ex

#### March 17, 2024

## 0.0.1 Loading the data

```
# Define a function to read a JSONL file
def read_jsonl(file_path):
    sentence_id = []; summary = []; sentence = []
    with jsonlines.open(file_path, 'r') as reader:
        for line in reader:
            sentence_id.append(line['id'])
            summary.append(line['summaries'][0])
            sentence.append(line['text'])
        return sentence_id,summary,sentence

# Example usage
file_path = 'newsroom.jsonl'
s_id,summary,sentence = read_jsonl(file_path)
```

### 0.0.2 Training Data

```
[12]: columns = ['s_id', 'sentence']
data = df[columns]
data.head()
```

```
[12]: s_id sentence

0 newsroom-val-title-0 Real Madrid have confirmed they have agreed to...

1 newsroom-val-title-1 American Pie singer Don McLean was arrested on...

2 newsroom-val-title-2 A candidate for governor of the northern Mexic...

3 newsroom-val-title-3 Bill Parcells, the two-time Super Bowl-winning...

4 newsroom-val-title-4 IBM's data crunching service for the healthcar...
```

# 0.0.3 Preprocessing the Data

```
[13]: import nltk
    from nltk.corpus import stopwords
    from nltk.tokenize import word_tokenize
    from nltk.stem import PorterStemmer
```

```
import string
      # Download NLTK resources
      nltk.download('punkt')
      nltk.download('stopwords')
      # Initialize stemmer and set of stopwords
      stemmer = PorterStemmer()
      stop_words = set(stopwords.words('english'))
      def preprocess text(text):
          # Convert text to lowercase
          text = text.lower()
          # Tokenization
          tokens = word_tokenize(text)
          # Join tokens back into a single string
          processed_text = ' '.join(tokens)
          return processed_text
      # Apply the preprocessing function to the 'sentence' column of the DataFrame
      data['processed_sentence'] = data['sentence'].apply(preprocess_text)
      # Display the DataFrame with the processed sentences
      data.head()
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk data]
                   Package punkt is already up-to-date!
     [nltk_data] Downloading package stopwords to /root/nltk_data...
                  Package stopwords is already up-to-date!
     [nltk_data]
[13]:
                         s_id
                                                                         sentence \
      O newsroom-val-title-O Real Madrid have confirmed they have agreed to ...
      1 newsroom-val-title-1 American Pie singer Don McLean was arrested on...
      2 newsroom-val-title-2 A candidate for governor of the northern Mexic...
      3 newsroom-val-title-3 Bill Parcells, the two-time Super Bowl-winning...
      4 newsroom-val-title-4 IBM's data crunching service for the healthcar...
                                        processed sentence
      O real madrid confirm agre sign mexican striker ...
      1 american pie singer mclean arrest misdemeanor ...
      2 candid governor northern mexican state tamauli...
      3 bill parcel two-tim super bowl-win coach rejoi...
      4 ibm 'data crunch servic healthcar industri wa...
```

```
[14]: print(data['sentence'][0])
```

Real Madrid have confirmed they have agreed to sign the Mexican striker Javier Hernández on a season-long loan from Manchester United.

```
[15]: print(data['processed_sentence'][0])
```

real madrid confirm agre sign mexican striker javier hernández season-long loan manchest unit

## 0.0.4 Creating word embeddings

```
[16]: import numpy as np

def load_glove_embeddings(file_path):
    embeddings_index = {}
    with open(file_path, 'r', encoding='utf-8') as f:
        for line in f:
            values = line.split()

            embeddings_index[word] = coefs
        return embeddings_index

word_embeddings = load_glove_embeddings(glove_path)
    embedding_dim = len(next(iter(word_embeddings.values())))
    print("Embedding_Dimension:", embedding_dim)
```

