## Problem 3

## April 15, 2024

```
[]:
    Patrick Ballou
    ID: 801130521
    ECGR 4106
    Homework 4
    Problem 3
    I I I
[]: '\nPatrick Ballou\nID: 801130521\nECGR 4106\nHomework 4\nProblem 3\n'
[]: import torch
    import torch.nn as nn
    import torch.optim as optim
    from torch import cuda
    from torch.utils.data import DataLoader, Dataset
[]: | #check if GPU is available and set the device accordingly
    #device = 'torch.device("cuda:0" if torch.cuda.is available() else "cpu")'
    device = 'cuda'
    print("Using GPU: ", cuda.get_device_name())
    gpu_info = !nvidia-smi
    gpu_info = '\n'.join(gpu_info)
    if gpu_info.find('failed') >= 0:
     print('Not connected to a GPU')
    else:
     print(gpu_info)
   Using GPU: Quadro T2000
   Mon Apr 15 14:04:41 2024
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   | NVIDIA-SMI 551.86
                                Driver Version: 551.86 CUDA Version:
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   | GPU Name
                         TCC/WDDM | Bus-Id Disp.A | Volatile
   Uncorr. ECC |
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                                     ...aam7r\AcrobatNotificationClient.exe
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```
[]: text = [
         ("J'ai froid", "I am cold"),
         ("Tu es fatigué", "You are tired"),
         ("Il a faim", "He is hungry"),
         ("Elle est heureuse", "She is happy"),
         ("Nous sommes amis", "We are friends"),
         ("Ils sont étudiants", "They are students"),
         ("Le chat dort", "The cat is sleeping"),
         ("Le soleil brille", "The sun is shining"),
         ("Nous aimons la musique", "We love music"),
         ("Elle parle français couramment", "She speaks French fluently"),
         ("Il aime lire des livres", "He enjoys reading books"),
         ("Ils jouent au football chaque week-end", "They play soccer everyu
      ⇔weekend"),
         ("Le film commence à 19 heures", "The movie starts at 7 PM"),
         ("Elle porte une robe rouge", "She wears a red dress"),
         ("Nous cuisinons le dîner ensemble", "We cook dinner together"),
         ("Il conduit une voiture bleue", "He drives a blue car"),
         ("Ils visitent souvent des musées", "They visit museums often"),
         ("Le restaurant sert une délicieuse cuisine", "The restaurant serves_{\sqcup}

delicious food"),
         ("Elle étudie les mathématiques à l'université", "She studies mathematics⊔
      ⇔at university"),
         ("Nous regardons des films le vendredi", "We watch movies on Fridays"),
         ("Il écoute de la musique en faisant du jogging", "He listens to music⊔
      ⇔while jogging"),
         ("Ils voyagent autour du monde", "They travel around the world"),
         ("Le livre est sur la table", "The book is on the table"),
         ("Elle danse avec grâce", "She dances gracefully"),
         ("Nous célébrons les anniversaires avec un gâteau", "We celebrate birthdays⊔
      ⇔with cake"),
```

```
("Il travaille dur tous les jours", "He works hard every day"),
  ("Ils parlent différentes langues", "They speak different languages"),
  ("Les fleurs fleurissent au printemps", "The flowers bloom in spring"),
  ("Elle écrit de la poésie pendant son temps libre", "She writes poetry in ⊔
⇔her free time"),
  ("Nous apprenons quelque chose de nouveau chaque jour", "We learn something,
→new every day"),
  ("Le chien aboie bruyamment", "The dog barks loudly"),
  ("Il chante magnifiquement", "He sings beautifully"),
  ("Ils nagent dans la piscine", "They swim in the pool"),
  ("Les oiseaux gazouillent le matin", "The birds chirp in the morning"),
  ("Elle enseigne l'anglais à l'école", "She teaches English at school"),
  ("Nous prenons le petit déjeuner ensemble", "We eat breakfast together"),
  ("Il peint des paysages", "He paints landscapes"),
  ("Ils rient de la blague", "They laugh at the joke"),
  ("L'horloge tic-tac bruyamment", "The clock ticks loudly"),
  ("Elle court dans le parc", "She runs in the park"),
  ("Nous voyageons en train", "We travel by train"),
  ("Il écrit une lettre", "He writes a letter"),
  ("Ils lisent des livres à la bibliothèque", "They read books at the \sqcup
⇔library"),
  ("Le bébé pleure", "The baby cries"),
  ("Elle étudie dur pour les examens", "She studies hard for exams"),
  ⇔garden"),
  ("Il répare la voiture", "He fixes the car"),
  ("Ils boivent du café le matin", "They drink coffee in the morning"),
  ("Le soleil se couche le soir", "The sun sets in the evening"),
  ("Elle danse à la fête", "She dances at the party"),
  ("Nous jouons de la musique au concert", "We play music at the concert"),
  ("Il cuisine le dîner pour sa famille", "He cooks dinner for his family"),
  ("Ils étudient la grammaire française", "They study French grammar"),
  ("La pluie tombe doucement", "The rain falls gently"),
  ("Elle chante une chanson", "She sings a song"),
  ("Nous regardons un film ensemble", "We watch a movie together"),
  ("Il dort profondément", "He sleeps deeply"),
  ("Ils voyagent à Paris", "They travel to Paris"),
  ("Les enfants jouent dans le parc", "The children play in the park"),
  ("Elle se promène le long de la plage", "She walks along the beach"),
  ("Nous parlons au téléphone", "We talk on the phone"),
  ("Il attend le bus", "He waits for the bus"),
  ("Ils visitent la tour Eiffel", "They visit the Eiffel Tower"),
  ("Les étoiles scintillent la nuit", "The stars twinkle at night"),
  ("Elle rêve de voler", "She dreams of flying"),
  ("Nous travaillons au bureau", "We work in the office"),
  ("Il étudie l'histoire", "He studies history"),
  ("Ils écoutent la radio", "They listen to the radio"),
```

```
("Le vent souffle doucement", "The wind blows gently"),
   ("Elle nage dans l'océan", "She swims in the ocean"),
   ("Nous dansons au mariage", "We dance at the wedding"),
   ("Il gravit la montagne", "He climbs the mountain"),
   ("Ils font de la randonnée dans la forêt", "They hike in the forest"),
   ("Le chat miaule bruyamment", "The cat meows loudly"),
   ("Elle peint un tableau", "She paints a picture"),
   ("Nous construisons un château de sable", "We build a sandcastle"),
   ("Il chante dans le chœur", "He sings in the choir")
]

SOS_token = 0
EOS_token = 1
```

