Language Translation

In this project, you're going to take a peek into the realm of neural network machine translation. You'll be training a sequence to sequence model on a dataset of English and French sentences that can translate new sentences from English to French.

Get the Data

Since translating the whole language of English to French will take lots of time to train, we have provided you with a small portion of the English corpus.

```
In [16]:
    """
    DON'T MODIFY ANYTHING IN THIS CELL
    """
    import helper
    import problem_unittests as tests

    source_path = 'data/small_vocab_en'
    target_path = 'data/small_vocab_fr'
    source_text = helper.load_data(source_path)
    target_text = helper.load_data(target_path)
```

Explore the Data

Play around with view_sentence_range to view different parts of the data.

```
In [17]: | view sentence range = (0, 10)
         DON'T MODIFY ANYTHING IN THIS CELL
         import numpy as np
         print('Dataset Stats')
         print('Roughly the number of unique words: {}'.format(len({word: None for word in source text.split()})))
         sentences = source_text.split('\n')
         word counts = [len(sentence.split()) for sentence in sentences]
         print('Number of sentences: {}'.format(len(sentences)))
         print('Average number of words in a sentence: {}'.format(np.average(word counts)))
         print()
         print('English sentences {} to {}:'.format(*view sentence range))
         print('\n'.join(source text.split('\n')[view sentence range[0]:view sentence range[1]]))
         print()
         print('French sentences {} to {}:'.format(*view sentence range))
         print('\n'.join(target text.split('\n')[view sentence range[0]:view sentence range[1]]))
         Dataset Stats
         Roughly the number of unique words: 227
         Number of sentences: 137861
         Average number of words in a sentence: 13.225277634719028
         English sentences 0 to 10:
         new jersey is sometimes quiet during autumn , and it is snowy in april .
         the united states is usually chilly during july , and it is usually freezing in november .
         california is usually quiet during march , and it is usually hot in june .
         the united states is sometimes mild during june , and it is cold in september .
         your least liked fruit is the grape , but my least liked is the apple .
         his favorite fruit is the orange, but my favorite is the grape.
         paris is relaxing during december, but it is usually chilly in july.
         new jersey is busy during spring , and it is never hot in march .
         our least liked fruit is the lemon , but my least liked is the grape .
         the united states is sometimes busy during january , and it is sometimes warm in november .
         French sentences 0 to 10:
         new jersey est parfois calme pendant l'automne, et il est neigeux en avril.
         les états-unis est généralement froid en juillet , et il gèle habituellement en novembre .
```

```
california est généralement calme en mars , et il est généralement chaud en juin . les états-unis est parfois légère en juin , et il fait froid en septembre . votre moins aimé fruit est le raisin , mais mon moins aimé est la pomme . son fruit préféré est l'orange , mais mon préféré est le raisin . paris est relaxant en décembre , mais il est généralement froid en juillet . new jersey est occupé au printemps , et il est jamais chaude en mars . notre fruit est moins aimé le citron , mais mon moins aimé est le raisin . les états-unis est parfois occupé en janvier , et il est parfois chaud en novembre .
```

Implement Preprocessing Function

Text to Word Ids

As you did with other RNNs, you must turn the text into a number so the computer can understand it. In the function <code>text_to_ids()</code> , you'll turn <code>source_text</code> and <code>target_text</code> from words to ids. However, you need to add the <code><EOS></code> word id at the end of each sentence from <code>target_text</code>. This will help the neural network predict when the sentence should end.

You can get the <EOS> word id by doing:

```
target_vocab_to_int['<EOS>']
```

You can get other word ids using source_vocab_to_int and target_vocab_to_int.

Preprocess all the data and save it

Running the code cell below will preprocess all the data and save it to file.

```
In [18]: def text to ids(source text, target text, source vocab to int, target vocab to int):
             Convert source and target text to proper word ids
              :param source text: String that contains all the source text.
              :param target text: String that contains all the target text.
              :param source vocab to int: Dictionary to go from the source words to an id
             :param target_vocab_to_int: Dictionary to go from the target words to an id
              :return: A tuple of lists (source id text, target id text)
             # TODO: Implement Function
             target text = target text.replace(".", ". <EOS>")
             source id text = []
             target id text = []
             for source sentence in source text.split("\n"):
                 source id sentence = [ source vocab to int[word] for word in source sentence.split() ]
                 source id text.append(source id sentence)
             for target sentence in target text.split("\n"):
                 target id sentence = [ target vocab to int[word] for word in target sentence.split() ]
                 target id text.append(target id sentence)
             return source id text, target id text
         DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
         tests.test_text_to_ids(text_to_ids)
```

Check Point

This is your first checkpoint. If you ever decide to come back to this notebook or have to restart the notebook, you can start from here. The preprocessed data has been saved to disk.