Embedding Dimension: 100

# [17]: print(word\_embeddings['the'])

```
[-0.038194 -0.24487
                    0.72812 -0.39961
                                       0.083172 0.043953 -0.39141
 0.3344
         -0.57545
                    0.087459 0.28787 -0.06731
                                                 0.30906 -0.26384
-0.13231 -0.20757
                    0.33395 -0.33848 -0.31743 -0.48336
                                                          0.1464
-0.37304
         0.34577
                    0.052041 0.44946 -0.46971
                                                 0.02628 -0.54155
-0.15518 -0.14107 -0.039722 0.28277
                                       0.14393
                                                0.23464 -0.31021
 0.086173 0.20397
                    0.52624
                              0.17164 -0.082378 -0.71787 -0.41531
 0.20335 -0.12763
                    0.41367
                              0.55187
                                       0.57908 -0.33477 -0.36559
-0.54857 -0.062892 0.26584
                                       0.99775 -0.80481 -3.0243
                             0.30205
 0.01254 -0.36942
                              0.72201 -0.24978
                                                0.92136
                                                          0.034514
                    2.2167
 0.46745
           1.1079
                   -0.19358 -0.074575 0.23353 -0.052062 -0.22044
 0.057162 -0.15806 -0.30798 -0.41625
                                       0.37972
                                                0.15006 - 0.53212
-0.2055
          -1.2526
                    0.071624 0.70565
                                       0.49744 -0.42063
                                                          0.26148
                                       0.37104 -0.25217
                                                          0.016215
-1.538
          -0.30223 -0.073438 -0.28312
-0.017099 -0.38984
                    0.87424 -0.72569 -0.51058 -0.52028 -0.1459
 0.8278
          0.27062 ]
```

```
[18]: import numpy as np
    from tensorflow.keras.preprocessing.sequence import pad_sequences
    from gensim.models import KeyedVectors
    word_vectors = word_embeddings
    # Function to convert a sentence to word embeddings
    def sentence_to_embeddings(sentence, word_vectors, embedding_dim):
       words = sentence.split()
       embeddings = []
       for word in words:
          if word in word vectors:
             embeddings.append(word_vectors[word])
          else:
             # If word not in vocabulary, use zero vector
             embeddings.append(np.zeros(embedding_dim))
       return embeddings
    # Display the preprocessed data
    data.head()
[18]:
                                                     sentence \
                  s_id
    O newsroom-val-title-O Real Madrid have confirmed they have agreed to ...
    1 newsroom-val-title-1 American Pie singer Don McLean was arrested on...
    2 newsroom-val-title-2 A candidate for governor of the northern Mexic...
    3 newsroom-val-title-3 Bill Parcells, the two-time Super Bowl-winning...
    4 newsroom-val-title-4 IBM's data crunching service for the healthcar...
                             processed_sentence \
    O real madrid confirm agre sign mexican striker ...
    1 american pie singer mclean arrest misdemeanor ...
    2 candid governor northern mexican state tamauli...
    3 bill parcel two-tim super bowl-win coach rejoi...
    4 ibm 'data crunch servic healthcar industri wa...
                                   embeddings \
    padded_embeddings
```

```
[28]: | temp = df['summary']
     temp = pd.DataFrame(temp)
     temp.head()
[28]:
                                                  summary
     O Real Madrid sign Javier Hernández on loan from...
     1 American Pie singer Don Mclean arrested on dom...
     2 Candidate for governor of Mexican state of Tam...
              Bill Parcells rejoining ESPN for third time
     4 IBM Watson Health now counts CVS Health as a p...
[33]: word_vectors = word_embeddings
      # Function to convert a sentence to word embeddings
     def sentence_to_embeddings(sentence, word_vectors, embedding_dim):
         words = sentence.split()
         embeddings = []
         for word in words:
             if word in word vectors:
                 embeddings.append(word_vectors[word])
             else:
                 # If word not in vocabulary, use zero vector
                 embeddings.append(np.zeros(embedding_dim))
         return embeddings
     max_seq_length = 50 # Maximum sequence length for padding/truncation
     temp['s_padded_embeddings'] = pad_sequences(temp['s_embeddings'],_
       maxlen=max_seq_length, padding='post', truncating='post').tolist()
      # Display the preprocessed data
     temp.head()
[33]:
                                                  summary \
     O Real Madrid sign Javier Hernández on loan from...
     1 American Pie singer Don Mclean arrested on dom...
     2 Candidate for governor of Mexican state of Tam...
              Bill Parcells rejoining ESPN for third time
     3
     4 IBM Watson Health now counts CVS Health as a p...
                                        processed_summary \
     O real madrid sign javier hernández loan manches...
     1 american pie singer mclean arrest domest viole...
     2 candid governor mexican state tamaulipa kill s...
     3
                       bill parcel rejoin espn third time
```

```
4
             ibm watson health count cv health partner
                                     s embeddings \
    0 [[0.45006, 0.15098, 0.31014, -0.20369, -0.2210...
    1 [[0.38666, 0.64827, 0.72807, -0.077056, 0.1545...
    2 [[-0.33871, -0.37143, 0.4443, 0.72357, -0.3119...
    3 [[-0.10535, -0.025048, 0.55525, -1.0371, 0.221...
    4 [[0.4875, 0.4214, 0.013491, 0.71504, 0.3708, -...
                               s_padded_embeddings
    1 [[0, 0, 0, 0, 0, 0, -1, 0, 0, 0, 0, 0, 0, 0...
    3 [[0, 0, 0, -1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0...
    [35]: final.columns
[35]: Index(['s_id', 'sentence', 'processed_sentence', 'embeddings',
          'padded_embeddings', 'summary', 'processed_summary', 's_embeddings',
          's_padded_embeddings'],
         dtype='object')
```

