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|-------------------------|---|
| <b>Ex. No. 9</b>        | <b>Transformer model for a text classification task</b> |
| <b>Date of Exercise</b> | <b>10/11/2025</b>                                       |

**Aim:**

To build a neural machine translation model using the Keras Functional API and evaluate translation quality using BLEU scores.

**Description:**

Neural Machine Translation (NMT) is based on deep learning techniques that translate text from one language to another. BLEU scores measure the accuracy of the generated translations.

**Code:**

```
import tensorflow as tf
from tensorflow.keras.layers import Input, LSTM, Dense
from tensorflow.keras.models import Model
from nltk.translate.bleu_score import sentence_bleu
# Define model architecture
encoder_inputs = Input(shape=(None, 100))
encoder_lstm = LSTM(256, return_state=True)
encoder_outputs, state_h, state_c = encoder_lstm(encoder_inputs)
encoder_states = [state_h, state_c]
decoder_inputs = Input(shape=(None, 100))
decoder_lstm = LSTM(256, return_sequences=True, return_state=True)
```

```

decoder_outputs, _, _ = decoder_lstm(decoder_inputs, initial_state=encoder_states)
decoder_dense = Dense(100, activation='softmax')
decoder_outputs = decoder_dense(decoder_outputs)

# Define the model
nmt_model = Model([encoder_inputs, decoder_inputs], decoder_outputs)
nmt_model.compile(optimizer='adam', loss='categorical_crossentropy')
print(nmt_model.summary())

# BLEU score evaluation
def evaluate_bleu(reference, candidate):
    return sentence_bleu([reference.split()], candidate.split())
reference_text = "hello world"
candidate_text = "hello earth"
print("BLEU Score:", evaluate_bleu(reference_text, candidate_text))

```

### Sample Output:

```

Model: "functional"



| Layer (type)                              | Output Shape                                                                                                                                              | Param #              | Connected to                                                                            |
|-------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------|----------------------|-----------------------------------------------------------------------------------------|
| input_layer ( <code>InputLayer</code> )   | ( <code>None</code> , <code>None</code> , <code>100</code> )                                                                                              | <code>0</code>       | -                                                                                       |
| input_layer_1 ( <code>InputLayer</code> ) | ( <code>None</code> , <code>None</code> , <code>100</code> )                                                                                              | <code>0</code>       | -                                                                                       |
| lstm ( <code>LSTM</code> )                | [ <code>(None</code> , <code>256</code> ),<br><code>(None</code> , <code>256</code> ),<br><code>(None</code> , <code>256</code> )]                        | <code>365,568</code> | <code>input_layer[0][0]</code>                                                          |
| lstm_1 ( <code>LSTM</code> )              | [ <code>(None</code> , <code>None</code> ,<br><code>256</code> ), <code>(None</code> ,<br><code>256</code> ), <code>(None</code> ,<br><code>256</code> )] | <code>365,568</code> | <code>input_layer_1[0]...</code><br><code>lstm[0][1],</code><br><code>lstm[0][2]</code> |
| dense ( <code>Dense</code> )              | ( <code>None</code> , <code>None</code> , <code>100</code> )                                                                                              | <code>25,700</code>  | <code>lstm_1[0][0]</code>                                                               |



Total params: 756,836 (2.89 MB)

Trainable params: 756,836 (2.89 MB)

Non-trainable params: 0 (0.00 B)

None

```

```
candidate_text = "hello earth"
print("BLEU Score:", evaluate_bleu(reference_text, candidate_text))
✓ 0.0s
BLEU Score: 1.5319719891192393e-231
/home/franz/Documents/Lab/.venv/lib/python3.12/site-packages/nltk/translate/bleu_score.py:577: UserWarning:
The hypothesis contains 0 counts of 2-gram overlaps.
Therefore the BLEU score evaluates to 0, independently of
how many N-gram overlaps of lower order it contains.
Consider using lower n-gram order or use SmoothingFunction()
    warnings.warn(_msg)
/home/franz/Documents/Lab/.venv/lib/python3.12/site-packages/nltk/translate/bleu_score.py:577: UserWarning:
The hypothesis contains 0 counts of 3-gram overlaps.
Therefore the BLEU score evaluates to 0, independently of
how many N-gram overlaps of lower order it contains.
Consider using lower n-gram order or use SmoothingFunction()
    warnings.warn(_msg)
/home/franz/Documents/Lab/.venv/lib/python3.12/site-packages/nltk/translate/bleu_score.py:577: UserWarning:
The hypothesis contains 0 counts of 4-gram overlaps.
Therefore the BLEU score evaluates to 0, independently of
how many N-gram overlaps of lower order it contains.
Consider using lower n-gram order or use SmoothingFunction()
    warnings.warn(_msg)
```

### Result:

The above experiment of Transformer model for a text classification task is done successfully and the output is been obtained