

TCS RIO 125 Internship project[+ Code](#)[+ Text](#)**Topic : Automate Sentiment Analysis of textual comments and feedback****Reference Links from TCS**<https://www.intel.com/content/www/us/en/developer/topic-technology/artificial-intelligence/training/courses.html><https://python-course.eu/machine-learning/><https://scikit-learn.org/stable/tutorial/index.html><https://pandas.pydata.org/pandas-docs/version/0.15/tutorials.html><https://www.tensorflow.org/tutorials><https://www.tensorflow.org/tutorials/keras/classification><https://www.tensorflow.org/tutorials/images/cnn>https://www.tensorflow.org/text/tutorials/text_classification_rnn<https://www.intel.com/content/www/us/en/developer/learn/course-deep-learning.html><https://www.intel.com/content/www/us/en/developer/learn/course-machine-learning.html>**Importing Libraries**

```
import numpy as np
import pandas as pd
print('Libraries Imported')
import tensorflow_datasets as tfds
import tensorflow as tf
```

Libraries Imported

Loading the Dataset

```
dataset, info = tfds.load('imdb_reviews/subwords8k', with_info=True, as_supervised=True)
train_dataset, test_dataset = dataset['train'], dataset['test']
```

WARNING:absl:TFDS datasets with text encoding are deprecated and will be removed in a future version. Instead, you should use tfds.load('imdb_reviews/subwords8k', with_info=True, as_supervised=True) to load the dataset. Downloading and preparing dataset 80.23 MiB (download: 80.23 MiB, generated: Unknown size, total: 80.23 MiB) to /root/tensorflow_datasets/imdb_reviews/subwords8k/1.0.0. Subsequent calls will

DI Completed...: 100% 1/1 [00:10<00:00, 10.07s/ url]

DI Size...: 100% 80/80 [00:10<00:00, 16.44 MiB/s]

WARNING:absl:Dataset is using deprecated text encoder API which will be removed soon. Please use the plain_text version of the Dataset imdb_reviews downloaded and prepared to /root/tensorflow_datasets/imdb_reviews/subwords8k/1.0.0. Subsequent calls will

```

train_examples_batch, train_labels_batch = next(iter(train_dataset))
print(train_examples_batch)
print(train_labels_batch)

tf.Tensor(
[ 62  18  41 604 927  65   3 644 7968  21  35 5096  36  11
 43 2948 5240 102  50 681 7862 1244   3 3266  29 122 640   2
 26  14  279 438  35  79 349 384  11 1991   3 492  79 122
188 117  33 4047 4531  14  65 7968   8 1819 3947   3  62  27
  9  41  577 5044 2629 2552 7193 7961 3642   3  19 107 3903 225
 85 198  72   1 1512 738 2347 102 6245   8  85 308  79 6936
7961 23 4981 8044   3 6429 7961 1141 1335 1848 4848  55 3601 4217
8050   2   5  59 3831 1484 8040 7974 174 5773  22 5240 102  18
247  26   4 3903 1612 3902 291  11   4  27  13  18 4092 4008
7961   6 119 213 2774   3  12 258 2306  13  91  29 171  52
229   2 1245 5790 995 7968   8  52 2948 5240 8039 7968   8  74
1249   3  12 117 2438 1369 192  39 7975], shape=(163,), dtype=int64)
tf.Tensor(0, shape=(), dtype=int64)

```

Text Encoding

```

encoder = info.features['text'].encoder
print('Vocabulary size: {}'.format(encoder.vocab_size))

```

Vocabulary size: 8185

```
sample_string = 'Hello TensorFlow.'
```

```

encoded_string = encoder.encode(sample_string)
print('Encoded string is {}'.format(encoded_string))

```

```

original_string = encoder.decode(encoded_string)
print('The original string: {}'.format(original_string))

```

Encoded string is [4025, 222, 6307, 2327, 4043, 2120, 7975]
The original string: "Hello TensorFlow."

```

assert original_string == sample_string
for index in encoded_string:
    print('{} ----> {}'.format(index, encoder.decode([index])))

