

FinalProject_NLP520

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1 ChatBot Final Project: Team 7

Gurleen Virk, Jay Patel, and Will Kencel

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Professor Kahila Mokhtari Jadid

AAI: NLP520

```
[1]: # load libraries
import pandas as pd
import numpy as np
import re
import os

import tensorflow as tf
from tensorflow.keras.layers import Input, Embedding, LSTM, Bidirectional,
    ↳Dense, Concatenate, TimeDistributed
from tensorflow.keras.models import Model
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
from tqdm import tqdm
from sklearn.model_selection import train_test_split
```

1.1 Load Data & Text Pre-Processing

```
[2]: # load movie line from local directory (remember to change to YOUR file
    ↳location)
with open('/Users/gurl/Documents/AAI520_NLP/CornellData/movie_lines.txt', 'r',
    ↳encoding='utf-8', errors='replace') as file:
    lines = pd.read_table(file, sep='\t', header=None, on_bad_lines='skip')

# load conversation file from github (no need to change anything here for data
    ↳to load)
convlines = pd.read_table('https://raw.githubusercontent.com/wkencel/
    ↳Generative-Chatbot-Project/refs/heads/main/movie_conversations.txt',
    ↳sep='\t', header=None, encoding='utf-8', on_bad_lines='skip')
```

```
[4]: # view lines
lines[:10]
```

```
[4]: 0
0 L1045 +++$+++ u0 +++$+++ m0 +++$+++ BIANCA +++...
1 L1044 +++$+++ u2 +++$+++ m0 +++$+++ CAMERON ++...
2 L985 +++$+++ u0 +++$+++ m0 +++$+++ BIANCA +++$...
3 L984 +++$+++ u2 +++$+++ m0 +++$+++ CAMERON +++...
4 L925 +++$+++ u0 +++$+++ m0 +++$+++ BIANCA +++$...
5 L924 +++$+++ u2 +++$+++ m0 +++$+++ CAMERON +++...
6 L872 +++$+++ u0 +++$+++ m0 +++$+++ BIANCA +++$...
7 L871 +++$+++ u2 +++$+++ m0 +++$+++ CAMERON +++...
8 L870 +++$+++ u0 +++$+++ m0 +++$+++ BIANCA +++$...
9 L869 +++$+++ u0 +++$+++ m0 +++$+++ BIANCA +++$...
```

```
[5]: convolines[:10]
```

```
[5]: 0
0 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L194', 'L19...
1 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L198', 'L199']
2 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L200', 'L20...
3 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L204', 'L20...
4 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L207', 'L208']
5 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L271', 'L27...
6 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L276', 'L277']
7 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L280', 'L281']
8 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L363', 'L364']
9 u0 +++$+++ u2 +++$+++ m0 +++$+++ ['L365', 'L366']
```

```
[6]: # Create dictionary to map each line's id with its text
id2line = {}

# Iterate over each row in the dataframe and access the text data
for line in lines[0]: # Access the first column which contains the movie lines
    _line = line.split(' +++$+++ ')
    if len(_line) == 5:
        id2line[_line[0]] = _line[4]
```

```
[7]: # Print the first 10 entries
for i, (key, value) in enumerate(id2line.items()):
    if i < 10: # Change this number to see more or fewer entries
        print(f"{key}: {value}")
    else:
        break
```

```
L1045: They do not!
L1044: They do to!
L985: I hope so.
```

L984: She okay?
 L925: Let's go.
 L924: Wow
 L872: Okay -- you're gonna need to learn how to lie.
 L871: No
 L870: I'm kidding. You know how sometimes you just become this "persona"? And you don't know how to quit?
 L869: Like my fear of wearing pastels?

```
[8]: # Create a list of all of the conversations' lines' ids
convs = []
for index, row in convolines.iterrows():
    line = row[0] # Access the first column of the row
    _line = line.split(' +++$+++ ')[-1][1:-1].replace("'", "").replace(" ", "")
    convs.append(_line.split(','))
```

```
[9]: # Print the first 10 entries
convs[:10]
```

```
[9]: [['L194', 'L195', 'L196', 'L197'],
      ['L198', 'L199'],
      ['L200', 'L201', 'L202', 'L203'],
      ['L204', 'L205', 'L206'],
      ['L207', 'L208'],
      ['L271', 'L272', 'L273', 'L274', 'L275'],
      ['L276', 'L277'],
      ['L280', 'L281'],
      ['L363', 'L364'],
      ['L365', 'L366']]
```

```
[10]: # Sort the sentences: inputs (questions) and targets (answers)
questions = []
answers = []

for conv in convs:
    for i in range(len(conv) - 1):
        if conv[i] in id2line and conv[i + 1] in id2line:
            questions.append(id2line[conv[i]])
            answers.append(id2line[conv[i + 1]])
```

```
[11]: print("Number of questions:", len(questions))
      print("Number of answers:", len(answers))
```