Check the Version of TensorFlow and Access to GPU

This will check to make sure you have the correct version of TensorFlow and access to a GPU

TensorFlow Version: 1.0.0 Default GPU Device: /gpu:0

Build the Neural Network

You'll build the components necessary to build a Sequence-to-Sequence model by implementing the following functions below:

- model_inputs
- process_decoding_input
- encoding layer
- decoding_layer_train
- decoding_layer_infer
- decoding_layer
- seq2seq_model

Input

Implement the model_inputs() function to create TF Placeholders for the Neural Network. It should create the following placeholders:

- Input text placeholder named "input" using the TF Placeholder name parameter with rank 2.
- Targets placeholder with rank 2.
- Learning rate placeholder with rank 0.
- Keep probability placeholder named "keep_prob" using the TF Placeholder name parameter with rank 0.

Return the placeholders in the following the tuple (Input, Targets, Learing Rate, Keep Probability)

```
In [22]: def model_inputs():
    """
    Create TF Placeholders for input, targets, and learning rate.
    :return: Tuple (input, targets, learning rate, keep probability)
    """
    # TODO: Implement Function

    inputs = tf.placeholder(tf.int32, [None, None], "input")
    targets = tf.placeholder(tf.int32, [None, None], "target")
    learning_rate = tf.placeholder(tf.float32, name="learning_rate")
    keep_probability = tf.placeholder(tf.float32, name="keep_prob")

    return inputs, targets, learning_rate, keep_probability

"""
    DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
    """
    tests.test_model_inputs(model_inputs)
```

Process Decoding Input

Implement process_decoding_input using TensorFlow to remove the last word id from each batch in target_data and concat the GO ID to the beginning of each batch.

```
In [23]: def process_decoding_input(target_data, target_vocab_to_int, batch_size):
    """
    Preprocess target data for dencoding
    :param target_data: Target Placehoder
    :param batch_size: Batch Size
    :return: Preprocessed target data
    """
    # TODO: Implement Function
    ending = tf.strided_slice(target_data, [0, 0], [batch_size, -1], [1, 1])
    decoded = tf.concat([tf.fill([batch_size, 1], source_vocab_to_int['<GO>']), ending], 1)
    return decoded
    """
    DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
    """
    tests.test_process_decoding_input(process_decoding_input)
```

Encoding

Implement encoding_layer() to create a Encoder RNN layer using tf.nn.dynamic_rnn(https://www.tensorflow.org/api_docs/python/tf/nn/dynamic_rnn).

```
In [24]: | def encoding layer(rnn inputs, rnn size, num layers, keep prob):
             Create encoding layer
              :param rnn inputs: Inputs for the RNN
              :param rnn size: RNN Size
              :param num layers: Number of layers
              :param keep prob: Dropout keep probability
              :return: RNN state
             # TODO: Implement Function
             lstm = tf.contrib.rnn.BasicLSTMCell(rnn size)
             dropout = tf.contrib.rnn.DropoutWrapper(lstm, output keep prob = keep prob)
             cell = tf.contrib.rnn.MultiRNNCell([dropout] * num layers)
             outputs, final state = tf.nn.dynamic rnn(cell, rnn inputs, dtype=tf.float32)
             return final state
         DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
         tests.test encoding layer(encoding layer)
```

Decoding - Training

```
Create training logits using <a href="mainto:tf">tf.contrib.seq2seq.simple_decoder_fn_train()</a>
(<a href="https://www.tensorflow.org/versions/r1.0/api_docs/python/tf/contrib/seq2seq/simple_decoder_fn_train">tf.contrib.seq2seq.dynamic_rnn_decoder_fn_train</a>) and <a href="mainto:tf">tf.contrib.seq2seq.dynamic_rnn_decoder_fn_train</a>) and <a href="
```

```
In [25]: def decoding_layer_train(encoder_state, dec_cell, dec_embed_input, sequence_length, decoding_scope,
                                   output fn, keep prob):
              0.00
              Create a decoding layer for training
              :param encoder state: Encoder State
              :param dec cell: Decoder RNN Cell
              :param dec_embed_input: Decoder embedded input
              :param sequence length: Sequence Length
              :param decoding scope: TenorFlow Variable Scope for decoding
              :param output fn: Function to apply the output layer
              :param keep prob: Dropout keep probability
              :return: Train Logits
              # TODO: Implement Function
             dropout = tf.contrib.rnn.DropoutWrapper(dec_cell, output_keep_prob = keep_prob)
              train decoder = tf.contrib.seq2seq.simple decoder fn train(encoder state)
              outputs, final state, final context state = tf.contrib.seq2seq.dynamic rnn decoder(
                  dropout,
                  train decoder,
                  dec_embed_input,
                  sequence length,
                  scope = decoding scope
              )
              return output fn(outputs)
          0.00
          DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
          tests.test decoding layer train(decoding layer train)
```

