nlp-assignment

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[1]: import pandas as pd
     import torch
     import torch.nn as nn
     from torch.utils.data import Dataset, DataLoader
     from sklearn.model_selection import train_test_split
     import numpy as np
     from tqdm import tqdm
     import time
     import nltk
     nltk.download('punkt')
    [nltk_data] Downloading package punkt to
                    C:\Users\samee\AppData\Roaming\nltk_data...
    [nltk_data]
    [nltk_data]
                  Package punkt is already up-to-date!
[1]: True
[2]: # Check if CUDA is available
     device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
     print(f"Using device: {device}")
    Using device: cuda
[3]: # Load and preprocess data
     df = pd.read_csv(r'C:\Users\samee\OneDrive\Documents\SEM5\NLP\Assignment_\u
      →3\Hindi_English_Truncated_Corpus.csv') # Replace with your dataset path
     df.dropna(inplace=True)
     df = df.sample(frac=1, random_state=42) # Shuffle the data
     src_lang = df['english_sentence'].astype(str).tolist()
     tgt_lang = df['hindi_sentence'].astype(str).tolist()
[4]: def create_vocab(sentences):
         vocab = set()
         for sentence in sentences:
             vocab.update(str(sentence).split())
         return vocab
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[5]: src_vocab = create_vocab(src_lang)
      tgt_vocab = create_vocab(tgt_lang)
      src_vocab_size = len(src_vocab) + 1 # Add 1 for padding
      tgt_vocab_size = len(tgt_vocab) + 1 # Add 1 for padding
 [6]: # Create word to index mappings
      src_word2idx = {word: idx + 1 for idx, word in enumerate(src_vocab)} # Start_
      ⇔indexing from 1
      tgt_word2idx = {word: idx + 1 for idx, word in enumerate(tgt_vocab)} # Start_
      →indexing from 1
      src_word2idx['<PAD>'] = 0 # Padding index
      tgt_word2idx['<PAD>'] = 0 # Padding index
      src_idx2word = {idx: word for word, idx in src_word2idx.items()}
      tgt_idx2word = {idx: word for word, idx in tgt_word2idx.items()}
 [7]: # Convert sentences to indices
      def sentence_to_indices(sentence, word2idx):
          return [word2idx.get(word, 0) for word in str(sentence).split()]
 [8]: src_indices = [sentence_to_indices(sentence, src_word2idx) for sentence in__
      ⇔src_lang]
      tgt_indices = [sentence_to_indices(sentence, tgt_word2idx) for sentence in_u
       →tgt_lang]
 [9]: # Pad sequences
      max_src_len = max(len(s) for s in src_indices)
      max_tgt_len = max(len(s) for s in tgt_indices)
      src_indices = [s + [0] * (max_src_len - len(s)) for s in src_indices]
      tgt_indices = [s + [0] * (max_tgt_len - len(s)) for s in tgt_indices]
[10]: # Create dataset
      class TranslationDataset(Dataset):
          def __init__(self, src, tgt):
             self.src = src
              self.tgt = tgt
          def __len__(self):
             return len(self.src)
          def __getitem__(self, idx):
              return torch.tensor(self.src[idx]), torch.tensor(self.tgt[idx])
[11]: # Split data into train and test sets
      X_train, X_test, y_train, y_test = train_test_split(src_indices, tgt_indices, u
       →test_size=0.2, random_state=42)
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[12]: # Create dataloaders
      train_dataset = TranslationDataset(X_train, y_train)
      test_dataset = TranslationDataset(X_test, y_test)
      train_loader = DataLoader(train_dataset, batch_size=16, shuffle=True)
      test_loader = DataLoader(test_dataset, batch_size=16, shuffle=False)
[13]: # Define the Seq2Seq model using LSTM with embedding layers
      class Seq2SeqLSTM(nn.Module):
          def __init__(self, src_vocab_size, tgt_vocab_size, hidden_size):
              super(Seq2SeqLSTM, self).__init__()
              self.src_embedding = nn.Embedding(src_vocab_size, hidden_size)
              self.tgt_embedding = nn.Embedding(tgt_vocab_size, hidden_size)
              self.encoder = nn.LSTM(hidden_size, hidden_size, batch_first=True)
              self.decoder = nn.LSTM(hidden_size, hidden_size, batch_first=True)
              self.fc = nn.Linear(hidden_size, tgt_vocab_size)
          def forward(self, src, tgt):
              src embedded = self.src embedding(src)
              tgt_embedded = self.tgt_embedding(tgt)
              _, (hidden, cell) = self.encoder(src_embedded)
              output, _ = self.decoder(tgt_embedded, (hidden, cell))
              return self.fc(output)
[14]: # Initialize the Seg2Seg model
      model = Seq2SeqLSTM(src_vocab_size, tgt_vocab_size, hidden_size=256).to(device)
      optimizer = torch.optim.Adam(model.parameters(), lr=0.001)
      criterion = nn.CrossEntropyLoss(ignore_index=0) # Use padding index
[15]: # Training loop
      num_epochs = 2
      for epoch in range(num_epochs):
          model.train()
          total loss = 0
          progress_bar = tqdm(enumerate(train_loader), total=len(train_loader),__