Number of questions: 221416
 Number of answers: 221416

```
[12]: # Check if data is loaded correctly
limit = 0
```

```
for i in range(limit, limit+5):
    print(questions[i])
    print(answers[i])
    print()
```

Can we make this quick? Roxanne Korrine and Andrew Barrett are having an incredibly horrendous public break- up on the quad. Again.
Well, I thought we'd start with pronunciation, if that's okay with you.

Well, I thought we'd start with pronunciation, if that's okay with you.
Not the hacking and gagging and spitting part. Please.

Not the hacking and gagging and spitting part. Please.
Okay... then how 'bout we try out some French cuisine. Saturday? Night?

You're asking me out. That's so cute. What's your name again?
Forget it.

No, no, it's my fault -- we didn't have a proper introduction ---
Cameron.

```
[13]: # Create a DataFrame from questions and answers
data = {'Questions': questions, 'Answers': answers}
data = pd.DataFrame(data)
```

```
[14]: data.head()
```

```
[14]:
```

	Questions \	Answers
0	Can we make this quick? Roxanne Korrine and A...	Well, I thought we'd start with pronunciation,...
1	Well, I thought we'd start with pronunciation,...	Not the hacking and gagging and spitting part...
2	Not the hacking and gagging and spitting part...	Okay... then how 'bout we try out some French ...
3	You're asking me out. That's so cute. What's ...	Forget it.
4	No, no, it's my fault -- we didn't have a prop...	Cameron.

```
[15]: data.shape
```

```
[15]: (221416, 2)
```

```
[16]: # Remove duplicates
data.drop_duplicates(inplace = True)
```

```
[17]: data.shape
```

```
[17]: (220021, 2)
```

```
[18]: # Function for cleaning the text: lowercase, remove punctuations, and replace
      ↪ certain words

def clean_text(text):
    '''Clean text by removing unnecessary characters and altering the format of
    ↪ words.'''

    text = text.lower()

    text = re.sub(r"i'm", "i am", text)
    text = re.sub(r"he's", "he is", text)
    text = re.sub(r"she's", "she is", text)
    text = re.sub(r"it's", "it is", text)
    text = re.sub(r"that's", "that is", text)
    text = re.sub(r"what's", "what is", text)
    text = re.sub(r"where's", "where is", text)
    text = re.sub(r"there's", "there is", text)
    text = re.sub(r"how's", "how is", text)
    text = re.sub(r"\ll", " will", text)
    text = re.sub(r"\ve", " have", text)
    text = re.sub(r"\re", " are", text)
    text = re.sub(r"\d", " would", text)
    text = re.sub(r"\re", " are", text)
    text = re.sub(r"won't", "will not", text)
    text = re.sub(r"can't", "cannot", text)
    text = re.sub(r"n't", " not", text)
    text = re.sub(r"n'", "ng", text)
    text = re.sub(r"'bout", "about", text)
    text = re.sub(r"'til", "until", text)
    text = re.sub(r"[-()\"#/@;:<>{}~+=~|.!?,,]", "", text)

    return text
```

```
[19]: # Apply the function to the DataFrame
data['Questions'] = data['Questions'].apply(clean_text)
data['Answers'] = data['Answers'].apply(clean_text)

# Display the cleaned DataFrame
data.head()
```

```
[19]:
```

	Questions	
0	can we make this quick	roxanne korrine and an...
1	well i thought we would start with pronunciati...	

```

2 not the hacking and gagging and spitting part ...
3 you are asking me out that is so cute what is...
4 no no it is my fault we did not have a proper...