Decoding - Inference

Create inference logits using tf.contrib.seq2seq.simple_decoder_fn_inference()

(https://www.tensorflow.org/versions/r1.0/api_docs/python/tf/contrib/seq2seq/simple_decoder_fn_inference) and tf.contrib.seq2seq.dynamic_rnn_decoder()

(https://www.tensorflow.org/versions/r1.0/api_docs/python/tf/contrib/seq2seq/dynamic_rnn_decoder).

```
In [26]: def decoding_layer_infer(encoder_state, dec_cell, dec_embeddings, start_of_sequence_id, end_of_sequence_id,
                                   maximum length, vocab size, decoding scope, output fn, keep prob):
              0.00
              Create a decoding layer for inference
              :param encoder state: Encoder state
              :param dec cell: Decoder RNN Cell
              :param dec_embeddings: Decoder embeddings
              :param start of sequence id: GO ID
              :param end of sequence id: EOS Id
              :param maximum length: The maximum allowed time steps to decode
              :param vocab size: Size of vocabulary
              :param decoding scope: TensorFlow Variable Scope for decoding
              :param output fn: Function to apply the output layer
              :param keep prob: Dropout keep probability
              :return: Inference Logits
              # TODO: Implement Function
              inference decoder fn = tf.contrib.seq2seq.simple decoder fn inference(
                  output fn,
                  encoder_state,
                  dec embeddings,
                  start_of_sequence_id,
                  end of sequence id,
                  maximum length,
                  vocab size
              )
              outputs, final state, final context state = tf.contrib.seq2seq.dynamic rnn decoder(
                  dec cell,
                  inference decoder fn,
                  scope = decoding scope
              )
              return outputs
          DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
         tests.test decoding layer infer(decoding layer infer)
```

Build the Decoding Layer

Implement decoding layer() to create a Decoder RNN layer.

- Create RNN cell for decoding using rnn_size and num_layers.
- Create the output fuction using 1ambda _(https://docs.python.org/3/tutorial/controlflow.html#lambda-expressions) to transform it's input, logits, to class logits.
- Use the your decoding_layer_train(encoder_state, dec_cell, dec_embed_input, sequence_length, decoding_scope, output_fn, keep_prob) function to get the training logits.
- Use your decoding_layer_infer(encoder_state, dec_cell, dec_embeddings, start_of_sequence_id, end_of_sequence_id, maximum_length, vocab_size, decoding_scope, output_fn, keep_prob) function to get the inference logits.

Note: You'll need to use <u>tf.variable_scope</u> (<u>https://www.tensorflow.org/api_docs/python/tf/variable_scope</u>) to share variables between training and inference.

```
In [27]: def decoding_layer(dec_embed_input, dec_embeddings, encoder_state, vocab_size, sequence_length, rnn_size,
                             num layers, target vocab to int, keep prob):
              0.00
              Create decoding layer
              :param dec embed input: Decoder embedded input
              :param dec embeddings: Decoder embeddings
              :param encoder state: The encoded state
              :param vocab size: Size of vocabulary
              :param sequence length: Sequence Length
              :param rnn size: RNN Size
              :param num layers: Number of layers
              :param target vocab to int: Dictionary to go from the target words to an id
              :param keep prob: Dropout keep probability
              :return: Tuple of (Training Logits, Inference Logits)
              # TODO: Implement Function
              lstm = tf.contrib.rnn.BasicLSTMCell(rnn size)
              dropout = tf.contrib.rnn.DropoutWrapper(lstm, output keep prob = keep prob)
              cell = tf.contrib.rnn.MultiRNNCell([dropout] * num layers)
             with tf.variable_scope('decoding') as decoding_scope:
                  output fn = lambda x: tf.contrib.layers.fully connected(
                      Χ,
                      vocab size,
                      None,
                      scope = decoding scope
                  train_logits = decoding_layer_train(
                      encoder_state,
                      cell,
                      dec embed input,
                      sequence_length,
                      decoding scope,
                      output fn,
                      keep_prob
```

```
decoding scope.reuse variables()
        inference logits = decoding layer infer(
            encoder state,
            cell,
            dec embeddings,
            target_vocab_to_int['<GO>'],
            target vocab to int['<EOS>'],
            sequence length,
            vocab_size,
            decoding scope,
            output fn,
            keep_prob
    return train logits, inference logits
0.00
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
tests.test decoding layer(decoding layer)
```

