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# Part 2: Implementation (Coding)
# 3. Model Setup
# a. Dataset Preprocessing
# Use a few English-French pairs:
data = [("hello", "bonjour"), ("how are you", "comment ça va"),
        ("thank you", "merci"), ("good morning", "bonjour"), ("i love you", "je t'aime")]
# b. Tokenization and Padding
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences

eng_sentences, fr_sentences = zip(*data)
eng_tokenizer = Tokenizer()
fr_tokenizer = Tokenizer()

eng_tokenizer.fit_on_texts(eng_sentences)
fr_tokenizer.fit_on_texts(fr_sentences)

eng_seq = eng_tokenizer.texts_to_sequences(eng_sentences)
fr_seq = fr_tokenizer.texts_to_sequences(fr_sentences)

max_eng_len = max(len(seq) for seq in eng_seq)
max_fr_len = max(len(seq) for seq in fr_seq)

eng_seq = pad_sequences(eng_seq, maxlen=max_eng_len, padding='post')
fr_seq = pad_sequences(fr_seq, maxlen=max_fr_len, padding='post')
# c. Encoder-Decoder with Attention (Simplified)
import tensorflow as tf
from tensorflow.keras.layers import Embedding, LSTM, Dense, Input, Attention
from tensorflow.keras.models import Model

# Parameters
vocab_eng = len(eng_tokenizer.word_index) + 1
vocab_fr = len(fr_tokenizer.word_index) + 1
embed_dim = 64
lstm_units = 64

# Encoder
encoder_inputs = Input(shape=(max_eng_len,))
enc_emb = Embedding(vocab_eng, embed_dim)(encoder_inputs)
encoder_lstm, state_h, state_c = LSTM(lstm_units, return_sequences=True, return_state=True)(enc_emb)

# Decoder
decoder_inputs = Input(shape=(max_fr_len,))
dec_emb = Embedding(vocab_fr, embed_dim)(decoder_inputs)
decoder_lstm_out, _, _ = LSTM(lstm_units, return_sequences=True, return_state=True)(dec_emb, initial_state=[state_h, state_c])

# Attention
attention = Attention()([decoder_lstm_out, encoder_lstm])
from tensorflow.keras.layers import Concatenate

concat = Concatenate(axis=-1)([decoder_lstm_out, attention])

# Output
outputs = Dense(vocab_fr, activation='softmax')(concat)

model = Model([encoder_inputs, decoder_inputs], outputs)
model.compile(optimizer='adam', loss='sparse_categorical_crossentropy')
model.summary()
# d. Training
import numpy as np
decoder_output = np.expand_dims(fr_seq, -1)
model.fit([eng_seq, fr_seq], decoder_output, epochs=100)

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input_layer_2 (InputLayer)	(None, 3)	0	-
input_layer_3 (InputLayer)	(None, 3)	0	-
embedding_2 (Embedding)	(None, 3, 64)	640	input_layer_2[0]...
embedding_3 (Embedding)	(None, 3, 64)	512	input_layer_3[0]...
lstm_2 (LSTM)	[(None, 3, 64), (None, 64), (None, 64)]	33,024	embedding_2[0][0]
lstm_3 (LSTM)	[(None, 3, 64), (None, 64), (None, 64)]	33,024	embedding_3[0][0].. lstm_2[0][1], lstm_2[0][2]
attention_1 (Attention)	(None, 3, 64)	0	lstm_3[0][0], lstm_2[0][0]
concatenate (Concatenate)	(None, 3, 128)	0	lstm_3[0][0], attention_1[0][0]
dense (Dense)	(None, 3, 8)	1,032	concatenate[0][0]

Total params: 68,232 (266.53 KB)

Trainable params: 68,232 (266.53 KB)

Non-trainable params: 0 (0.00 B)