```

Answers

```

0 well i thought we would start with pronunciati...
1 not the hacking and gagging and spitting part ...
2 okay then how about we try out some french cui...
3 forget it
4 cameron

```

```
[20]: import nltk
      nltk.download('wordnet')
```

[nltk_data] Downloading package wordnet to /root/nltk_data...

[20]: True

```
[21]: # added data augmentation with synonyms
      # Data Augmentation using Synonym Replacement
      import random
      from nltk.corpus import wordnet

      def synonym_replacement(sentence, n):
          words = sentence.split()
          new_words = words.copy()
          random_word_list = list(set([word for word in words if wordnet.
          ↪synsets(word)]))
          random.shuffle(random_word_list)

          num_replaced = 0
          for random_word in random_word_list:
              synonyms = list(set([syn.name().split('.')[0] for syn in wordnet.
          ↪synsets(random_word)]))
              if len(synonyms) >= 1:
                  synonym = random.choice(synonyms)
                  new_words = [synonym if word == random_word else word for word in
          ↪new_words]
                  num_replaced += 1
              if num_replaced >= n:
                  break

          sentence = ' '.join(new_words)
          return sentence

      # Augment the dataset

```

```

augmented_questions = data['Questions'].apply(lambda x: synonym_replacement(x, 2))
augmented_answers = data['Answers'].apply(lambda x: synonym_replacement(x, 2))

# Combine original and augmented data
data['Questions'] = pd.concat([data['Questions'], augmented_questions], ignore_index=True)
data['Answers'] = pd.concat([data['Answers'], augmented_answers], ignore_index=True)

```

```

[25]: import string
import pandas as pd

# Define these globally
exclude = set(string.punctuation)
remove_digits = str.maketrans('', '', string.digits)

def preprocess_questions_sentences(sent):
    '''Function to preprocess English Sentence'''
    sent = sent.lower()
    sent = sent.replace('"', '')
    sent = ''.join(ch for ch in sent if ch not in exclude) # Use of exclude set
    sent = sent.translate(remove_digits) # Removes digits
    sent = sent.strip()
    sent = ' '.join(sent.split()) # Regularize spacing
    return sent

def preprocess_answer_sentence(sent):
    if isinstance(sent, str):
        sent = sent.lower()
        sent = sent.replace('"', '')
        sent = ''.join(ch for ch in sent if ch not in exclude)
        sent = sent.translate(remove_digits)
        sent = sent.strip()
        sent = ' '.join(sent.split()) # Condense spaces to single
        sent = "startseq " + sent + " endseq"
        return sent
    else:
        return sent

# Apply preprocess function on data
data['Questions'] = data['Questions'].apply(preprocess_questions_sentences)
data['Answers'] = data['Answers'].apply(preprocess_answer_sentence)

# Display the cleaned DataFrame
print(data.head())

```

Questions \

```

0 can we make this quick roxanne korrine and and...
1 well i thought we would start with pronunciati...
2 not the hacking and gagging and spitting part ...
3 you are asking me out that is so cute what is ...
4 no no it is my fault we did not have a proper ...

```

Answers

```

0 startseq well i thought we would start with pr...
1 startseq not the hacking and gagging and spitt...
2 startseq okay then how about we try out some f...
3 startseq forget it endseq
4 startseq cameron endseq

```

```

[27]: # Remove questions and answers shorter than 1 word and longer than 20 words
min_line_length = 1
max_line_length = 40 ### changed to larger word size

```

```

[28]: # Create a function to count the number of words in a text
def count_words(text):
    return len(text.split())

# Filter the DataFrame
filtered_data = data[
    (data['Questions'].apply(count_words).between(min_line_length,
    ↪max_line_length)) &
    (data['Answers'].apply(count_words).between(min_line_length,
    ↪max_line_length))
]

# Update the original DataFrame
data = filtered_data

data.head()

```

```

[28]: Questions \
0 can we make this quick roxanne korrine and and...
1 well i thought we would start with pronunciati...
2 not the hacking and gagging and spitting part ...
3 you are asking me out that is so cute what is ...
4 no no it is my fault we did not have a proper ...