desc=f"Epoch {epoch+1}/{num_epochs}")

          for batch_idx, (src, tgt) in progress_bar:
              src, tgt = src.to(device), tgt.to(device)
              optimizer.zero_grad()
              # Forward pass
              output = model(src, tgt[:, :-1])
              loss = criterion(output.reshape(-1, tgt_vocab_size), tgt[:, 1:].
       →reshape(-1))
              loss.backward()
              optimizer.step()
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total_loss += loss.item()
              avg_loss = total_loss / (batch_idx + 1)
              # Update progress bar
              progress_bar.set_postfix({
                  'Loss': f'{avg_loss:.4f}',
                  'Batch': f'{batch_idx+1}/{len(train_loader)}'
              })
          print(f"Epoch {epoch+1}/{num_epochs} completed. Average Loss: {avg_loss:.

4f}")
          # Validation
          model.eval()
          val loss = 0
          with torch.no_grad():
              for src, tgt in test_loader:
                  src, tgt = src.to(device), tgt.to(device)
                  output = model(src, tgt[:, :-1])
                  loss = criterion(output.reshape(-1, tgt_vocab_size), tgt[:, 1:].
       \rightarrowreshape(-1))
                  val_loss += loss.item()
          val_loss /= len(test_loader)
          print(f"Validation Loss: {val_loss:.4f}")
          # Save the best model
          torch.save(model.state_dict(), 'best_translation_model.pth')
          print("Model saved!")
     Epoch 1/2: 100%|
                           | 6381/6381 [6:06:21<00:00, 3.44s/it, Loss=6.4746,
     Batch=6381/6381]
     Epoch 1/2 completed. Average Loss: 6.4746
     Validation Loss: 5.8262
     Model saved!
     Epoch 2/2: 100%|
                           | 6381/6381 [5:21:41<00:00, 3.02s/it, Loss=5.1379,
     Batch=6381/6381]
     Epoch 2/2 completed. Average Loss: 5.1379
     Validation Loss: 5.3763
     Model saved!
[16]: # Inference function for translation
      def translate(model, test_loader, src_idx2word, tgt_idx2word, device,_
       →max_tgt_len):
          model.eval()
```

```
all_translations = []
  all_references = []
  for src, tgt in tqdm(test_loader, desc="Translating"):
      src, tgt = src.to(device), tgt.to(device)
      for i in range(len(src)):
          src_sentence = ' '.join([src_idx2word.get(idx.item(), "") for idx_
→in src[i] if idx.item() != 0])
          tgt_sentence = ' '.join([tgt_idx2word.get(idx.item(), "") for idx_
→in tgt[i] if idx.item() != 0])
          src tensor = torch.tensor([src[i].tolist()], device=device)
          with torch.no grad():
               _, (hidden, cell) = model.encoder(model.
⇔src_embedding(src_tensor))
              tgt_tensor = torch.zeros(1, 1, dtype=torch.long, device=device)
              output_sentence = []
              for _ in range(max_tgt_len):
                  output, (hidden, cell) = model.decoder(model.

