

NIRF-2024 Engineering Rank Band (151-200) Pharmacy Rank - 77 Innovation Rank Band (11-50)











INTRODUCTION TO AI(AI101B)

Even Semester Session 2024-25

LANGUAGE TRANSLATION WITH SEQUENCE MODEL

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Introduction

Language is the foundation of communication, and the ability to automatically translate one language to another is a significant milestone in artificial intelligence and natural language processing. This project explores the use of sequence models, particularly sequence-to-sequence (Seq2Seq) models with LSTM layers, to perform language translation. By training on English–French sentence pairs, the model learns to map input sequences in French to target sequences in English. The project demonstrates how neural networks can be used to learn patterns in linguistic data and apply them to real-world translation tasks.

Objective of the Project

The objective of this project is to build a machine learning-based model that translates English sentences into French using a sequence-to-sequence architecture. The model utilizes LSTM layers to capture the temporal dependencies in sentence structures and generates meaningful translations. This project aims to train the model on real French-English translation datasets.

Methodology

1.Data Collection: A dataset containing English–French sentence pairs was used. The source is the ManyThings.org version of the Tatoeba dataset.

2. Data Preprocessing:

- 1. Lowercasing, trimming, and adding special start/end tokens.
- 2. Tokenization and padding of input and output sequences.

3. Model Architecture:

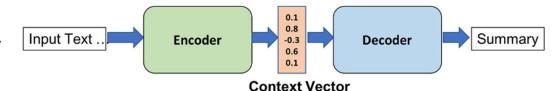
- 1. Encoder: Embedding layer + LSTM
- 2. Decoder: Embedding + LSTM + Dense layer with Softmax

4.Training:

- 1. Teacher forcing technique
- 2. Loss function: Categorical crossentropy
- 3. Optimizer: RMSProp

5.Inference:

- 1. Encoder encodes input sentence into states.
- 2. Decoder predicts one word at a time until token.



Code

```
[1] import string
     import re
     import numpy as np
     from numpy import array, argmax, random, take
     import pandas as pd
[2] from tensorflow.keras.models import Sequential
     from tensorflow.keras.layers import Dense, LSTM, Embedding, RepeatVector
     from tensorflow.keras.preprocessing.text import Tokenizer
     from tensorflow.keras.callbacks import ModelCheckpoint
     from tensorflow.keras.preprocessing.sequence import pad sequences
     from tensorflow.keras.models import load model
     from tensorflow.keras import optimizers
[3] import matplotlib.pyplot as plt
     %matplotlib inline
     pd.set_option('display.max_colwidth',200)
[5] data_path='fra.txt'
     with open(data_path, 'r', encoding='utf-8') as f:
      lines = f.read()
     lines
'Go.\tVa !\tCC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #1158250 (Wittydev)\nGo.\tMarc
    ba.org #2877272 (CM) & #8090732 (Micsmithel)\nGo.\tEn route !\tCC-BY 2.0 (France) Attribution: tatoeba
    o.\tBouge !\tCC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #9022935 (Micsmithel)\nHi.\tS
```

```
[6] def to lines(text):
      sents=text.strip().split('\n')
      sents=[i.split('\t') for i in sents ]
       return sents
[7] fra eng=to lines(lines)
    fra eng[:5]
→ [['Go.',
       'Va !',
      'CC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #1158250 (Wittydev)'],
      ['Go.',
       'Marche.',
      'CC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #8090732 (Micsmithel)'],
      ['Go.',
      'En route!',
      'CC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #8267435 (felix63)'],
      ['Go.',
       'Bouge!',
      'CC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #9022935 (Micsmithel)'],
     ['Hi.',
       'Salut !',
      'CC-BY 2.0 (France) Attribution: tatoeba.org #538123 (CM) & #509819 (Aiji)']]
[8] fra eng=array(fra eng)
    fra_eng[:5]
→ array([['Go.', 'Va !',
             'CC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #1158250 (Wittydev)'],
           ['Go.', 'Marche.',
            'CC-BY 2.0 (France) Attribution: tatoeba.org #2877272 (CM) & #8090732 (Micsmithel)'],
                     Connected to Python 3 Google Compute Engine backend (GPU)
```

```
fra_eng.shape
(232736, 3)
fra_eng=fra_eng[:50000,:]
fra eng=fra eng[:,[0,1]]
fra_eng[:5]
array([['Go.', 'Va !'],
       ['Go.', 'Marche.'],
['Go.', 'En route !'],
       ['Go.', 'Bouge !'],
['Hi.', 'Salut !']], dtype='<U349')
fra_eng[:,0]=[s.translate(str.maketrans('','',string.punctuation)) for s in fra_eng[:,
fra eng[:,1]=[s.translate(str.maketrans('','',string.punctuation)) for s in fra eng[:,
fra eng[:5]
array([['Go', 'Va '],
       ['Go', 'Marche'],
       ['Go', 'En route '],
       ['Go', 'Bouge '],
       ['Hi', 'Salut ']], dtype='<U349')
for i in range(len(fra_eng)):
  fra_eng[i,0]=fra_eng[i,0].lower()
  fra_eng[i,1]=fra_eng[i,1].lower()
```

```
[14] def tokenization(lines):
         tokenizer=Tokenizer()
         tokenizer.fit_on_texts(lines)
         return tokenizer
       eng tokenizer=tokenization(fra eng[:,0])
       eng_vocab_size=len(eng_tokenizer.word_index)+1
       eng length=8
       print('English vocabulary size:', eng vocab size)
   → English vocabulary size: 1907
  [15] fra_tokenizer=tokenization(fra_eng[:,1])
       fra vocab size=len(fra tokenizer.word index)+1
       fra length=8
       print('French Vocabulary Size: ', fra vocab size)
   French Vocabulary Size: 4422
  [16] def encode sequences(tokenizer,length, lines):
         seq=tokenizer.texts to sequences(lines)
         seq=pad sequences(seq,maxlen=length,padding='post')
         return seq
  [17] from sklearn.model selection import train test split
       train, test= train test split(fra eng,test size=0.2,random state=12)
```

```
trainX=encode sequences(fra tokenizer,fra length,train[:,1])
trainY= encode_sequences(eng_tokenizer,eng_length,train[:,0])
testX=encode sequences(fra tokenizer,fra length,test[:,1])
testY=encode sequences(eng tokenizer, eng length, test[:,0])
def define_model(in_vocab,out_vocab, in_timesteps,out_timesteps, units):
  model=Sequential()
  model.add(Embedding(in vocab, units, input length=in timesteps, mask zero=True))
 model.add(LSTM(units))
  model.add(RepeatVector(out timesteps))
  model.add(LSTM(units, return sequences=True))
  model.add(Dense(out vocab, activation='softmax'))
  return model
model=define model(fra vocab size, eng vocab size, fra length, eng length, 512)
rms=optimizers.RMSprop(learning rate=0.001)
model.compile(optimizer=rms,loss='sparse_categorical crossentropy')
/usr/local/lib/python3.11/dist-packages/keras/src/layers/core/embedding.py:90: UserWarning: Argument `input length` is deprecate
 warnings.warn(
history=model.fit(trainX, trainY.reshape(trainY.shape[0], trainY.shape[1],1), epochs=30, batch size=512, validation split=0.2)
Epoch 1/30
63/63 -
                          20s 219ms/step - loss: 5.6975 - val loss: 3.1559
Epoch 2/30
63/63 -
                          13s 212ms/step - loss: 2.9637 - val loss: 2.7826
```