Build the Neural Network

Apply the functions you implemented above to:

- Apply embedding to the input data for the encoder.
- Encode the input using your encoding_layer(rnn_inputs, rnn_size, num_layers, keep_prob).
- Process target data using your process_decoding_input(target_data, target_vocab_to_int, batch_size) function.
- Apply embedding to the target data for the decoder.
- Decode the encoded input using your decoding_layer(dec_embed_input, dec_embeddings, encoder_state, vocab_size, sequence_length, rnn_size, num_layers, target_vocab_to_int, keep_prob).

```
In [28]: def seq2seq model(input_data, target_data, keep_prob, batch_size, sequence_length, source_vocab_size, target_voc
                            enc embedding size, dec embedding size, rnn size, num layers, target vocab to int):
              0.00
              Build the Sequence-to-Sequence part of the neural network
              :param input data: Input placeholder
              :param target data: Target placeholder
              :param keep prob: Dropout keep probability placeholder
              :param batch size: Batch Size
              :param sequence length: Sequence Length
              :param source vocab size: Source vocabulary size
              :param target vocab size: Target vocabulary size
              :param enc_embedding_size: Encoder embedding size
              :param dec embedding size: Decoder embedding size
              :param rnn_size: RNN Size
              :param num layers: Number of layers
              :param target vocab to int: Dictionary to go from the target words to an id
              :return: Tuple of (Training Logits, Inference Logits)
              # TODO: Implement Function
              encoder embed input = tf.contrib.layers.embed sequence(
                  input_data,
                  source_vocab_size,
                  enc embedding size)
              encoder state = encoding layer(
                  encoder_embed_input,
                  rnn_size,
                  num layers,
                  keep prob=keep prob)
              target_data = process_decoding_input(
                  target data,
                  target_vocab_to_int,
                  batch size)
              decoder embed = tf.Variable(tf.random uniform([target vocab size, dec embedding size]))
              decoder embed input = tf.nn.embedding lookup(decoder embed, target data)
              decoder layer = decoding layer(
                  decoder embed input,
```

```
decoder_embed,
    encoder_state,
    target_vocab_size,
    sequence_length,
    rnn_size,
    num_layers,
    target_vocab_to_int,
    keep_prob)

return decoder_layer

"""

DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
"""

tests.test_seq2seq_model(seq2seq_model)
```

Neural Network Training

Hyperparameters

Tune the following parameters:

- Set epochs to the number of epochs.
- Set batch_size to the batch size.
- Set rnn_size to the size of the RNNs.
- Set num_layers to the number of layers.
- $\bullet\,$ Set <code>encoding_embedding_size</code> to the size of the embedding for the encoder.
- Set decoding_embedding_size to the size of the embedding for the decoder.
- Set learning_rate to the learning rate.
- Set keep_probability to the Dropout keep probability

```
In [37]: # Number of Epochs
epochs = 8
    # Batch Size
batch_size = 512
    # RNN Size
    rnn_size = 256
    # Number of Layers
    num_layers = 2
    # Embedding Size
    encoding_embedding_size = 256
    decoding_embedding_size = 256
    # Learning Rate
    learning_rate = 0.001
    # Dropout Keep Probability
    keep_probability = 0.5
```