## 1 Problem 3A: French to English without attention

```
[]: | # Vocabulary class to handle mapping between words and numerical indices
     class Vocabulary:
         def __init__(self):
             # Initialize dictionaries for word to index and index to word mappings
             self.word2index = {"<SOS>": SOS_token, "<EOS>": EOS_token}
             self.index2word = {SOS_token: "<SOS>", EOS_token: "<EOS>"}
             self.word_count = {} # Keep track of word frequencies
             self.n words = 2  # Start counting from 2 to account for special tokens
         def add sentence(self, sentence):
             # Add all words in a sentence to the vocabulary
             for word in sentence.split(' '):
                 self.add_word(word)
         def add_word(self, word):
             # Add a word to the vocabulary
             if word not in self.word2index:
                 # Assign a new index to the word and update mappings
                 self.word2index[word] = self.n_words
                 self.index2word[self.n_words] = word
                 self.word_count[word] = 1
                 self.n_words += 1
             else:
                 # Increment word count if the word already exists in the vocabulary
                 self.word count[word] += 1
     # Custom Dataset class for English to French sentences
     class EngFrDataset(Dataset):
         def __init__(self, pairs):
             self.eng_vocab = Vocabulary()
```

```
self.fr_vocab = Vocabulary()
             self.pairs = []
             # Process each English-French pair
             for eng, fr in pairs:
                 self.eng_vocab.add_sentence(eng)
                 self.fr_vocab.add_sentence(fr)
                 self.pairs.append((eng, fr))
             # Separate English and French sentences
             self.eng sentences = [pair[0] for pair in self.pairs]
             self.fr_sentences = [pair[1] for pair in self.pairs]
         # Returns the number of pairs
         def __len__(self):
             return len(self.pairs)
         # Get the sentences by index
         def __getitem__(self, idx):
             input_sentence = self.eng_sentences[idx]
             target_sentence = self.fr_sentences[idx]
             input_indices = [self.eng_vocab.word2index[word] for word in_
      →input_sentence.split()] + [EOS_token]
             target_indices = [self.fr_vocab.word2index[word] for word in_
      →target_sentence.split()] + [EOS_token]
             return torch.tensor(input_indices, dtype=torch.long), torch.
      →tensor(target_indices, dtype=torch.long)
     # Initialize the dataset and DataLoader
     e2f_dataset = EngFrDataset(text)
     dataloader = DataLoader(e2f_dataset, batch_size=1, shuffle=True)
[]: class Encoder(nn.Module):
         """The Encoder part of the seq2seq model."""
         def __init__(self, input_size, hidden_size):
             super(Encoder, self).__init__()
             self.hidden_size = hidden_size
             self.embedding = nn.Embedding(input_size, hidden_size) # Embedding_
      \hookrightarrow layer
             self.gru = nn.GRU(hidden_size, hidden_size) # GRU layer
         def forward(self, input, hidden):
             embedded = self.embedding(input).view(1, 1, -1)
             output, hidden = self.gru(embedded, hidden)
             return output, hidden
```

```
def initHidden(self):
             return torch.zeros(1, 1, self.hidden_size, device=device)
     class Decoder(nn.Module):
         """The Decoder part of the seq2seq model."""
         def __init__(self, hidden_size, output_size):
             super(Decoder, self).__init__()
             self.hidden_size = hidden_size
             self.embedding = nn.Embedding(output_size, hidden_size)
             self.gru = nn.GRU(hidden_size, hidden_size)
             self.out = nn.Linear(hidden size, output size)
             self.softmax = nn.LogSoftmax(dim=1)
         def forward(self, input, hidden):
             embedded = self.embedding(input).view(1, 1, -1)
             output, hidden = self.gru(embedded, hidden)
             output = self.softmax(self.out(output[0]))
             return output, hidden
         def initHidden(self):
             return torch.zeros(1, 1, self.hidden_size, device=device)
[]: def train(input_tensor, target_tensor, encoder, decoder, encoder_optimizer,_
      →decoder optimizer, criterion):
         # Initialize encoder hidden state
         encoder_hidden = encoder.initHidden()
         encoder_optimizer.zero_grad()
         decoder_optimizer.zero_grad()
         input_length = input_tensor.size(0)
         target_length = target_tensor.size(0)
         loss = 0
         # Encoding each character in the input
         for ei in range(input_length):
             encoder_output, encoder_hidden = encoder(input_tensor[ei].unsqueeze(0),_
      ⇔encoder hidden)
         # Decoder's first input is the SOS token
```

