### ▼ Imports

```
from io import open
import unicodedata
import string
import re
import random
import pandas as pd
import torch
import torch.nn as nn
from torch import optim
import torch.nn.functional as F
import numpy as np
import operator
from gensim.models import KeyedVectors
%matplotlib inline
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
device
    device(type='cuda')
```

# Loading data files

```
!wget https://download.pytorch.org/tutorial/data.zip
!unzip data.zip
    --2021-07-11 11:08:04-- <a href="https://download.pytorch.org/tutorial/data.zip">https://download.pytorch.org/tutorial/data.zip</a>
    Resolving download.pytorch.org (download.pytorch.org)... 99.86.37.116, 99.86.
    Connecting to download.pytorch.org (download.pytorch.org)|99.86.37.116|:443...
    HTTP request sent, awaiting response... 200 OK
    Length: 2882130 (2.7M) [application/zip]
    Saving to: 'data.zip'
                          100%[======]
                                                         2.75M --.-KB/s
                                                                            in 0.05s
    data.zip
    2021-07-11 11:08:04 (53.5 MB/s) - 'data.zip' saved [2882130/2882130]
    Archive: data.zip
        creating: data/
       inflating: data/eng-fra.txt
        creating: data/names/
       inflating: data/names/Arabic.txt
       inflating: data/names/Chinese.txt
       inflating: data/names/Czech.txt
       inflating: data/names/Dutch.txt
```

```
inflating: data/names/English.txt
inflating: data/names/French.txt
inflating: data/names/German.txt
inflating: data/names/Irish.txt
inflating: data/names/Italian.txt
inflating: data/names/Japanese.txt
inflating: data/names/Korean.txt
inflating: data/names/Polish.txt
inflating: data/names/Portuguese.txt
inflating: data/names/Russian.txt
inflating: data/names/Scottish.txt
inflating: data/names/Spanish.txt
inflating: data/names/Spanish.txt
inflating: data/names/Vietnamese.txt
```

### Glove Vectors

```
# !unzip glove.6B.zip
# !wget https://nlp.stanford.edu/data/glove.42B.300d.zip
# !unzip glove.42B.300d.zip
# !rm -rf glove.42B.300d.zip
# !wget https://nlp.stanford.edu/data/glove.twitter.27B.zip
# !unzip glove.twitter.27B.zip
# !rm -rf glove.twitter.27B.zip
from google.colab import drive
drive.mount('/content/drive')
    Mounted at /content/drive
glove_path = f'/content/drive/MyDrive/glove'
# glove_words = []
\# idx = 0
# glove_word2id = {}
# glove_vectors = []
# with open(f'{glove_path}/glove.6B.100d.txt', 'rb') as f:
      for l in f:
#
#
          line = l.decode().split()
          word = line[0]
#
#
          glove_words.append(word)
#
          glove_word2id[word] = idx
#
          idx += 1
#
          word_vector = np.array(line[1:]).astype(np.float)
          glove_vectors.append(word_vector)
```