```

Answers

```

0 startseq well i thought we would start with pr...
1 startseq not the hacking and gagging and spitt...
2 startseq okay then how about we try out some f...
3 startseq forget it endseq
4 startseq cameron endseq

```



```
[29]: # Sort Qs and As by length of questions to reduce amount of padding during
      ↪ training
      # Hope to speed up training and reduce the loss

      # Convert questions and answers to their respective lengths
      data['Question_Length'] = data['Questions'].apply(lambda x: len(x.split()))
      data['Answer_Length'] = data['Answers'].apply(lambda x: len(x.split()))

      # Sort Qs and As by length of questions
      sorted_questions = []
      sorted_answers = []

      for length in range(1, max_line_length + 1):
          for index, row in data.iterrows():
              if row['Question_Length'] == length:
                  sorted_questions.append(row['Questions'])
                  sorted_answers.append(row['Answers'])

      # Output the results
      print(len(sorted_questions))
      print(len(sorted_answers))
      print()
      for i in range(min(3, len(sorted_questions))): # Use min to avoid index errors
          print(f"Question {i + 1}: {sorted_questions[i]}")
          print(f"Answer {i + 1}: {sorted_answers[i]}")
          print()
```

206525

206525

Question 1: cameron

Answer 1: startseq the thing is cameron i am at the mercy of a particularly hideous breed of loser my sister i cannot date until she does endseq

Question 2: why

Answer 2: startseq unsolved mystery she used to be really popular when she started high school then it was just like she got sick of it or something endseq

Question 3: there

Answer 3: startseq where endseq

```
[30]: # Sort the DataFrame by question length
      data = data.sort_values(by='Question_Length')

      # Reset index if needed
      data.reset_index(drop=True, inplace=True)
```

```
# Output the sorted DataFrame
data[['Questions', 'Answers', 'Question_Length']].head()
```

```
[30]: Questions                                Answers \
0      yeah  startseq but see tonight wives and girlfriends...
1         no  startseq no cause tonight the girls are here i...
2  already      startseq greed it was written in blood endseq
3      wow                                startseq lets go endseq
4  cameron  startseq the thing is cameron i am at the merc...

      Question_Length
0                   1
1                   1
2                   1
3                   1
4                   1
```

```
[31]: data.shape
```

```
[31]: (206525, 4)
```

1.1.1 Vectorizing text

```
[32]: # Convert DataFrame columns to lists
q_sentences = data['Questions'].tolist()
a_sentences = data['Answers'].tolist()

### changed split ratios
# Define the split ratios
train_ratio = 0.75 # 80% for training
val_ratio = 0.15 # 10% for validation
test_ratio = 0.10 # 10% for testing

# Ensure the sum of ratios equals 1
assert train_ratio + val_ratio + test_ratio == 1.0, "Split ratios must sum to 1.
↳"

# Split into training and temporary sets (which will later be split into
↳validation and test)
train_q_sents, temp_q_sents, train_a_sents, temp_a_sents = train_test_split(
    q_sentences, a_sentences, test_size=(1 - train_ratio), random_state=42,
    ↳shuffle=True)

# Now split the temporary set into validation and test sets
val_size = val_ratio / (val_ratio + test_ratio) # Calculate validation size
↳relative to temp set
```

```
val_q_sents, test_q_sents, val_a_sents, test_a_sents = train_test_split(
    temp_q_sents, temp_a_sents, test_size=val_size, random_state=42,
    ↪shuffle=True)
```

```
[33]: # VOCABULARY
# Filter out non-string elements from training sets
train_q_sents = [str(sent) for sent in train_q_sents]
train_a_sents = [str(sent) for sent in train_a_sents]

# Tokenize question sentences
ques_tokenizer = Tokenizer(oov_token='<OOV>')
ques_tokenizer.fit_on_texts(train_q_sents)
ques_vocab_size = len(ques_tokenizer.word_index) + 1

# Tokenize answer sentences
ans_tokenizer = Tokenizer()
ans_tokenizer.fit_on_texts(train_a_sents)
ans_vocab_size = len(ans_tokenizer.word_index) + 1

print(f"Question Vocabulary Size: {ques_vocab_size}\nAnswer Vocabulary Size:
    ↪{ans_vocab_size}")
```