decoder\_input = torch.tensor([[SOS\_token]], device=device)

# Decoder starts with encoder's last hidden state

decoder hidden = encoder hidden

# Decoding loop

```
for di in range(target_length):
             decoder_output, decoder_hidden = decoder(decoder_input, decoder_hidden)
             # Choose top1 word from decoder's output
             topv, topi = decoder_output.topk(1)
             decoder_input = topi.squeeze().detach() # Detach from history as input
             # Calculate loss
             loss += criterion(decoder_output, target_tensor[di].unsqueeze(0))
             if decoder_input.item() == EOS_token:
                 break
         # Backpropagation
         loss.backward()
         # Update encoder and decoder
         encoder_optimizer.step()
         decoder_optimizer.step()
         # Return average loss
         return loss.item() / target_length
     def train__loop(encoder, decoder, encoder_optimizer, decoder_optimizer,_u
      ⇔criterion, dataloader, epochs):
         for epoch in range(epochs):
             total_loss = 0
             for input_tensor, target_tensor in dataloader:
                 input_tensor = input_tensor[0].to(device)
                 target_tensor = target_tensor[0].to(device)
                 # Perform a single training step and update loss
                 loss = train(input_tensor, target_tensor, encoder, decoder, u
      ⇔encoder_optimizer, decoder_optimizer, criterion)
                 total_loss += loss
             # Print loss every 5 epochs
             if epoch % 5 == 0:
                 print(f'Epoch {epoch}, Loss: {total_loss / len(dataloader)}')
[]: def evaluate_and_show_examples(encoder, decoder, dataloader, criterion,_
      \rightarrown_examples=10):
         # Switch model to evaluation mode
         encoder.eval()
         decoder.eval()
         total_loss = 0
         correct_predictions = 0
```

```
# No gradient calculation
  with torch.no_grad():
      for i, (input_tensor, target_tensor) in enumerate(dataloader):
          # Move tensors to the correct device
          input_tensor = input_tensor[0].to(device)
          target_tensor = target_tensor[0].to(device)
          encoder_hidden = encoder.initHidden()
          input_length = input_tensor.size(0)
          target_length = target_tensor.size(0)
          loss = 0
          # Encoding step
          for ei in range(input_length):
              encoder_output, encoder_hidden = encoder(input_tensor[ei].
→unsqueeze(0), encoder_hidden)
          # Decoding step
          decoder_input = torch.tensor([[SOS_token]], device=device)
          decoder_hidden = encoder_hidden
          predicted_indices = []
          for di in range(target_length):
              decoder_output, decoder_hidden = decoder(decoder_input,_
→decoder_hidden)
              topv, topi = decoder_output.topk(1)
              predicted_indices.append(topi.item())
              decoder_input = topi.squeeze().detach()
              loss += criterion(decoder_output, target_tensor[di].
unsqueeze(0))
              if decoder_input.item() == EOS_token:
                   break
          # Calculate and print loss and accuracy for the evaluation
          total_loss += loss.item() / target_length
          if predicted_indices == target_tensor.tolist():
              correct_predictions += 1
          # Print some examples
          if i < n examples:</pre>
              predicted_sentence = ' '.join([dataloader.dataset.fr_vocab.
index2word[index] for index in predicted indices if index not in (SOS_token,

→EOS_token)])
```

```