# !wget https://nlp.stanford.edu/data/glove.6B.zip

```
# glove_words = []
\# idx = 0
# glove word2id = {}
# glove_vectors = []
# with open(f'{glove path}/glove.42B.300d.txt', 'rb') as f:
      for l in f:
#
          line = l.decode().split()
          word = line[0]
#
#
          glove words.append(word)
          glove_word2id[word] = idx
#
          idx += 1
#
          word_vector = np.array(line[1:]).astype(np.float)
#
#
          glove_vectors.append(word_vector)
glove_words = []
idx = 0
glove word2id = {}
glove_vectors = []
with open(f'{glove_path}/glove.twitter.27B.200d.txt', 'rb') as f:
    for l in f:
        line = l.decode().split()
        word = line[0]
        glove words.append(word)
        glove_word2id[word] = idx
        idx += 1
        word_vector = np.array(line[1:]).astype(np.float)
        glove_vectors.append(word_vector)
glove_word2vec = {word: glove_vectors[glove_word2id[word]] for word in glove_words
# n_glove_vectors = len(glove_words)
# dim_glove_vectors = 100
# class InputLang:
      def __init__(self, name):
#
#
          self.name = name
          self.word2index = { k : v for k , v in sorted(glove_word2id.items(), key:
#
          self.word2count = { word : 1 for word in glove_words }
#
          self.index2word = { i : word for word, i in glove_word2id.items() }
#
#
          self.n_words = len(glove_words)
#
      def addSentence(self, sentence):
#
          for word in sentence.split(' '):
#
              self.addWord(word)
      def addWord(self, word):
#
#
          if word not in self.word2index:
              print(word)
```

```
def unicodeToAscii(s):
    return ''.join(
        c for c in unicodedata.normalize('NFD', s)
        if unicodedata.category(c) != 'Mn'
# Lowercase, trim, and remove non-letter characters
def normalizeString(s):
    s = unicodeToAscii(s.lower().strip())
    s = re.sub(r"([.!?])", r" \1", s)
    s = re.sub(r"[^a-zA-Z.!?]+", r" ", s)
    return s
def readLangs():
    print("Reading lines...")
    # Read the file and split into lines
    lines = open('data/eng-fra.txt', encoding='utf-8').\
        read().strip().split('\n')
    # Split every line into pairs and normalize
    pairs = [[normalizeString(s) for s in l.split('\t')] for l in lines]
    pairs = [list((p)) for p in pairs]
    input_lang = InputLang('eng')
    output lang = OutputLang('fra')
    return input_lang, output_lang, pairs
MAX LENGTH = 10
eng_prefixes = (
    "i am ", "i m ",
    "he is", "he s "
    "she is", "she s ",
              , "you re ",
    "you are",
    "we are", "we re ",
    "they are", "they re "
)
def filterPair(p):
    return len(p[0].split(' ')) < MAX_LENGTH and \</pre>
        len(p[1].split(' ')) < MAX_LENGTH and \</pre>
        p[0].startswith(eng_prefixes)
def filterPairs(pairs):
    return [pair for pair in pairs if filterPair(pair)]
def prepareData():
    input_lang, output_lang, pairs = readLangs()
    print("Read %s sentence pairs" % len(pairs))
    pairs = filterPairs(pairs)
    print("Trimmed to %s sentence pairs" % len(pairs))
    print("Counting words...")
```

```
for pair in pairs:
        input_lang.addSentence(pair[0])
        output_lang.addSentence(pair[1])
    print("Counted words:")
    print(input_lang.name, input_lang.n_words)
    print(output_lang.name, output_lang.n_words)
    return input_lang, output_lang, pairs
input_lang, output_lang, pairs = prepareData()
print(random.choice(pairs))
    Reading lines...
    Read 135842 sentence pairs
    Trimmed to 10599 sentence pairs
    Counting words...
    Counted words:
    eng 2803
    fra 4345
    ['i am yours and you are mine .', 'je suis votre et vous etes mien .']
input_vocab_size = input_lang.n_words
weights matrix = np.zeros((input vocab size, 200))
words found = 0
for i, word in enumerate(input lang.word2index):
    try:
        weights_matrix[i] = glove_word2vec[word]
        words_found += 1
    except KeyError:
        weights_matrix[i] = np.random.normal(scale=0.6, size=(200, ))
```

### French Word Embeddings

```
# !wget https://s3.us-east-2.amazonaws.com/embeddings.net/embeddings/frWac_non_lem
# !cp frWac_non_lem_no_postag_no_phrase_200_skip_cut100.bin /content/drive/MyDrive,
     --2021-07-11 11:08:52-- <a href="https://s3.us-east-2.amazonaws.com/embeddings.net/em">https://s3.us-east-2.amazonaws.com/embeddings.net/em</a>
    Resolving s3.us-east-2.amazonaws.com (s3.us-east-2.amazonaws.com)... 52.219.8
     Connecting to s3.us-east-2.amazonaws.com (s3.us-east-2.amazonaws.com)|52.219.
    HTTP request sent, awaiting response... 200 OK
    Length: 126052447 (120M) [application/octet-stream]
     Saving to: 'frWac_non_lem_no_postag_no_phrase_200_skip_cut100.bin'
     frWac_non_lem_no_po 100%[===========] 120.21M 35.0MB/s
                                                                             in 3.4s
     2021-07-11 11:08:56 (35.0 MB/s) - 'frWac_non_lem_no_postag_no_phrase_200_skip
fr_w2v_path = f'/content/drive/MyDrive/glove/frWac_non_lem_no_postag_no_phrase_200
fr_w2v = KeyedVectors.load_word2vec_format(fr_w2v_path, binary=True)
output_vocab_size = output_lang.n_words
out weights matrix = np.zeros((output vocab size, 200))
```

```
7/11/2021
                                  END2 Assign 10 S Eng+FreEmb.ipynb - Colaboratory
   for i, word in enumerate(output_lang.word2index):
            out_weights_matrix[i] = fr_w2v[word]
        except KeyError:
            out_weights_matrix[i] = np.random.normal(scale=0.6, size=(200, ))
```