```

4025 ----> Hell
222 ----> o
6307 ----> Ten
2327 ----> sor
4043 ----> Fl
2120 ----> ow
7975 ----> .

```

BUFFER_SIZE = 10000
BATCH_SIZE = 64

```

```

train_dataset = train_dataset.shuffle(BUFFER_SIZE)
train_dataset = train_dataset.padded_batch(BATCH_SIZE)

```

```
test_dataset = test_dataset.padded_batch(BATCH_SIZE)
```

Building the Model

```
model = tf.keras.Sequential([
    tf.keras.layers.Embedding(encoder.vocab_size, 64),
    tf.keras.layers.Bidirectional(tf.keras.layers.LSTM(64, return_sequences=True)),
    tf.keras.layers.Bidirectional(tf.keras.layers.LSTM(32)),
    tf.keras.layers.Dense(64, activation='relu'),
    tf.keras.layers.Dropout(0.5),
    tf.keras.layers.Dense(1)
])

model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
=====		
embedding (Embedding)	(None, None, 64)	523840
bidirectional (Bidirectional)	(None, None, 128)	66048
bidirectional_1 (Bidirectional)	(None, 64)	41216
dense (Dense)	(None, 64)	4160
dropout (Dropout)	(None, 64)	0
dense_1 (Dense)	(None, 1)	65
=====		
Total params: 635329 (2.42 MB)		
Trainable params: 635329 (2.42 MB)		
Non-trainable params: 0 (0.00 Byte)		

```
model.compile(loss = tf.keras.losses.BinaryCrossentropy(from_logits=True),
              optimizer = tf.keras.optimizers.Adam(1e-4),
              metrics = ['accuracy'])
```

Training the model

```
history = model.fit(train_dataset, epochs = 10, validation_data = test_dataset, validation_steps = 30)
```

```
Epoch 1/10
391/391 [=====] - 148s 343ms/step - loss: 0.6589 - accuracy: 0.5544 - val_loss: 0.5067 - val_accuracy: 0.7672
Epoch 2/10
391/391 [=====] - 100s 254ms/step - loss: 0.3694 - accuracy: 0.8493 - val_loss: 0.3569 - val_accuracy: 0.8516
Epoch 3/10
391/391 [=====] - 90s 229ms/step - loss: 0.2664 - accuracy: 0.9035 - val_loss: 0.3465 - val_accuracy: 0.8677
Epoch 4/10
391/391 [=====] - 80s 204ms/step - loss: 0.2159 - accuracy: 0.9282 - val_loss: 0.3524 - val_accuracy: 0.8630
```

```

Epoch 5/10
391/391 [=====] - 75s 191ms/step - loss: 0.1827 - accuracy: 0.9423 - val_loss: 0.4467 - val_accuracy: 0.8578
Epoch 6/10
391/391 [=====] - 70s 178ms/step - loss: 0.1574 - accuracy: 0.9535 - val_loss: 0.4124 - val_accuracy: 0.8542
Epoch 7/10
391/391 [=====] - 67s 170ms/step - loss: 0.1423 - accuracy: 0.9579 - val_loss: 0.4605 - val_accuracy: 0.8536
Epoch 8/10
391/391 [=====] - 66s 169ms/step - loss: 0.1226 - accuracy: 0.9661 - val_loss: 0.4713 - val_accuracy: 0.8531
Epoch 9/10
391/391 [=====] - 65s 167ms/step - loss: 0.1063 - accuracy: 0.9735 - val_loss: 0.4894 - val_accuracy: 0.8573
Epoch 10/10
391/391 [=====] - 63s 161ms/step - loss: 0.0934 - accuracy: 0.9776 - val_loss: 0.5212 - val_accuracy: 0.8542

```

Test for Accuracy

```
test_loss, test_acc = model.evaluate(test_dataset)
```

```
print('Test Loss: {}'.format(test_loss))
print('Test Accuracy: {}'.format(test_acc))
```

```

391/391 [=====] - 25s 63ms/step - loss: 0.5205 - accuracy: 0.8490
Test Loss: 0.5205340385437012
Test Accuracy: 0.8489999771118164

```

Prediction Functions

```

def pad_to_size(vec, size):
    zeros = [0] * (size - len(vec))
    vec.extend(zeros)
    return vec

```

