Problem 3

April 15, 2024

```
[]:
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    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 14:18:26 2024
   +----
    _____
   | NVIDIA-SMI 551.86
                                 Driver Version: 551.86 CUDA Version:
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   ----+
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                          TCC/WDDM | Bus-Id Disp.A | Volatile
   Uncorr. ECC |
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     0
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                                       ...les\Microsoft OneDrive\OneDrive.exe
    21636
                                   C+G
    N/A
[]: text = [
         ("J'ai froid", "I am cold"),
         ("Tu es fatigué", "You are tired"),
```

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("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_{\sqcup}
→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"),
  ⇔library"),
  ("Le bébé pleure", "The baby cries"),
  ("Elle étudie dur pour les examens", "She studies hard for exams"),
  ("Nous plantons des fleurs dans le jardin", "We plant flowers in the
⇔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 seg2seg 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)
 →decoder optimizer, criterion):
    # Initialize encoder hidden state
    encoder hidden = encoder.initHidden()
   encoder_optimizer.zero_grad()
```

```
[]: def train(input_tensor, target_tensor, encoder, decoder, encoder_optimizer,_
         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,_
      ⇔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,__
      \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.

→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:.

<pr
[]:  # 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: 3.021917926491081
          Epoch 5, Loss: 2.431328812025916
          Epoch 10, Loss: 1.1711633006621083
          Epoch 15, Loss: 0.1914339971532222
          Epoch 20, Loss: 0.04979745171544871
          Epoch 25, Loss: 0.02994382827736819
          Epoch 30, Loss: 0.021387013262880766
          Epoch 35, Loss: 0.016593719381739976
          Epoch 40, Loss: 0.013516508670259976
          Epoch 45, Loss: 0.011382935190632085
          Epoch 50, Loss: 0.009814835214110789
```

```
[]: evaluate and show examples (encoder, decoder, dataloader, criterion)
    Input: Ils lisent des livres à la bibliothèque, Target: They read books at the
    library, Predicted: They read books at the library
    Input: Le soleil se couche le soir, Target: The sun sets in the evening,
    Predicted: The sun sets in the evening
    Input: Nous prenons le petit déjeuner ensemble, Target: We eat breakfast
    together, Predicted: We eat breakfast together
    Input: Il répare la voiture, Target: He fixes the car, Predicted: He fixes the
    car
    Input: Elle peint un tableau, Target: She paints a picture, Predicted: She
    paints a picture
    Input: Les fleurs fleurissent au printemps, Target: The flowers bloom in spring,
    Predicted: The flowers bloom in spring
    Input: Nous regardons un film ensemble, Target: We watch a movie together,
    Predicted: We watch a movie together
    Input: Ils nagent dans la piscine, Target: They swim in the pool, Predicted:
    They swim in the pool
    Input: Ils écoutent la radio, Target: They listen to the radio, Predicted: They
    listen to the radio
    Input: Il aime lire des livres, Target: He enjoys reading books, Predicted: He
    enjoys reading books
    Evaluation Loss: 0.0094, Accuracy: 100.00%
[]: torch.save(encoder.state dict(), '../../Models/hw4 3a encoder.pth')
     torch.save(decoder.state_dict(), '../../Models/hw4_3a_decoder.pth')
```

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:.
      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: 3.074753296397392
    Epoch 5, Loss: 2.427136123777542
    Epoch 10, Loss: 1.1876769432405228
    Epoch 15, Loss: 0.2797134742523206
```

```
Epoch 20, Loss: 0.06737830363485868
    Epoch 25, Loss: 0.033576714494896284
    Epoch 30, Loss: 0.022193639288064893
    Epoch 35, Loss: 0.016341955855663868
    Epoch 40, Loss: 0.01290227331695692
    Epoch 45, Loss: 0.01056426606108201
    Epoch 50, Loss: 0.008934163938270132
[]: evaluate_and_show_examples(encoder, decoder, dataloader, criterion)
    Input: Il gravit la montagne, Target: He climbs the mountain, Predicted: He
    climbs the mountain
    Input: Elle parle français couramment, Target: She speaks French fluently,
    Predicted: She speaks French fluently
    Input: Elle danse à la fête, Target: She dances at the party, Predicted: She
    dances at the party
    Input: Ils visitent la tour Eiffel, Target: They visit the Eiffel Tower,
    Predicted: They visit the Eiffel Tower
    Input: Elle enseigne l'anglais à l'école, Target: She teaches English at school,
    Predicted: She teaches English at school
    Input: Il cuisine le dîner pour sa famille, Target: He cooks dinner for his
    family, Predicted: He cooks dinner for his family
    Input: Ils voyagent à Paris, Target: They travel to Paris, Predicted: They
    travel to Paris
    Input: Ils écoutent la radio, Target: They listen to the radio, Predicted: They
    listen to the radio
    Input: Nous parlons au téléphone, Target: We talk on the phone, Predicted: We
    talk on the phone
    Input: Les étoiles scintillent la nuit, Target: The stars twinkle at night,
    Predicted: The stars twinkle at night
    Evaluation Loss: 0.0083, Accuracy: 100.00%
[]: torch.save(encoder.state dict(), '../../Models/hw4 3b encoder.pth')
     torch.save(decoder.state_dict(), '../../Models/hw4_3b_decoder.pth')
```