# The Seq2Seq Model

#### ▼ The Encoder

```
class EncoderRNN(nn.Module):
    def __init__(self, input_size, hidden_size):
        super(EncoderRNN, self).__init__()
        self.hidden_size = hidden_size
        self.embedding = nn.Embedding(input_size, hidden_size)
        self.embedding.weight.data.copy_(torch.from_numpy(weights_matrix))
        # Make this vector non-trainable
        for param in self.embedding.parameters():
            param.requires_grad = False
        self.gru = nn.GRU(hidden_size, hidden_size)
   def forward(self, input, hidden):
        embedded = self.embedding(input).view(1, 1, -1)
        output = embedded
        output, hidden = self.gru(output, hidden)
        return output, hidden
   def initHidden(self):
        return torch.zeros(1, 1, self.hidden_size, device=device)
```

#### Attention Decoder

```
class AttnDecoderRNN(nn.Module):
   def __init__(self, hidden_size, output_size, dropout_p=0.1, max_length=MAX_LEN
        super(AttnDecoderRNN, 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)
        self.embedding.weight.data.copy_(torch.from_numpy(out_weights_matrix))
        # Make this vector non-trainable
        for param in self.embedding.parameters():
            param.requires grad = False
```

```
self.attn = nn.Linear(self.hidden_size * 2, self.max_length)
    self.attn_combine = nn.Linear(self.hidden_size * 2, self.hidden_size)
    self.dropout = nn.Dropout(self.dropout_p)
    self.gru = nn.GRU(self.hidden size, self.hidden size)
    self.out = nn.Linear(self.hidden_size, self.output_size)
def forward(self, input, hidden, encoder_outputs):
    embedded = self.embedding(input).view(1, 1, -1)
    embedded = self.dropout(embedded)
    attn weights = F.softmax(
        self.attn(torch.cat((embedded[0], hidden[0]), 1)), dim=1)
    attn_applied = torch.bmm(attn_weights.unsqueeze(0),
                             encoder_outputs.unsqueeze(0))
    output = torch.cat((embedded[0], attn_applied[0]), 1)
    output = self.attn_combine(output).unsqueeze(0)
    output = F.relu(output)
    output, hidden = self.gru(output, hidden)
    output = F.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)
```

# Training

### **Preparing Training Data**

```
def indexesFromSentence(lang, sentence):
    return [lang.word2index[word] for word in sentence.split(' ')]

def tensorFromSentence(lang, sentence):
    indexes = indexesFromSentence(lang, sentence)
    indexes.append(EOS_token)
    return torch.tensor(indexes, dtype=torch.long, device=device).view(-1, 1)

def tensorsFromPair(pair):
    input_tensor = tensorFromSentence(input_lang, pair[0])
    target_tensor = tensorFromSentence(output_lang, pair[1])
    return (input_tensor, target_tensor)
```

## Training the Model

```
teacher_forcing_ratio = 0.5
def train(input_tensor, target_tensor, encoder, decoder, encoder_optimizer, decode
    encoder_hidden = encoder.initHidden()
    encoder_optimizer.zero_grad()
    decoder_optimizer.zero_grad()
    input length = input tensor.size(0)
    target length = target tensor.size(0)
    encoder_outputs = torch.zeros(max_length, encoder.hidden_size, device=device)
    loss = 0
    for ei in range(input_length):
        encoder_output, encoder_hidden = encoder(
            input_tensor[ei], encoder_hidden)
        encoder_outputs[ei] = encoder_output[0, 0]
    decoder_input = torch.tensor([[SOS_token]], device=device)
    decoder hidden = encoder hidden
    use_teacher_forcing = True if random.random() < teacher_forcing_ratio else Fal:</pre>
    if use_teacher_forcing:
        # Teacher forcing: Feed the target as the next input
        for di in range(target length):
            decoder_output, decoder_hidden, decoder_attention = decoder(
                decoder_input, decoder_hidden, encoder_outputs)
            loss += criterion(decoder_output, target_tensor[di])
            decoder_input = target_tensor[di] # Teacher forcing
    else:
        # Without teacher forcing: use its own predictions as the next input
        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 inpu
            loss += criterion(decoder_output, target_tensor[di])
            if decoder_input.item() == EOS_token:
                break
    loss.backward()
    encoder_optimizer.step()
    decoder_optimizer.step()
    return loss.item() / target_length
```

```
import time
import math
def asMinutes(s):
   m = math.floor(s / 60)
    s -= m * 60
    return '%dm %ds' % (m, s)
def timeSince(since, percent):
   now = time.time()
    s = now - since
    es = s / (percent)
    rs = es - s
    return '%s (- %s)' % (asMinutes(s), asMinutes(rs))
def trainIters(encoder, decoder, n_iters, print_every=1000, plot_every=100, learni
    start = time.time()
    plot_losses = []
    print_loss_total = 0 # Reset every print_every
    plot_loss_total = 0 # Reset every plot_every
    encoder_optimizer = optim.SGD(encoder.parameters(), lr=learning_rate)
    decoder_optimizer = optim.SGD(decoder.parameters(), lr=learning_rate)
    training_pairs = [tensorsFromPair(random.choice(pairs))
                      for i in range(n_iters)]
    criterion = nn.NLLLoss()
    for iter in range(1, n_iters + 1):
        training_pair = training_pairs[iter - 1]
        input_tensor = training_pair[0]
        target_tensor = training_pair[1]
        loss = train(input_tensor, target_tensor, encoder,
                     decoder, encoder_optimizer, decoder_optimizer, criterion)
        print_loss_total += loss
        plot_loss_total += loss
        if iter % print_every == 0:
            print_loss_avg = print_loss_total / print_every
            print loss total = 0
            print('%s (%d %d%%) %.4f' % (timeSince(start, iter / n_iters),
                                         iter, iter / n_iters * 100, print_loss_ave
        if iter % plot_every == 0:
            plot_loss_avg = plot_loss_total / plot_every
            plot_losses.append(plot_loss_avg)
            plot_loss_total = 0
    showPlot(plot_losses)
```