```

def sample_predict(sample_pred_text, pad):
    encoded_sample_pred_text = encoder.encode(sample_pred_text)

    if pad:
        encoded_sample_pred_text = pad_to_size(encoded_sample_pred_text, 64)
    encoded_sample_pred_text = tf.cast(encoded_sample_pred_text, tf.float32)
    predictions = model.predict(tf.expand_dims(encoded_sample_pred_text, 0))
    print("Prediction Score: ", predictions)
    output = ""
    if predictions[0][0] >= 0.5: output = "POSITIVE"
    elif predictions[0][0] <= -1: output = "NEGATIVE"
    else: output = "NEUTRAL"

    return output

```

Prediction with samples

```

sample_pred_text = ('The movie was not good. The animation and the graphics were terrible. I would not recommend this movie.')
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)

```

```
1/1 [=====] - 3s 3s/step
Prediction Score: [[-1.96656]]
NEGATIVE
```

```
sample_pred_text = ('The movie was cool. The animation and the graphics were out of this world. I would recommend this movie.')
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 1s 1s/step
Prediction Score: [[0.797149]]
POSITIVE
```

```
sample_pred_text = ('This movie is awesome. The acting was incredible. Highly recommend')
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 24ms/step
Prediction Score: [[2.0023649]]
POSITIVE
```

```
sample_pred_text = ('This movie was so so. The acting was mediocre. Kind of recommend')
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 21ms/step
Prediction Score: [[-0.20042937]]
NEUTRAL
```

```
# AVENGERS: ENDGAME 5 STAR COMMENT
```

```
sample_pred_text = ("I loved the movie a lot as I am great fan of marvel! Avengers: Endgame, which marks the end of the Infinity Saga, is spellbounding and surely an enthralling experience. The last film of the 'Avengers' franchise is remarkable and doesn't disappoint. Watching all our favourite superheroes in one film is just surpassing. Marvel has been working on this grand culmination ever since they released 'Iron Man'. I'm damn sure that all of their hard work and ambition has paid off. The directors, Anthony and Joe Russo, have made sure that it delivers an unforgettable experience. Christopher Markus and Stephen McFeely have come up with a screenplay full of epic and unpredictable moments. The film has a great balance of humour, emotions and action. The biggest strength of the film is the emotions. This is the most emotional superhero film I have ever seen. It's just perfect. The action sequences were jaw-dropping. The climatic battle left me amazed. It's just filled with memorable moments and cannot be described with words. The visuals are gorgeous and have a great impact on the film. The humour doesn't look exaggerated and manages to entertain throughout the film. The plot twists were very impressive and suspenseful. The film features many cameos of characters from the previous MCU films, which just gives a double dose of excitement. The background score gives me goosebumps, though I've listened to it several times. It was really clever to make changes to the characterization of the Hulk. I enjoyed that a lot. But the show-stealer is Robert Downey Jr, who plays the role of Tony Stark/Iron Man. The man who started it all proves yet again that there's no one else who can perfect his role. Do not miss his powerful moments in the final battle.")
```

```
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 38ms/step
Prediction Score: [[4.885022]]
POSITIVE
```

```
# AVENGERS: ENDGAME 3 STAR COMMENT
```

```
sample_pred_text = ("Overrated Sequel, But Still Good, But Violent! Beloved characters die, a character gets their head cut off, someone falls off a cliff, stabbings, punching, shooting, and more. The characters swear a bit. Even Captain America does! Role models include, Captain America, Iron Man, Hulk, AntMan,
```

```
Black Widow, Hawkeye. Thor not so much because he SPOILER ALERT: got fat and played Fortnite the entire 5 year period. He was funny, but you may not want your kid wanting to say 'I want to be like the God of Thunder and play fortnite all day'. Characters did kiss. Thor drank a lot of beer. Captain America proved himself of something but not saying what. With reviewing the movie, the first half hour was good, then the next half hour was slow. The second half hour was pretty good, but the last hour was epic. From just starting out with Iron Man, Cap, Thor, and Hulk, to this amazing fight scene at the end, Avengers Endgame really pulled it off. Overall, pretty good for families and a good finale for the Infinity Saga.""")
```