target_sentence = ' '.join([dataloader.dataset.fr_vocab.
               oindex2word[index.item()] for index in target_tensor if index.item() not in in index.item() not in index.item() not in index.item() not in index.item() not in
               →(SOS_token, EOS_token)])
                                                   input_sentence = ' '.join([dataloader.dataset.eng_vocab.
               index2word[index.item()] for index in input_tensor if index.item() not in input_tensor if index.item()
               ⇔(SOS token, EOS token)])
                                                   print(f'Input: {input_sentence}, Target: {target_sentence},
               →Predicted: {predicted_sentence}')
                                # Print overall evaluation results
                                average_loss = total_loss / len(dataloader)
                               accuracy = correct_predictions / len(dataloader)
                               print(f'Evaluation Loss: {average_loss:.4f}, Accuracy: {100*accuracy:.

<p
[]:  # Params
            epochs = 51
            learning_rate = 0.01
            hidden_size = 1028
[]: input_size = len(e2f_dataset.eng_vocab.word2index)
            output size = len(e2f dataset.fr vocab.word2index)
            encoder = Encoder(input_size=input_size, hidden_size=hidden_size).to(device)
            decoder = Decoder(hidden_size=hidden_size, output_size=output_size).to(device)
            # Optimizers
            encoder_optimizer = optim.SGD(encoder.parameters(), lr=learning_rate)
            decoder_optimizer = optim.SGD(decoder.parameters(), lr=learning_rate)
            criterion = nn.NLLLoss()
            train_loop(encoder, decoder, encoder_optimizer, decoder_optimizer, criterion, u
               ⇔dataloader, epochs)
          Epoch 0, Loss: 3.0847776353690977
          Epoch 5, Loss: 2.391458537877896
          Epoch 10, Loss: 1.2405039028171851
          Epoch 15, Loss: 0.20794280146697305
          Epoch 20, Loss: 0.05064842285716246
          Epoch 25, Loss: 0.03017518640154293
          Epoch 30, Loss: 0.021479093558837505
          Epoch 35, Loss: 0.016620042057098543
          Epoch 40, Loss: 0.01352441101706481
          Epoch 45, Loss: 0.011378455714851927
          Epoch 50, Loss: 0.009802569154806633
```

```
Input: Nous travaillons au bureau, Target: We work in the office, Predicted: We
    work in the office
    Input: Le soleil brille, Target: The sun is shining, Predicted: The sun is
    Input: Il peint des paysages, Target: He paints landscapes, Predicted: He paints
    landscapes
    Input: Nous jouons de la musique au concert, Target: We play music at the
    concert, Predicted: We play music at the concert
    Input: Nous parlons au téléphone, Target: We talk on the phone, Predicted: We
    talk on the phone
    Input: Elle écrit de la poésie pendant son temps libre, Target: She writes
    poetry in her free time, Predicted: She writes poetry in her free time
    Input: Il répare la voiture, Target: He fixes the car, Predicted: He fixes the
    car
    Input: Nous regardons des films le vendredi, Target: We watch movies on Fridays,
    Predicted: We watch movies on Fridays
    Input: Il attend le bus, Target: He waits for the bus, Predicted: He waits for
    Input: Elle se promène le long de la plage, Target: She walks along the beach,
    Predicted: She walks along the beach
    Evaluation Loss: 0.0094, Accuracy: 100.00%
[]: torch.save(encoder.state dict(), '../../Models/hw4 3a encoder.pth')
     torch.save(decoder.state_dict(), '../../Models/hw4_3b_decoder.pth')
```