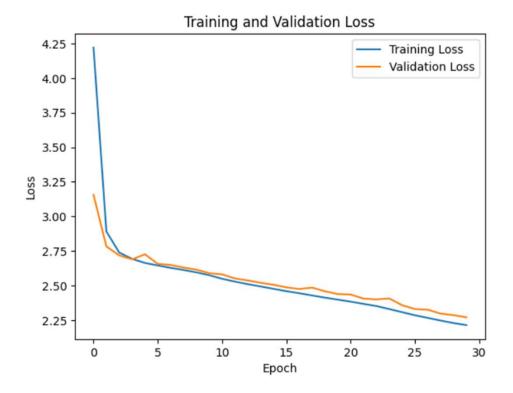
```
[22] preds_probs = model.predict(testX)
     preds = np.argmax(preds_probs, axis=-1)
→ 58/58 -
                          1s 14ms/step
[23] preds
\rightarrow array([[ 1, 5, 0, ..., 0, 0, 0],
           [1, 7, 0, ..., 0, 0, 0],
           [2, 2, 0, ..., 0, 0, 0],
           [2, 2, 0, \ldots, 0, 0, 0],
           [1, 6, 0, \ldots, 0, 0, 0],
           [1, 32, 0, \ldots, 0, 0, 0]
[24] def get_word(n,tokenizer):
       for word,index in tokenizer.word index.items():
         if index==n:
           return word
       return None
```

```
preds text=[]
for i in preds:
  temp=[]
 for j in range(len(i)):
    t=get_word(i[j],eng_tokenizer)
   if j>0:
      if(t==get_word(i[j-1], eng_tokenizer)) or (t==None):
        temp.append('')
      else:
        temp.append(t)
    else:
     if(t==None):
       temp.append('')
      else:
        temp.append(t)
  preds_text.append(' '.join(temp))
pred df=pd.DataFrame({'actual': test[:,0], 'predicted': preds text})
```

Output

pred_df.sample <mark>(</mark> 10)		
	actual	predicted
4522	are you courageous	are you
8109	that was a secret	he was a
5487	tom is in luck	tom is a
9812	dont be naive	were
3857	theyre very happy	theyre are
6536	everyone watched	youre not
2536	are we sinking	you
2001	whats it good for	we me
642	are you ambitious	are you
3001	i remain doubtful	i am

```
import matplotlib.pyplot as plt
plt.plot(history.history['loss'], label='Training Loss')
plt.plot(history.history['val_loss'], label='Validation Loss')
plt.title('Training and Validation Loss')
plt.xlabel('Epoch')
plt.ylabel('Loss')
plt.legend()
plt.show()
```



Outcome

This project successfully demonstrates how sequence models can be used for basic language translation. The model trained on sample data can generate understandable French translations for English sentences. The project lays the foundation for scaling up to more advanced NLP tasks like multilingual translation, attention mechanisms, or Transformer-based models.

References

- 1. Many Things.org English French Dataset: http://www.manythings.org/
- 2.TensorFlow Documentation https://www.tensorflow.org/
- 3. Keras Seq2Seq Guide https://keras.io/examples/nlp/lstm_seq2seq/