Build the Graph

Build the graph using the neural network you implemented.

```
0.000
In [38]:
          DON'T MODIFY ANYTHING IN THIS CELL
          save path = 'checkpoints/dev'
          (source_int_text, target_int_text), (source_vocab_to_int, target_vocab_to_int), _ = helper.load_preprocess()
         max source sentence length = max([len(sentence) for sentence in source int text])
         train graph = tf.Graph()
         with train graph.as default():
             input data, targets, lr, keep prob = model inputs()
             sequence length = tf.placeholder with default(max source sentence length, None, name='sequence length')
             input shape = tf.shape(input data)
             train logits, inference logits = seq2seq model(
                 tf.reverse(input_data, [-1]), targets, keep_prob, batch_size, sequence_length, len(source_vocab_to_int),
                 encoding_embedding_size, decoding_embedding_size, rnn_size, num_layers, target_vocab_to_int)
             tf.identity(inference logits, 'logits')
             with tf.name scope("optimization"):
                 # Loss function
                 cost = tf.contrib.seq2seq.sequence loss(
                     train_logits,
                     targets,
                     tf.ones([input shape[0], sequence length]))
                 # Optimizer
                 optimizer = tf.train.AdamOptimizer(lr)
                 # Gradient Clipping
                  gradients = optimizer.compute gradients(cost)
                 capped gradients = [(tf.clip by value(grad, -1., 1.), var) for grad, var in gradients if grad is not Nor
                 train op = optimizer.apply gradients(capped gradients)
```

Train

Train the neural network on the preprocessed data. If you have a hard time getting a good loss, check the forms to see if anyone is having the same problem.

```
0.000
In [39]:
         DON'T MODIFY ANYTHING IN THIS CELL
         import time
         def get_accuracy(target, logits):
             Calculate accuracy
             max seq = max(target.shape[1], logits.shape[1])
             if max_seq - target.shape[1]:
                  target = np.pad(
                      target,
                      [(0,0),(0,max_seq - target.shape[1])],
                      'constant')
             if max_seq - logits.shape[1]:
                  logits = np.pad(
                      logits,
                      [(0,0),(0,\max_{0} - \log its.shape[1]),(0,0)],
                      'constant')
             return np.mean(np.equal(target, np.argmax(logits, 2)))
         train source = source int text[batch size:]
         train target = target int text[batch size:]
         valid source = helper.pad sentence batch(source int text[:batch size])
         valid target = helper.pad sentence batch(target int text[:batch size])
         with tf.Session(graph=train graph) as sess:
             sess.run(tf.global_variables_initializer())
             for epoch i in range(epochs):
                  for batch i, (source batch, target batch) in enumerate(
                          helper.batch data(train source, train target, batch size)):
                      start time = time.time()
                      _, loss = sess.run(
                          [train_op, cost],
                          {input_data: source_batch,
                           targets: target batch,
                           lr: learning rate,
```

```
sequence length: target batch.shape[1],
             keep prob: keep probability})
        batch train logits = sess.run(
            inference logits,
            {input data: source_batch, keep_prob: 1.0})
        batch valid logits = sess.run(
            inference logits,
            {input data: valid source, keep prob: 1.0})
        train acc = get accuracy(target batch, batch train logits)
        valid acc = get accuracy(np.array(valid target), batch valid logits)
        end time = time.time()
        print('Epoch {:>3} Batch {:>4}/{} - Train Accuracy: {:>6.3f}, Validation Accuracy: {:>6.3f}, Loss:
              .format(epoch i, batch i, len(source int text) // batch size, train acc, valid acc, loss))
# Save Model
saver = tf.train.Saver()
saver.save(sess, save path)
print('Model Trained and Saved')
```

```
0 Batch
                   0/269 - Train Accuracy:
                                           0.242, Validation Accuracy: 0.310, Loss:
                                                                                      5.868
Epoch
Epoch
        0 Batch
                   1/269 - Train Accuracy:
                                           0.233, Validation Accuracy: 0.310, Loss:
                                                                                      5.494
Epoch
        0 Batch
                   2/269 - Train Accuracy:
                                           0.266, Validation Accuracy: 0.310, Loss:
                                                                                      5.034
Epoch
        0 Batch
                   3/269 - Train Accuracy:
                                           0.245, Validation Accuracy: 0.310, Loss:
                                                                                      4.795
Epoch
        0 Batch
                   4/269 - Train Accuracy:
                                           0.244, Validation Accuracy:
                                                                        0.322, Loss:
                                                                                      4.728
Epoch
       0 Batch
                   5/269 - Train Accuracy:
                                           0.265, Validation Accuracy:
                                                                        0.339, Loss:
                                                                                      4.594
Epoch
        0 Batch
                   6/269 - Train Accuracy:
                                           0.313, Validation Accuracy:
                                                                        0.342, Loss:
                                                                                      4.186
Epoch
        0 Batch
                  7/269 - Train Accuracy:
                                           0.320, Validation Accuracy:
                                                                        0.351, Loss:
                                                                                      4.101
Epoch
        0 Batch
                   8/269 - Train Accuracy:
                                           0.289, Validation Accuracy:
                                                                        0.351, Loss:
                                                                                      4.159
Epoch
        0 Batch
                   9/269 - Train Accuracy:
                                           0.314, Validation Accuracy:
                                                                        0.354, Loss:
                                                                                      3.921
Epoch
        0 Batch
                 10/269 - Train Accuracy:
                                           0.283, Validation Accuracy: 0.354, Loss:
                                                                                      3.971
Epoch
        0 Batch
                 11/269 - Train Accuracy:
                                           0.333, Validation Accuracy:
                                                                        0.367, Loss:
                                                                                      3.760
                                                                        0.372, Loss:
Epoch
        0 Batch
                 12/269 - Train Accuracy:
                                           0.311, Validation Accuracy:
                                                                                      3.824
                                                                        0.373, Loss:
Epoch
        0 Batch
                 13/269 - Train Accuracy:
                                           0.375, Validation Accuracy:
                                                                                      3.464
Epoch
        0 Batch
                 14/269 - Train Accuracy:
                                           0.342, Validation Accuracy:
                                                                        0.376, Loss: 3.574
Epoch
        0 Batch
                 15/269 - Train Accuracy:
                                           0.331, Validation Accuracy:
                                                                        0.374, Loss: 3.551
Epoch
                 16/269 - Train Accuracy:
                                           0.353, Validation Accuracy:
                                                                        0.382, Loss:
        0 Batch
                                                                                      3.443
Epoch
        0 Batch
                 17/269 - Train Accuracy:
                                           0.343, Validation Accuracy:
                                                                        0.380, Loss:
Epoch
        0 Batch
                  18/269 - Train Accuracy:
                                           0.320, Validation Accuracy: 0.386, Loss:
```