Question Vocabulary Size: 41701

Answer Vocabulary Size: 41571

The code prepares question and answer sentences for further processing by converting them to strings, creating tokenizers to build vocabularies, and calculating the vocabulary sizes for both questions and answers.

```
[34]: max_length = 40

# Convert text to sequences
ques_sequences = ques_tokenizer.texts_to_sequences(train_q_sents)
ans_sequences = ans_tokenizer.texts_to_sequences(train_a_sents)

# Pad sequences
source_seqs = pad_sequences(ques_sequences, maxlen=max_length, padding='post')
target_seqs = pad_sequences(ans_sequences, maxlen=max_length, padding='post')
```

```
[35]: # Create training dataset
train_dataset = tf.data.Dataset.from_tensor_slices((source_seqs, target_seqs))
train_dataset = train_dataset.shuffle(buffer_size=len(source_seqs)).batch(16,
    ↪drop_remainder=True)

# Create validation dataset
val_sequences = ques_tokenizer.texts_to_sequences(val_q_sents)
val_sequences = pad_sequences(val_sequences, maxlen=max_length, padding='post')
val_target_sequences = ans_tokenizer.texts_to_sequences(val_a_sents)
```

```

val_target_sequences = pad_sequences(val_target_sequences, maxlen=max_length,
    ↪padding='post')
val_dataset = tf.data.Dataset.from_tensor_slices((val_sequences,
    ↪val_target_sequences))
val_dataset = val_dataset.batch(16, drop_remainder=True)

# Create test dataset
test_sequences = ques_tokenizer.texts_to_sequences(test_q_sents)
test_sequences = pad_sequences(test_sequences, maxlen=max_length,
    ↪padding='post')
test_target_sequences = ans_tokenizer.texts_to_sequences(test_a_sents)
test_target_sequences = pad_sequences(test_target_sequences, maxlen=max_length,
    ↪padding='post')
test_dataset = tf.data.Dataset.from_tensor_slices((test_sequences,
    ↪test_target_sequences))
test_dataset = test_dataset.batch(16, drop_remainder=True)

```

```

[36]: # Print sizes of the datasets
print(f"Training set size: {len(train_q_sents)}")
print(f"Validation set size: {len(val_q_sents)}")
print(f"Test set size: {len(test_q_sents)}")

```

Training set size: 154893
 Validation set size: 20652
 Test set size: 30980

Model Training

```

[57]: import tensorflow as tf
from tensorflow.keras.layers import Layer, Input, Embedding, Dense,
    ↪LayerNormalization, Dropout, MultiHeadAttention
from tensorflow.keras.models import Model

def positional_encoding(max_len, d_model):
    position = tf.range(max_len, dtype=tf.float32)[: , tf.newaxis]
    div_term = tf.exp(tf.range(0, d_model, 2, dtype=tf.float32) * -(tf.math.
    ↪log(10000.0) / d_model))
    angle_rads = position * div_term
    sines = tf.sin(angle_rads)
    cosines = tf.cos(angle_rads)
    pos_encoding = tf.concat([sines, cosines], axis=-1)
    pos_encoding = pos_encoding[tf.newaxis, ...]
    return tf.cast(pos_encoding, dtype=tf.float32)

class PositionalEncodingLayer(Layer):
    def __init__(self, max_len, d_model):
        super(PositionalEncodingLayer, self).__init__()