→tgt_embedding(tgt_tensor), (hidden, cell))
                  output = model.fc(output)
                  predicted = output.argmax(2).item()
                  if predicted == 0:
                      break
                  output_sentence.append(tgt_idx2word.get(predicted, ""))
                  tgt_tensor = torch.tensor([[predicted]], device=device)
          all_translations.append(' '.join(output_sentence))
          all_references.append(tgt_sentence)
  return all_translations, all_references
```

```
[17]: # Translate and calculate BLEU score
from nltk.translate.bleu_score import corpus_bleu
model.load_state_dict(torch.load('best_translation_model.pth'))
translations, references = translate(model, test_loader, src_idx2word, used_idx2word, device, max_tgt_len=20)
```

C:\Users\samee\AppData\Local\Temp\ipykernel 22028\3140131770.py:3:

FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), which uses the default pickle module implicitly. It is possible to construct malicious pickle data which will execute arbitrary code during

unpickling (See

https://github.com/pytorch/pytorch/blob/main/SECURITY.md#untrusted-models for more details). In a future release, the default value for `weights_only` will be flipped to `True`. This limits the functions that could be executed during

unpickling. Arbitrary objects will no longer be allowed to be loaded via this mode unless they are explicitly allowlisted by the user via `torch.serialization.add_safe_globals`. We recommend you start setting `weights_only=True` for any use case where you don't have full control of the loaded file. Please open an issue on GitHub for any issues related to this experimental feature.

model.load_state_dict(torch.load('best_translation_model.pth'))
Translating: 100%| | 1596/1596 [13:17<00:00, 2.00it/s]</pre>

```
[79]: import nltk
      import os
      from urllib.request import urlretrieve
      # Define the directory for nltk_data
      nltk_data_dir = 'C:/nltk_data'
      if not os.path.exists(nltk_data_dir):
          os.makedirs(nltk_data_dir)
      # Redownload punkt manually if necessary
      punkt_url = 'https://raw.githubusercontent.com/nltk/nltk_data/gh-pages/packages/
       ⇔tokenizers/punkt.zip'
      punkt_path = os.path.join(nltk_data_dir, 'punkt.zip')
      urlretrieve(punkt_url, punkt_path)
      # Extract the punkt package
      import zipfile
      with zipfile.ZipFile(punkt_path, 'r') as zip_ref:
          zip_ref.extractall(nltk_data_dir)
      # Load NLTK data path and tokenizer
      nltk.data.path.append(nltk_data_dir)
      nltk.download('punkt', download_dir=nltk_data_dir)
      # Now perform tokenization
      from nltk.tokenize import word_tokenize
      processed_translations = [word_tokenize(t.lower()) for t in translations]
      processed_references = [[word_tokenize(r.lower())] for r in references]
```

[nltk_data] Downloading package punkt to C:/nltk_data...
[nltk_data] Package punkt is already up-to-date!

```
[80]: # Calculate BLEU score
bleu_score = corpus_bleu(processed_references, processed_translations)
print(f"BLEU Score: {bleu_score:.4f}")
```

BLEU Score: 0.0030

```
[81]: # Print some example translations
      num_examples = 5
      print("\nExample Translations:")
     for i in range(min(num_examples, len(translations))):
         print(f"Source: {references[i]}")
         print(f"Translation: {translations[i]}")
         print(f"Reference: {references[i]}")
          print()
     Example Translations:
     Source:
     Translation:
     Reference:
     Source:
     Translation:
     Reference:
     Source:
     Translation:
     Reference:
     Source:
     Translation:
     563
     Reference:
     Source:
     Translation:
     Reference:
```