[]: evaluate and show examples (encoder, decoder, dataloader, criterion)

## 2 Problem 3B: French to English with attention

```
[]: # Decoder with attention
     class AttentionDecoder(nn.Module):
         def __init__(self, hidden_size, output_size, max_length=16, dropout_p=0.1):
             super(AttentionDecoder, self).__init__()
             self.hidden_size = hidden_size
             self.output_size = output_size
             self.dropout_p = dropout_p
             self.max_length = max_length
             self.embedding = nn.Embedding(self.output_size, self.hidden_size) #__
      →Embedding layer
             self.attn = nn.Linear(self.hidden_size * 2, self.max_length) #__
      → Attention layer
             self.attn_combine = nn.Linear(self.hidden_size * 2, self.hidden_size) #_
      → Combining layer
             self.dropout = nn.Dropout(self.dropout_p)
             self.gru = nn.GRU(self.hidden_size, self.hidden_size)
             self.out = nn.Linear(self.hidden_size, output_size)
```

```
def forward(self, input, hidden, encoder_outputs):
             embedded = self.embedding(input).view(1, 1, -1)
             embedded = self.dropout(embedded)
             # Calculating attention weights
             attn_weights = torch.softmax(self.attn(torch.cat((embedded[0],__
      \hookrightarrowhidden[0]), 1)), dim=1)
             attn_applied = torch.bmm(attn_weights.unsqueeze(0), encoder_outputs.
      unsqueeze(0))
             # Combining embedded input with attention output
             output = torch.cat((embedded[0], attn_applied[0]), 1)
             output = self.attn_combine(output).unsqueeze(0)
             output = torch.relu(output)
             output, hidden = self.gru(output, hidden)
             output = torch.log_softmax(self.out(output[0]), dim=1)
             return output, hidden, attn_weights
         def initHidden(self):
             return torch.zeros(1, 1, self.hidden_size, device=device)
[]: def train(input_tensor, target_tensor, encoder, decoder, encoder_optimizer,__

→decoder_optimizer, criterion, max_length=16):
         # Initialize encoder hidden state
         encoder_hidden = encoder.initHidden()
         encoder_optimizer.zero_grad()
         decoder_optimizer.zero_grad()
         input_length = input_tensor.size(0)
         target_length = target_tensor.size(0)
         loss = 0
         # Encoding each character in the input
         encoder_outputs = torch.zeros(max_length, encoder.hidden_size,_
      →device=device)
```