```

```

        self.pos_encoding = positional_encoding(max_len, d_model)

    def call(self, inputs):
        seq_len = tf.shape(inputs)[1]
        return inputs + self.pos_encoding[:, :seq_len, :]

class TransformerBlock(Layer):
    def __init__(self, d_model, num_heads, dff, rate=0.2): # Adjusted dropout
        super(TransformerBlock, self).__init__()
        self.attention = MultiHeadAttention(num_heads=num_heads,
        ↪key_dim=d_model)
        self.ffn = tf.keras.Sequential([
            Dense(dff, activation='relu'),
            Dense(d_model)
        ])
        self.layernorm1 = LayerNormalization(epsilon=1e-6)
        self.layernorm2 = LayerNormalization(epsilon=1e-6)
        self.dropout1 = Dropout(rate)
        self.dropout2 = Dropout(rate)

    def call(self, inputs, training, mask):
        attn_output = self.attention(inputs, inputs, inputs,
        ↪attention_mask=mask)
        attn_output = self.dropout1(attn_output, training=training)
        out1 = self.layernorm1(inputs + attn_output)
        ffn_output = self.ffn(out1)
        ffn_output = self.dropout2(ffn_output, training=training)
        return self.layernorm2(out1 + ffn_output)

def build_transformer_model(input_vocab_size, output_vocab_size, max_len,
        ↪d_model, num_heads, dff, rate=0.2):
    inputs = Input(shape=(None,))
    x = Embedding(input_vocab_size, d_model)(inputs)
    x = PositionalEncodingLayer(max_len, d_model)(x)

    # Increase model complexity by adding more transformer blocks
    for _ in range(4): # Increasing depth
        x = TransformerBlock(d_model, num_heads, dff, rate)(x, training=True,
        ↪mask=None)

    outputs = Dense(output_vocab_size, activation='softmax')(x)
    return Model(inputs=inputs, outputs=outputs)

# Hyperparameters
max_len = 50
d_model = 768 # Increased dimension
num_heads = 12 # Increased heads

```

```

dff = 3072 # Increased FFN
rate = 0.2 # Dropout increased

# Vocabulary sizes (example sizes)
input_vocab_size = 10000
output_vocab_size = 10000

# Updated learning rate with decay schedule
initial_learning_rate = 1e-3
lr_schedule = tf.keras.optimizers.schedules.ExponentialDecay(
    initial_learning_rate,
    decay_steps=1000,
    decay_rate=0.96,
    staircase=True
)

optimizer = tf.keras.optimizers.Adam(learning_rate=lr_schedule)

# Build and compile model
transformer_model = build_transformer_model(input_vocab_size,
    ↪output_vocab_size, max_len, d_model, num_heads, dff, rate)
transformer_model.compile(optimizer=optimizer,
    ↪loss='sparse_categorical_crossentropy', metrics=['accuracy'])
transformer_model.summary()

# Assume train_dataset and val_dataset are properly prepared
EPOCHS = 2
BATCH_SIZE = 128 # Changed batch size
transformer_model.fit(train_dataset, epochs=EPOCHS, validation_data=val_dataset)

```

Model: "functional_16"

Layer (type)	Output Shape	
↪Param #		
input_layer_12 (InputLayer)	(None, None)	↪
↪ 0		
embedding_5 (Embedding)	(None, None, 768)	↪
↪7,680,000		
positional_encoding_layer_3	(None, None, 768)	↪
↪ 0		
(PositionalEncodingLayer)		↪
↪		

```

transformer_block_9                (None, None, 768)      ㄣ
↳33,065,472
(TransformerBlock)                  ㄣ
↳

transformer_block_10               (None, None, 768)      ㄣ
↳33,065,472
(TransformerBlock)                  ㄣ
↳

transformer_block_11               (None, None, 768)      ㄣ
↳33,065,472
(TransformerBlock)                  ㄣ
↳

transformer_block_12               (None, None, 768)      ㄣ
↳33,065,472
(TransformerBlock)                  ㄣ
↳

dense_94 (Dense)                   (None, None, 10000)    ㄣ
↳7,690,000

```

Total params: 147,631,888 (563.17 MB)

Trainable params: 147,631,888 (563.17 MB)

Non-trainable params: 0 (0.00 B)

```

Epoch 1/2
9680/9680          550s 53ms/step
- accuracy: 0.7130 - loss: nan - val_accuracy: 0.7178 - val_loss: nan
Epoch 2/2
9680/9680          503s 52ms/step
- accuracy: 0.7153 - loss: nan - val_accuracy: 0.7178 - val_loss: nan

```

[57]: <keras.src.callbacks.history.History at 0x78b5e8786a10>

[59]: !pip install rouge-score

```

Collecting rouge-score
  Downloading rouge_score-0.1.2.tar.gz (17 kB)
  Preparing metadata (setup.py) ... done
Requirement already satisfied: absl-py in /usr/local/lib/python3.10/dist-
packages (from rouge-score) (1.4.0)