# Plotting results

Plotting is done with matplotlib, using the array of loss values plot\_losses saved while training

```
import matplotlib.pyplot as plt
plt.switch_backend('agg')
import matplotlib.ticker as ticker
import numpy as np

def showPlot(points):
    plt.figure()
    fig, ax = plt.subplots()
    # this locator puts ticks at regular intervals
    loc = ticker.MultipleLocator(base=0.2)
    ax.yaxis.set_major_locator(loc)
    plt.plot(points)
```

### Evaluation

```
def evaluate(encoder, decoder, sentence, max_length=MAX_LENGTH):
    with torch.no_grad():
        input_tensor = tensorFromSentence(input_lang, sentence)
        input_length = input_tensor.size()[0]
        encoder_hidden = encoder.initHidden()
        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],
                                                      encoder_hidden)
            encoder_outputs[ei] += encoder_output[0, 0]
        decoder_input = torch.tensor([[SOS_token]], device=device) # SOS
        decoder_hidden = encoder_hidden
        decoded_words = []
        decoder_attentions = torch.zeros(max_length, max_length)
        for di in range(max_length):
            decoder_output, decoder_hidden, decoder_attention = decoder(
                decoder_input, decoder_hidden, encoder_outputs)
            decoder_attentions[di] = decoder_attention.data
            topv, topi = decoder_output.data.topk(1)
            if topi.item() == EOS_token:
                decoded_words.append('<E0S>')
                break
            else:
                decoded_words.append(output_lang.index2word[topi.item()])
```

```
decoder_input = topi.squeeze().detach()

return decoded_words, decoder_attentions[:di + 1]

def evaluateRandomly(encoder, decoder, n=10):
    for i in range(n):
        pair = random.choice(pairs)
        print('>', pair[0])
        print('=', pair[1])
        output_words, attentions = evaluate(encoder, decoder, pair[0])
        output_sentence = ' '.join(output_words)
        print('<', output_sentence)
        print('')</pre>
```