Epoch 1/100

1/1 ————— 5s 5s/step - loss: 2.0846

Epoch 2/100

1/1 ————— 0s 184ms/step - loss: 2.0741

Epoch 3/100

1/1 ————— 0s 140ms/step - loss: 2.0636

Epoch 4/100

1/1 ————— 0s 142ms/step - loss: 2.0530

Epoch 5/100

1/1 ————— 0s 145ms/step - loss: 2.0420

Epoch 6/100

1/1 ————— 0s 135ms/step - loss: 2.0306

Epoch 7/100

1/1 ————— 0s 114ms/step - loss: 2.0187

Epoch 8/100

1/1 ————— 0s 73ms/step - loss: 2.0060

Epoch 9/100

1/1 ————— 0s 157ms/step - loss: 1.9924

Epoch 10/100

1/1 ————— 0s 75ms/step - loss: 1.9779

Epoch 11/100

1/1 ————— 0s 80ms/step - loss: 1.9621

Epoch 12/100

1/1 ————— 0s 146ms/step - loss: 1.9450

Epoch 13/100

1/1 ————— 0s 89ms/step - loss: 1.9263

Epoch 14/100

1/1 ————— 0s 129ms/step - loss: 1.9059

Epoch 15/100

1/1 ————— 0s 91ms/step - loss: 1.8835

Epoch 16/100

1/1 ————— 0s 61ms/step - loss: 1.8588

Epoch 17/100

1/1 ————— 0s 65ms/step - loss: 1.8316

Epoch 18/100

1/1 ————— 0s 52ms/step - loss: 1.8016

Epoch 19/100

1/1 ————— 0s 60ms/step - loss: 1.7686

Epoch 20/100

1/1 ————— 0s 63ms/step - loss: 1.7323

Epoch 21/100

1/1 ————— 0s 62ms/step - loss: 1.6926

Epoch 22/100

1/1 ————— 0s 59ms/step - loss: 1.6493

Epoch 23/100











































1/1 ————— 0s 55ms/step - loss: 1.6025

Epoch 24/100

1/1 ————— 0s 55ms/step - loss: 1.5527

Epoch 25/100

1/1 ————— 0s 52ms/step - loss: 1.5005

Epoch 26/100  
1/1  0s 61ms/step - loss: 1.4471  
Epoch 27/100  
1/1  0s 64ms/step - loss: 1.3943  
Epoch 28/100  
1/1  0s 54ms/step - loss: 1.3441  
Epoch 29/100  
1/1  0s 53ms/step - loss: 1.2989  
Epoch 30/100  
1/1  0s 54ms/step - loss: 1.2607  
Epoch 31/100  
1/1  0s 56ms/step - loss: 1.2307  
Epoch 32/100  
1/1  0s 54ms/step - loss: 1.2090  
Epoch 33/100  
1/1  0s 53ms/step - loss: 1.1943  
Epoch 34/100  
1/1  0s 70ms/step - loss: 1.1840  
Epoch 35/100  
1/1  0s 57ms/step - loss: 1.1753  
Epoch 36/100  
1/1  0s 55ms/step - loss: 1.1656  
Epoch 37/100  
1/1  0s 53ms/step - loss: 1.1533  
Epoch 38/100  
1/1  0s 63ms/step - loss: 1.1376  
Epoch 39/100  
1/1  0s 59ms/step - loss: 1.1186  
Epoch 40/100  
1/1  0s 56ms/step - loss: 1.0970  
Epoch 41/100  
1/1  0s 54ms/step - loss: 1.0737  
Epoch 42/100  
1/1  0s 53ms/step - loss: 1.0496  
Epoch 43/100  
1/1  0s 60ms/step - loss: 1.0254  
Epoch 44/100  
1/1  0s 56ms/step - loss: 1.0019  
Epoch 45/100  
1/1  0s 146ms/step - loss: 0.9794  
Epoch 46/100  
1/1  0s 55ms/step - loss: 0.9581  
Epoch 47/100  
1/1  0s 54ms/step - loss: 0.9383  
Epoch 48/100  
1/1  0s 62ms/step - loss: 0.9199  
Epoch 49/100  
1/1  0s 60ms/step - loss: 0.9027  
Epoch 50/100  
1/1  0s 73ms/step - loss: 0.8867  
Epoch 51/100  
1/1  0s 53ms/step - loss: 0.8717  
Epoch 52/100  
1/1  0s 57ms/step - loss: 0.8572  
Epoch 53/100  
1/1  0s 58ms/step - loss: 0.8430  
Epoch 54/100  
1/1  0s 59ms/step - loss: 0.8286  
Epoch 55/100  
1/1  0s 57ms/step - loss: 0.8137  
Epoch 56/100  
1/1  0s 62ms/step - loss: 0.7981  
Epoch 57/100  
1/1  0s 60ms/step - loss: 0.7818  
Epoch 58/100  
1/1  0s 62ms/step - loss: 0.7651  
Epoch 59/100  
1/1  0s 59ms/step - loss: 0.7482  
Epoch 60/100  
1/1  0s 54ms/step - loss: 0.7317  
Epoch 61/100  
1/1  0s 55ms/step - loss: 0.7157  
Epoch 62/100  
1/1  0s 55ms/step - loss: 0.7005  
Epoch 63/100  
1/1  0s 60ms/step - loss: 0.6862  
Epoch 64/100  
1/1  0s 55ms/step - loss: 0.6727  
Epoch 65/100  
1/1  0s 55ms/step - loss: 0.6597  
Epoch 66/100  
1/1  0s 59ms/step - loss: 0.6472  
Epoch 67/100  
1/1  0s 69ms/step - loss: 0.6349  
Epoch 68/100