## 0.0.5 Running the Model & Evaluation Metrics

```
[121]: import numpy as np
       from tensorflow.keras.layers import Input, Embedding, LSTM, Dense
       # Tokenization
       max_words = 10000 # Maximum number of words to tokenize
       max_seq_len = 100 # Maximum sequence length for padding
       tokenizer = Tokenizer(num words=max words)
       tokenizer.fit_on_texts(final['processed_sentence'])
       tokenizer.fit_on_texts(final['processed_summary'])
       # Padding sequences to make them of uniform length
       sentence_sequences = pad_sequences(sentence_sequences, maxlen=max_seq_len,_u
        →padding='post')
       summary_sequences = pad_sequences(summary_sequences, maxlen=max_seq_len,__
        →padding='post')
       # Model architecture
       embedding_dim = 100  # Dimension of word embeddings
       hidden_units = 128
                            # Number of units in LSTM layer
       # Embedding layers
```

```
embedding_layer = Embedding(input_dim=max_words, output_dim=embedding_dim)

# Embedding lookup for sentence and summary
sentence_embedding = embedding_layer(sentence_input)
summary_embedding = embedding_layer(summary_input)

# RNN for sentence and summary
sentence_rnn = lstm_layer(sentence_embedding)
summary_rnn = lstm_layer(summary_embedding)
```

```
Epoch 1/10
7/7 [=========== ] - 9s 1s/step - loss: 9.1619 - accuracy:
0.6582 - val_loss: 9.0024 - val_accuracy: 0.9184
Epoch 2/10
7/7 [=========== ] - 8s 1s/step - loss: 8.3912 - accuracy:
0.9211 - val_loss: 7.4508 - val_accuracy: 0.9184
Epoch 3/10
7/7 [============ ] - 7s 1s/step - loss: 6.6197 - accuracy:
0.9211 - val_loss: 5.4238 - val_accuracy: 0.9184
Epoch 4/10
7/7 [============ ] - 7s 1s/step - loss: 4.3803 - accuracy:
0.9211 - val_loss: 3.0768 - val_accuracy: 0.9184
0.9211 - val_loss: 1.3990 - val_accuracy: 0.9184
Epoch 6/10
0.9211 - val_loss: 0.9076 - val_accuracy: 0.9184
Epoch 7/10
0.9211 - val loss: 0.8369 - val accuracy: 0.9184
Epoch 8/10
```

```
7/7 [========== ] - 7s 1s/step - loss: 0.7835 - accuracy:
            0.9211 - val_loss: 0.8246 - val_accuracy: 0.9184
            Epoch 9/10
            0.9211 - val_loss: 0.8179 - val_accuracy: 0.9184
            Epoch 10/10
            0.9211 - val_loss: 0.8114 - val_accuracy: 0.9184
[122]: <keras.src.callbacks.History at 0x7d9fc3d53af0>
[143]: from nltk.translate.bleu_score import sentence_bleu
             from nltk.translate.bleu_score import SmoothingFunction
             from rouge import Rouge
             from tensorflow.keras.preprocessing.sequence import pad_sequences
             # Define a function to preprocess the input sentence and expected summary
             def preprocess_input_and_summary(input_sentence, expected_summary):
                     processed_input_sentence = preprocess_text(input_sentence)
                     processed_expected_summary = preprocess_text(expected_summary)
                     return processed_input_sentence, processed_expected_summary
             # Define a function to decode the summary sequence
             def decode_summary(summary_sequence, tokenizer):
                     decoded_summary = tokenizer.sequences_to_texts(summary_sequence)
                     decoded_summary = [sentence.split() for sentence in decoded_summary]
                     decoded_summary = [' '.join(sentence) for sentence in decoded_summary]
                     return decoded_summary[0]
             def compress(input_sentence, expected_summary):
                     # Preprocess input sentence and expected summary
                     processed input sentence, processed expected summary = 11