Save Parameters

Save the batch size and save path parameters for inference.

Checkpoint

```
In [41]:
    DON'T MODIFY ANYTHING IN THIS CELL
    import tensorflow as tf
    import numpy as np
    import helper
    import problem_unittests as tests
    _, (source_vocab_to_int, target_vocab_to_int), (source_int_to_vocab, target_int_to_vocab) = helper.load_preprocelload_path = helper.load_params()
```

Sentence to Sequence

To feed a sentence into the model for translation, you first need to preprocess it. Implement the function sentence_to_seq() to preprocess new sentences.

- Convert the sentence to lowercase
- Convert words into ids using vocab_to_int
 - Convert words not in the vocabulary, to the <UNK> word id.

```
In [42]: def sentence_to_seq(sentence, vocab_to_int):
    """
    Convert a sentence to a sequence of ids
    :param sentence: String
    :param vocab_to_int: Dictionary to go from the words to an id
    :return: List of word ids
    """
    # TODO: Implement Function

    unk = vocab_to_int['<UNK>']
    sentence = sentence.lower()
    ids = [vocab_to_int.get(w, unk) for w in sentence.split()]
    return ids

DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
    """
    tests.test_sentence_to_seq(sentence_to_seq)
```

Translate

This will translate translate_sentence from English to French.

```
In [44]: | translate sentence = 'he saw a old yellow truck .'
         0.00
         DON'T MODIFY ANYTHING IN THIS CELL
         translate sentence = sentence to seq(translate sentence, source vocab to int)
         loaded graph = tf.Graph()
         with tf.Session(graph=loaded_graph) as sess:
             # Load saved model
             loader = tf.train.import meta graph(load path + '.meta')
             loader.restore(sess, load path)
             input data = loaded graph.get tensor by name('input:0')
             logits = loaded graph.get tensor by name('logits:0')
             keep prob = loaded graph.get tensor by name('keep prob:0')
             translate logits = sess.run(logits, {input data: [translate sentence], keep prob: 1.0})[0]
         print('Input')
         print(' Word Ids:
                             {}'.format([i for i in translate sentence]))
                  English Words: {}'.format([source int to vocab[i] for i in translate sentence]))
         print('
         print('\nPrediction')
         print(' Word Ids:
                                 {}'.format([i for i in np.argmax(translate logits, 1)]))
         print(' French Words: {}'.format([target int to vocab[i] for i in np.argmax(translate logits, 1)]))
         Input
           Word Ids:
                          [114, 76, 50, 60, 192, 129, 179]
           English Words: ['he', 'saw', 'a', 'old', 'yellow', 'truck', '.']
         Prediction
           Word Ids:
                          [212, 297, 318, 145, 112, 106, 292, 166, 1]
           French Words: ['il', 'a', 'vu', 'un', 'camion', 'jaune', 'brillant', '.', '<EOS>']
In [ ]:
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