encoder\_output, encoder\_hidden = encoder(input\_tensor[ei].unsqueeze(0),\_

for ei in range(input\_length):

encoder\_outputs[ei] = encoder\_output[0, 0]

decoder\_input = torch.tensor([[SOS\_token]], device=device)

# Decoder's first input is the SOS token

⇔encoder\_hidden)

```
# Decoder starts with encoder's last hidden state
         decoder_hidden = encoder_hidden
        # Decoding loop with attention
         for di in range(target_length):
             decoder_output, decoder_hidden, decoder_attention = decoder(
                 decoder_input, decoder_hidden, encoder_outputs)
             topv, topi = decoder_output.topk(1)
             decoder_input = topi.squeeze().detach() # Detach from history as input
             loss += criterion(decoder_output, target_tensor[di].unsqueeze(0))
             if decoder_input.item() == EOS_token:
                 break
         # Backpropagation
         loss.backward()
         # Update encoder and decoder
         encoder_optimizer.step()
         decoder_optimizer.step()
         # Return average loss
         return loss.item() / target_length
     def train__loop(encoder, decoder, encoder_optimizer, decoder_optimizer,_u
      ⇔criterion, dataloader, epochs):
         for epoch in range(epochs):
             total_loss = 0
             for input_tensor, target_tensor in dataloader:
                 input_tensor = input_tensor[0].to(device)
                 target_tensor = target_tensor[0].to(device)
                 # Perform a single training step and update loss
                 loss = train(input_tensor, target_tensor, encoder, decoder,__
      ⇔encoder_optimizer, decoder_optimizer, criterion)
                 total_loss += loss
             # Print loss every 5 epochs
             if epoch % 5 == 0:
                 print(f'Epoch {epoch}, Loss: {total_loss / len(dataloader)}')
[]: def evaluate_and_show_examples(encoder, decoder, dataloader, criterion,_
      on_examples=10, max_length=16):
         # Switch model to evaluation mode
         encoder.eval()
         decoder.eval()
```

```
total_loss = 0
  correct_predictions = 0
  # No gradient calculation
  with torch.no_grad():
      for i, (input_tensor, target_tensor) in enumerate(dataloader):
          # Move tensors to the correct device
          input_tensor = input_tensor[0].to(device)
          target_tensor = target_tensor[0].to(device)
          encoder_hidden = encoder.initHidden()
          input_length = input_tensor.size(0)
          target_length = target_tensor.size(0)
          loss = 0
          # Encoding each character in the input
          encoder_outputs = torch.zeros(max_length, encoder.hidden_size,__

device=device)
          for ei in range(input_length):
              encoder_output, encoder_hidden = encoder(input_tensor[ei].
→unsqueeze(0), encoder_hidden)
              encoder_outputs[ei] = encoder_output[0, 0]
          # Decoding step
          decoder_input = torch.tensor([[SOS_token]], device=device)
          decoder_hidden = encoder_hidden
          predicted_indices = []
          for di in range(target_length):
              decoder_output, decoder_hidden, decoder_attention =_
decoder(decoder_input, decoder_hidden, encoder_outputs)
              topv, topi = decoder_output.topk(1)
              predicted_indices.append(topi.item())
              decoder_input = topi.squeeze().detach()
              loss += criterion(decoder_output, target_tensor[di].

unsqueeze(0))
              if decoder_input.item() == EOS_token:
                  break
          # Calculate and print loss and accuracy for the evaluation