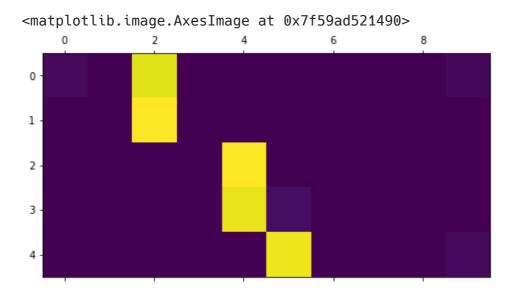
## Training and Evaluating

```
enc_hidden_size = 200
dec hidden size = 200
encoder1 = EncoderRNN(input_lang.n_words, enc_hidden_size).to(device)
attn_decoder1 = AttnDecoderRNN(dec_hidden_size, output_lang.n_words, dropout_p=0.1
trainIters(encoder1, attn_decoder1, 75000, print_every=5000)
    1m 6s (- 15m 34s) (5000 6%) 3.7662
    2m 10s (- 14m 6s) (10000 13%) 3.1396
    3m 14s (- 12m 56s) (15000 20%) 2.8985
    4m 19s (- 11m 53s) (20000 26%) 2.6651
    5m 24s (- 10m 48s) (25000 33%) 2.4912
    6m 28s (- 9m 43s) (30000 40%) 2.3679
    7m 34s (- 8m 38s) (35000 46%) 2.2003
    8m 38s (- 7m 33s) (40000 53%) 2.0445
    9m 42s (- 6m 28s) (45000 60%) 1.9527
    10m 47s (- 5m 23s) (50000 66%) 1.8294
    11m 51s (- 4m 18s) (55000 73%) 1.7395
    12m 56s (- 3m 14s) (60000 80%) 1.6617
    14m Os (- 2m 9s) (65000 86%) 1.5625
    15m 5s (- 1m 4s) (70000 93%) 1.4985
    16m 10s (- 0m 0s) (75000 100%) 1.4342
evaluateRandomly(encoder1, attn_decoder1)
    > i m not easily impressed .
    = je ne suis pas facilement impressionne .
    < je ne suis pas facilement impressionne . <EOS>
    > he is no friend of mine .
    = ce n est pas un ami a moi .
    < ce n est pas un qu . <EOS>
    > i m sorry i broke my promise .
    = je suis desole d avoir rompu ma promesse .
    < je suis desole d avoir rompu ma promesse . <EOS>
```

```
> i m not tall .
= je ne suis pas grand .
< je ne suis pas aussi . <EOS>
> she is in a green dress .
= elle porte une robe verte .
< elle est des robe de . <EOS>
> i m a salesperson .
= je suis un vendeur .
< je suis musulman . <EOS>
> you re careless .
= tu es negligent .
< vous etes idiot . <EOS>
> he is reading .
= il est en train de lire .
< il est en train de lire . <EOS>
> we re going to the movies .
= nous nous rendons au cinema .
< nous allons au la . . <EOS>
> he s away on vacation .
= il est absent en conges .
< il est en train . <EOS>
```

### Visualizing Attention

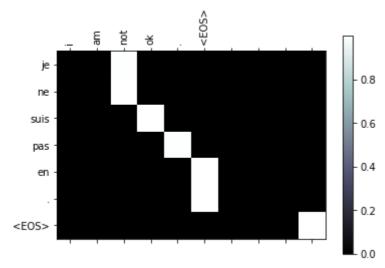
```
%matplotlib inline
output_words, attentions = evaluate(
    encoder1, attn_decoder1, "i am sorry .")
plt.matshow(attentions.numpy())
```



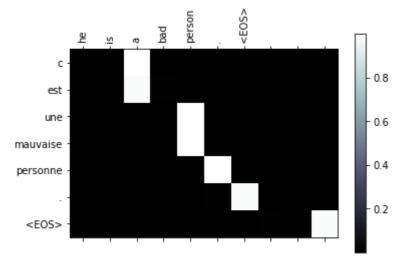
For a better viewing experience we will do the extra work of adding axes and labels:

```
def showAttention(input_sentence, output_words, attentions):
    # Set up figure with colorbar
    fig = plt.figure()
    ax = fig.add subplot(111)
    cax = ax.matshow(attentions.numpy(), cmap='bone')
    fig.colorbar(cax)
    # Set up axes
    ax.set_xticklabels([''] + input_sentence.split(' ') +
                       ['<E0S>'], rotation=90)
    ax.set_yticklabels([''] + output_words)
    # Show label at every tick
    ax.xaxis.set_major_locator(ticker.MultipleLocator(1))
    ax.yaxis.set_major_locator(ticker.MultipleLocator(1))
    plt.show()
def evaluateAndShowAttention(input sentence):
    output_words, attentions = evaluate(
        encoder1, attn_decoder1, input_sentence)
    print('input =', input_sentence)
    print('output =', ' '.join(output_words))
    showAttention(input_sentence, output_words, attentions)
evaluateAndShowAttention("i am not ok .")
evaluateAndShowAttention("he is a bad person .")
evaluateAndShowAttention("she is fast .")
evaluateAndShowAttention("we are going tomorrow .")
```

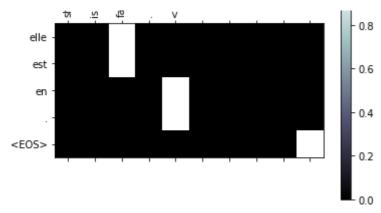
input = i am not ok .
output = je ne suis pas en . <EOS>



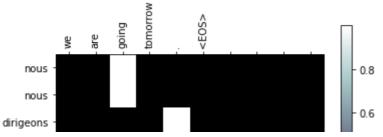
input = he is a bad person .
output = c est une mauvaise personne . <EOS>



input = she is fast .
output = elle est en . <EOS>



input = we are going tomorrow .
output = nous nous dirigeons . . <EOS>



X