                General content of the second content o
                     # Compute evaluation metrics
                     bleu_score = sentence_bleu([processed_expected_summary.split()],_
                decoded_summary.split(), smoothing_function=SmoothingFunction().method1)
                     rouge = Rouge()
                     rouge_scores = rouge.get_scores(decoded_summary, processed_expected_summary)
                     print("Summary:", decoded_summary)
                     # Print evaluation metrics
                     print("BLEU Score:", bleu_score)
```

```
[]: # Sample data
     sentences = [
         "New jobless numbers are a bit of a mixed bag for President Obama and \operatorname{his}_\sqcup
      ⇔reelection bid."]
     # Preprocessing
     stop_words = set(stopwords.words('english'))
     def preprocess_text(text):
         text = text.lower() # Convert to lowercase
         text = re.sub(r'[^a-zA-Z\s]', '', text) # Remove non-alphabetic characters
         tokens = word_tokenize(text) # Tokenization
         filtered tokens = [word for word in tokens if word not in stop words] #__
      \hookrightarrowRemove stopwords
         return ' '.join(filtered_tokens)
     # Tokenization and preprocessing
     sentences = [preprocess_text(sentence) for sentence in sentences]
     # Tokenization
     tokenizer = tf.keras.preprocessing.text.Tokenizer()
     tokenizer.fit_on_texts(sentences)
     sequences = tokenizer.texts_to_sequences(sentences)
     # Padding sequences for uniform input length
     max_seq_len = max([len(seq) for seq in sequences])
     padded_sequences = pad_sequences(sequences, maxlen=max_seq_len, padding='post')
     # Model parameters
     vocab size = len(tokenizer.word index) + 1
     embedding_dim = 64
     hidden_units = 128
     # Define models
     models = {
         "LSTM": Sequential([
             Embedding(vocab_size, embedding_dim, input_length=max_seq_len),
             LSTM(hidden_units, return_sequences=True),
             Dense(vocab_size, activation='softmax')
         ])
     }
     # Compile and train models
     for model_name, model in models.items():
         model.compile(loss='sparse_categorical_crossentropy', optimizer=Adam(lr=0.
      ⇔001), metrics=['accuracy'])
         print(f"Training {model_name}...")
```

```
model.fit(padded_sequences, np.expand_dims(padded_sequences, -1),_
      ⇔epochs=100, verbose=0)
        print(f"{model_name} trained.\n")
     # Example compression
    test sentences = [
        "New jobless numbers are a bit of a mixed bag for President Obama and \operatorname{his}_\sqcup
      ⇔reelection bid."]
    test_sentences = [preprocess_text(sentence) for sentence in test_sentences]
    test_sequences = tokenizer.texts_to_sequences(test_sentences)
    test_padded_sequences = pad_sequences(test_sequences, maxlen=max_seq_len,_u
      ⇔padding='post')
[3]: for model_name, model in models.items():
        print(f"\n{model_name} compressed sentences:")
        for i, test_sentence in enumerate(test_sentences):
            compressed_sequence = model.predict(test_padded_sequences[i:i+1])
            compressed_sequence = np.argmax(compressed_sequence, axis=-1)
            compressed_sentence = ' '.join([tokenizer.index_word[idx] for idx in_
      print(f"{i+1}. {compressed_sentence}")
    WARNING:absl:`lr` is deprecated in Keras optimizer, please use `learning_rate`
    or use the legacy optimizer, e.g., tf.keras.optimizers.legacy.Adam.
    Training LSTM...
    LSTM trained.
    LSTM compressed sentences:
    1/1 [======] - Os 484ms/step
    1. New jobless numbers a mixed bag for Obama
[]:
```