%s]' % re.escape(string.punctuation))  
    # remove punctuation from each word  
    stripped = [re_punc.sub('', w) for w in words]  
    return stripped  
  
def keep_alphabetic(words):  
    # remove remaining tokens that are not alphabetic  
    words = [word for word in words if word.isalpha()]  
    return words  
  
def remove_stopwords(words):  
    # filter out stop words  
    stop_words = set(stopwords.words('english'))  
    words = [w for w in words if not w in stop_words]  
    return words  
  
def to_sentence(words):  
    # join words to a sentence  
    return ' '.join(words)
```

### Removing the noisy text

```
In [13]: def denoise_text(text):  
        words = split_into_words(text)  
        words = to_lower_case(words)  
        words = remove_punctuation(words)  
        words = keep_alphabetic(words)  
        words = remove_stopwords(words)  
        return to_sentence(words)
```

```
In [14]: df['headline'] = df['headline'].apply(denoise_text)
```

Apply function on review column

```
In [15]: labels = (df['is_sarcastic'])
         data = (df['headline'])
```

```
In [16]: train_ratio = 0.80

         train_size = int(len(labels)*train_ratio)

         train_data = data[:train_size]
         train_labels= labels[:train_size]

         test_data = data[train_size:]
         test_labels = labels[train_size:]
```

```
In [17]: tokenizer = Tokenizer(oov_token='<OOV>')
         tokenizer.fit_on_texts(train_data)

         vocab_size = len(tokenizer.word_index)
         print(vocab_size)

         train_sequences = tokenizer.texts_to_sequences(train_data)
         test_sequences = tokenizer.texts_to_sequences(test_data)
```

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```
In [18]: maxlen=max([len(i) for i in train_sequences])
         train_padded = pad_sequences(train_sequences, maxlen=maxlen, padding='post')
         test_padded = pad_sequences(test_sequences, maxlen=maxlen, padding='post')
```

**Print a sample headline**

```
In [19]: index = 10
         print(f'sample headline: {train_sequences[index]}')
         print(f'padded sequence: {train_padded[index]} \n')

         print(f'Original Sentence: \n {tokenizer.sequences_to_texts(train_sequences[index:index+1])} \n')

         # Print dimensions of padded sequences
         print(f'shape of padded sequences: {train_padded.shape}')
```

sample headline: [1972, 2572, 315, 3022, 943, 7]

```

padded sequence: [1972 2572 315 3022 943 7 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0 0 0 0 0 0 0
0 0 0 0 0 0 0 0]

```

Original Sentence:

['lesbian considered father indiana amazing one']

shape of padded sequences: (22802, 106)

### Model Building:

```

In [20]: model = tf.keras.Sequential([
    tf.keras.layers.Embedding(vocab_size+1,100,input_length=maxlen),
    tf.keras.layers.Bidirectional( tf.keras.layers.LSTM(128)),
    tf.keras.layers.Flatten(),
    tf.keras.layers.Dropout(0.50),
    tf.keras.layers.Dense(64,activation='relu'),
    tf.keras.layers.Dense(1,activation='sigmoid')
])
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
model.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 106, 100)	2566300
bidirectional (Bidirectional)	(None, 256)	234496
flatten (Flatten)	(None, 256)	0
dropout (Dropout)	(None, 256)	0
dense (Dense)	(None, 64)	16448
dense_1 (Dense)	(None, 1)	65

=====  
 Total params: 2817309 (10.75 MB)  
 Trainable params: 2817309 (10.75 MB)  
 Non-trainable params: 0 (0.00 Byte)

In [21]: `history=model.fit(train_padded, np.array(train_labels),validation_data = (test_padded,np.array(test_labels)) , epochs = 5 , verbo`

```
Epoch 1/5
713/713 - 119s - loss: 0.4809 - accuracy: 0.7563 - val_loss: 0.4040 - val_accuracy: 0.8155 - 119s/epoch - 167ms/step
Epoch 2/5
713/713 - 108s - loss: 0.2264 - accuracy: 0.9109 - val_loss: 0.4603 - val_accuracy: 0.8100 - 108s/epoch - 152ms/step
Epoch 3/5
713/713 - 106s - loss: 0.0890 - accuracy: 0.9683 - val_loss: 0.6091 - val_accuracy: 0.8025 - 106s/epoch - 149ms/step
Epoch 4/5
713/713 - 105s - loss: 0.0449 - accuracy: 0.9849 - val_loss: 0.7368 - val_accuracy: 0.7951 - 105s/epoch - 147ms/step
Epoch 5/5
713/713 - 100s - loss: 0.0263 - accuracy: 0.9915 - val_loss: 1.0178 - val_accuracy: 0.7850 - 100s/epoch - 140ms/step
```

In [22]: `import matplotlib.pyplot as plt`

```
# Plot utility
def plot_graphs(model, string):
    plt.plot(model.history[string])
    plt.plot(model.history['val_'+string])
    plt.xlabel("Epochs")
    plt.ylabel(string)
```



```
plt.legend([string, 'val_'+string])  
plt.show()
```

```
# Plot the accuracy and loss  
plot_graphs(history, "accuracy")  
plot_graphs(history, "loss")
```