```

Requirement already satisfied: nltk in /usr/local/lib/python3.10/dist-packages (from rouge-score) (3.8.1)
 Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages (from rouge-score) (1.26.4)
 Requirement already satisfied: six>=1.14.0 in /usr/local/lib/python3.10/dist-packages (from rouge-score) (1.16.0)
 Requirement already satisfied: click in /usr/local/lib/python3.10/dist-packages (from nltk->rouge-score) (8.1.7)
 Requirement already satisfied: joblib in /usr/local/lib/python3.10/dist-packages (from nltk->rouge-score) (1.4.2)
 Requirement already satisfied: regex>=2021.8.3 in /usr/local/lib/python3.10/dist-packages (from nltk->rouge-score) (2024.9.11)
 Requirement already satisfied: tqdm in /usr/local/lib/python3.10/dist-packages (from nltk->rouge-score) (4.66.5)
 Building wheels for collected packages: rouge-score
 Building wheel for rouge-score (setup.py) ... done
 Created wheel for rouge-score: filename=rouge_score-0.1.2-py3-none-any.whl size=24935 sha256=cd23497c209b8272f9c8dc93501a9d984972fbfefb1e0076fd62d4d037f7adf9
 Stored in directory: /root/.cache/pip/wheels/5f/dd/89/461065a73be61a532ff8599a28e9beef17985c9e9c31e541b4
 Successfully built rouge-score
 Installing collected packages: rouge-score
 Successfully installed rouge-score-0.1.2

```
[60]: from rouge_score import rouge_scorer
import numpy as np

reference_texts = ["The cat sat on the mat.", "There is a dog in the house."]
    # Example ground truth summaries
candidate_texts = ["A cat sat on the mat.", "A dog is inside the house."]
    # Example model-generated summaries

# Initialize the ROUGE scorer
scorer = rouge_scorer.RougeScorer(['rouge1', 'rouge2', 'rougeL'],
    use_stemmer=True)

# Calculate ROUGE scores for each summary pair
scores = [scorer.score(ref, cand) for ref, cand in zip(reference_texts,
    candidate_texts)]

# Calculate average ROUGE scores
avg_rouge1 = np.mean([score['rouge1'].fmeasure for score in scores])
avg_rouge2 = np.mean([score['rouge2'].fmeasure for score in scores])
avg_rougeL = np.mean([score['rougeL'].fmeasure for score in scores])

print(f"Average ROUGE-1 Score: {avg_rouge1}")
```



```
print(f"Average ROUGE-2 Score: {avg_rouge2}")
print(f"Average ROUGE-L Score: {avg_rougeL}")
```

Average ROUGE-1 Score: 0.8012820512820513

Average ROUGE-2 Score: 0.5818181818181819

Average ROUGE-L Score: 0.7243589743589743

```
[61]: import tensorflow as tf

class CustomSchedule(tf.keras.optimizers.schedules.LearningRateSchedule):
    def __init__(self, d_model, warmup_steps=4000):
        super(CustomSchedule, self).__init__()

        self.d_model = tf.cast(d_model, tf.float32)
        self.warmup_steps = tf.cast(warmup_steps, tf.float32)

    def __call__(self, step):
        step = tf.cast(step, tf.float32)
        arg1 = tf.math.rsqrt(step)
        arg2 = step * (self.warmup_steps ** -1.5)

        return tf.math.rsqrt(self.d_model) * tf.math.minimum(arg1, arg2)

# Instantiate and use the learning rate schedule
d_model = 128 # Example value for d_model
learning_rate = CustomSchedule(d_model)
optimizer = tf.keras.optimizers.Adam(learning_rate,
                                      beta_1=0.9,
                                      beta_2=0.98,
                                      epsilon=1e-9)

# Optional: To visualize how the learning rate changes over steps
import matplotlib.pyplot as plt

steps = range(1, 10000)
lrs = [learning_rate(step) for step in steps]

plt.plot(steps, lrs)
plt.ylabel("Learning Rate")
plt.xlabel("Train Step")
plt.title("Learning Rate Schedule")
plt.show()
```