          total_loss += loss.item() / target_length
          if predicted_indices == target_tensor.tolist():
```

```
correct_predictions += 1
                    # Print some examples
                    if i < n_examples:</pre>
                         predicted_sentence = ' '.join([dataloader.dataset.fr_vocab.
       →EOS token)])
                         target_sentence = ' '.join([dataloader.dataset.fr_vocab.
       index2word[index.item()] for index in target_tensor if index.item() not in in index.item() not in index.item()
       ⇔(SOS_token, EOS_token)])
                         input_sentence = ' '.join([dataloader.dataset.eng_vocab.
       index2word[index.item()] for index in input_tensor if index.item() not in input_tensor if index.item()
       →(SOS_token, EOS_token)])
                         print(f'Input: {input_sentence}, Target: {target_sentence},__
       →Predicted: {predicted_sentence}')
               # Print overall evaluation results
               average_loss = total_loss / len(dataloader)
               accuracy = correct_predictions / len(dataloader)
               print(f'Evaluation Loss: {average_loss:.4f}, Accuracy: {100*accuracy:.

<pre
[ ]:  # Params
      epochs = 51
      learning_rate = 0.01
      hidden_size = 1028
[]: input_size = len(e2f_dataset.eng_vocab.word2index)
      output_size = len(e2f_dataset.fr_vocab.word2index)
      encoder = Encoder(input_size=input_size, hidden_size=hidden_size).to(device)
      decoder = AttentionDecoder(hidden_size=hidden_size, output_size=output_size).
       →to(device)
      # Optimizers
      encoder_optimizer = optim.SGD(encoder.parameters(), lr=learning_rate)
      decoder_optimizer = optim.SGD(decoder.parameters(), lr=learning_rate)
      criterion = nn.NLLLoss()
      train_loop(encoder, decoder_optimizer, decoder_optimizer, criterion, u
       ⇔dataloader, epochs)
     Epoch 0, Loss: 3.0278098774310176
     Epoch 5, Loss: 2.410202463983974
     Epoch 10, Loss: 1.2605623107820512
     Epoch 15, Loss: 0.30540076114960707
```

```
Epoch 20, Loss: 0.06631536839498772
    Epoch 25, Loss: 0.033590564190945145
    Epoch 30, Loss: 0.021999660827998185
    Epoch 35, Loss: 0.01609092903627002
    Epoch 40, Loss: 0.012645105581302092
    Epoch 45, Loss: 0.01031990974096739
    Epoch 50, Loss: 0.008761231182459866
[]: evaluate_and_show_examples(encoder, decoder, dataloader, criterion)
    Input: Ils voyagent à Paris, Target: They travel to Paris, Predicted: They
    travel to Paris
    Input: Nous aimons la musique, Target: We love music, Predicted: We love music
    Input: Tu es fatigué, Target: You are tired, Predicted: You are tired
    Input: Il cuisine le dîner pour sa famille, Target: He cooks dinner for his
    family, Predicted: He cooks dinner for his family
    Input: Il conduit une voiture bleue, Target: He drives a blue car, Predicted: He
    drives a blue car
    Input: Il chante magnifiquement, Target: He sings beautifully, Predicted: He
    sings beautifully
    Input: Nous jouons de la musique au concert, Target: We play music at the
    concert, Predicted: We play music at the concert
    Input: Ils boivent du café le matin, Target: They drink coffee in the morning,
    Predicted: They drink coffee in the morning
    Input: Le chat miaule bruyamment, Target: The cat meows loudly, Predicted: The
    cat meows loudly
    Input: Il aime lire des livres, Target: He enjoys reading books, Predicted: He
    enjoys reading books
    Evaluation Loss: 0.0082, Accuracy: 100.00%
[]: torch.save(encoder.state dict(), '../../Models/hw4 3b encoder.pth')
     torch.save(decoder.state_dict(), '../../Models/hw4_3b_decoder.pth')
```