```
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 30ms/step
Prediction Score: [[-0.50655866]]
NEUTRAL
```

```
# AVENGERS: ENDGAME 3 STAR COMMENT
```

```
sample_pred_text = ("""I don't get why so many people like this movie so much, all they did was go back in time, so it didn't add much to the story. They also just added a whole bunch of scenes of previous marvel movies, and that is how they were able to make the movie three hours long. Now, getting to the inappropriate content for the parents. This is just your average superhero movie with sci-fi violence and some language. There are also strong roll models, but if you have a kid that is in elementary school or higher, you should let them watch this movie if they want to. But, I am not trying to parent your child, I am just giving my personal opinion, so you can choose to let them watch it, or you can prevent them from watching it.""")
```

```
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 41ms/step
Prediction Score: [[-1.22496]]
NEGATIVE
```

```
# AVENGERS ENDGAME COMMENT
```

```
sample_pred_text = ("""What a great way to end several major storylines that they invested in over the past 11 years. For people who have watched the saga, I feel like this is just the cherry on the top. My only complaint is something that you can't really get away from in superhero stories: the character 'powers' are totally inconsistent from scene to scene, and movie to movie. This is a trope that there's no getting away from, because if characters like Captain Marvel, Thor, Scarlet Witch, and Hulk were always as powerful as they show flashes of, then the story wouldn't even be a thing. Any one of them could destroy Thanos in the blink of an eye, and have done similar feats in other stories (and even in other scenes within a given story). That they sometimes 'reduce' their power to a lower level, without an explained mechanism, is pretty laughable, and makes some parts of the story a bit nonsensical. Yes, this constant Ex Machina is needed to maintain the drama and keep the plot going, but it's still something that takes me out of the story.""")
```

```
predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 31ms/step
Prediction Score: [[-0.14023513]]
NEUTRAL
```

```
# AVENGERS: ENDGAME 1 STAR COMMENT
```

```
sample_pred_text = ("""Disappointing storyline - too many sad crying scenes - too much shit swearing compared to other great Marvel movies! Even as an adult I don't appreciate swearing in movies. There are MANY people who don't use cuss words in their lives except maybe in adrenaline traffic moments. To hear Robert Downey jr's 'moment' with his young child using and laughing at the fact that she uses 'adult' language is teaching the new impressionable ages watching this, that it is okay when it isn't. They seemed to want to use their cuss word quota for the rating for this movie. Sad writing when that's how they get their best laughs from audience. Bring back your creative, quirky writers from the two The Defenders of the Galaxy. Now that's smart character development and writing without resorting to desperate shock value. We have loved every movie of that series and eagerly await the next one. The Storyline was soooo boring in this. All of us watching kept hoping it would improve and it didn't. I think the only real laugh we had was the encounter on the ship
```

```
between Quill (StarLord) and Thor. Subtle, but funny. We all could've cared less if anyone died. That's how checked out we were watching this LONG 3 HRS!!!
Why! It was torture and I felt robbed of my time in the end.""")

predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 35ms/step
Prediction Score:  [[-4.302364]]
NEGATIVE
```

```
sample_pred_text = (""""Superhero comics, and much of their adaptations, have long taken an outsized, soap opera-like approach to storytelling.
At their best, they can take these fantastical ideas and make them emotionally resonant, even if there's obviously no real-world phenomenon to connect them to.
In some respects, Endgame pulls this off beautifully, like how the character Nebula confronts her past self through time travel, giving physical form to
her personal growth. But as fun as the movie is, there's an undeniable hollowness at its core induced by its unwillingness to follow through on certain
ideas and symbols.""")

predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 30ms/step
Prediction Score:  [[2.2788894]]
POSITIVE
```

```
# AUGUST to JUNE: Bringing Life to School
```

```
sample_pred_text = (""""After we screened August To June, Boynton, Boca Democratic Party Movie Night presenters said in all their years of doing this had
they never seen an audience response so serene. People wanted to stay. No one was angry. Conversations that never occurred did so after this viewing. People
who had known each other for years discovered commonalities that previously they did not know existed. When this was shared I asked for any theories.
The response, in unison, August To June is warm. It touches people. Real life school situations gave the audience hope not only for what could be, but
already is. People were reminded of good times and the challenges that helped them grow greater. As a Teacher, Amy taught as I think wise to live. Reciprocal
reverence.""")

predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

```
1/1 [=====] - 0s 28ms/step
Prediction Score:  [[3.4267743]]
POSITIVE
```

```
# AUGUST to JUNE: Bringing Life to School (http://augusttojune.com/press-media/audience-comments/)
```

```
sample_pred_text = (""""The film's ever-present focus on the 'big picture' of education and life! was much appreciated here and provided lots of talking points.
In particular, we really liked seeing how you conferenced with parents, e.g., paraphrased: 'In a few more years, none of you will even remember who the early
readers and late readers were', the overall approach to literacy- holistically focused rather than merely skills-focused, the ample time you provided to social
growth/experiences/conflict resolution, and the fact that you did not choose to hide those moments where you, as a teacher, needed to separate or redirect
children when disruptive. There is a whole, whole, whole lot more that I wish to say! It is sad to know that this marked your retirement, but very
heart-warming to know that your school carries on.""")

predictions = sample_predict(sample_pred_text, pad = False)
print(predictions)
```

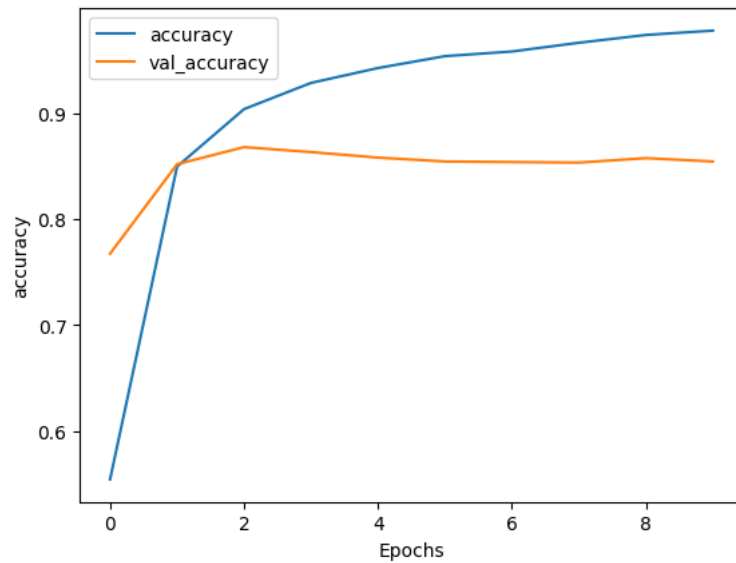
```
1/1 [=====] - 0s 27ms/step
Prediction Score:  [[2.7171698]]
POSITIVE
```

Plotting Accuracy

```
import matplotlib.pyplot as plt

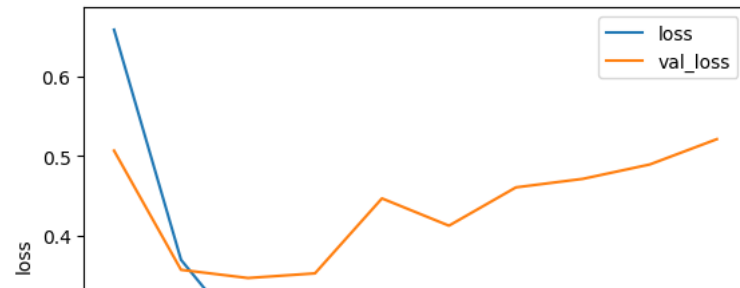
def plot_graphs(history, metric):
    plt.plot(history.history[metric])
    plt.plot(history.history['val_'+metric], '')
    plt.xlabel("Epochs")
    plt.ylabel(metric)
    plt.legend([metric, 'val_'+metric])
    plt.show()

plot_graphs(history, 'accuracy')
```



Plotting loss function

```
plot_graphs(history, 'loss')
```

Test accuracy achieved with the RNN LSTM Layers is 84.90 %

The developed Algorithm is able to detect different types of sentiments. It is also able to predict overall sentiment of a paragraph with a reasonable accuracy.

THE END OF THE PROJECT

Epochs