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

ge[1]]))

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.

```
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 [10]:
view_sentence_range = (1100, 1110)
,,,,,,
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_ran
ge[1]]))
print()
print('French sentences {} to {}:'.format(*view_sentence_range))
print('\n'.join(target_text.split('\n')[view_sentence_range[0]:view_sentence_ran
```

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In [1]:

Language Translation

```
Dataset Stats
Roughly the number of unique words: 227
Number of sentences: 137861
Average number of words in a sentence: 13.225277634719028
English sentences 1100 to 1110:
he likes grapes , mangoes , and bananas.
French sentences 1100 to 1110:
il aime les raisins , les mangues et les bananes .
```

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 $\texttt{text_to_ids}()$, you'll turn $\texttt{source_text}$ and $\texttt{target_text}$ from words to ids. However, you need to add the <EOS> word id at the end of each sentence from $\texttt{target_text}$. 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.
                                                                          In [13]:
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
    source_id_text = [[source_vocab_to_int[word]
                       for word in sentence.split()]
                       for sentence in source_text.split('\n')]
    target_id_text = [[target_vocab_to_int[word] for word in sentence.split()] +
                      [target_vocab_to_int['<EOS>'] ]
                       for sentence in target_text.split('\n')]
    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)
Tests Passed
```

Preprocess all the data and save it

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

```
In [14]:
,,,,,,
DON'T MODIFY ANYTHING IN THIS CELL
helper.preprocess_and_save_data(source_path, target_path, 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.

```
,,,,,,
DON'T MODIFY ANYTHING IN THIS CELL
import numpy as np
import helper
(source_int_text, target_int_text), (source_vocab_to_int, target_vocab_to_int),
_ = helper.load_preprocess()
```

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

```
In [16]:
```

In [15]:

```
,,,,,,
DON'T MODIFY ANYTHING IN THIS CELL
from distutils.version import LooseVersion
import warnings
import tensorflow as tf
# Check TensorFlow Version
assert LooseVersion(tf.__version__) in [LooseVersion('1.0.0'),
LooseVersion('1.0.1')], 'This project requires TensorFlow version 1.0 You are
using {}'.format(tf.__version__)
print('TensorFlow Version: {}'.format(tf.__version__))
# Check for a GPU
if not tf.test.gpu_device_name():
    warnings.warn('No GPU found. Please use a GPU to train your neural
network.')
else:
    print('Default GPU Device: {}'.format(tf.test.gpu_device_name()))
TensorFlow Version: 1.0.0
C:\Users\VadymSerpak\Anaconda3\envs\Python35\lib\site-
```

packages\ipykernel launcher.py:14: UserWarning: No GPU found. Please use a GPU to train your neural network.

Build the Neural Network

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

- model_inputsprocess_decoding_inputencoding_layerdecoding_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)

```
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, shape=(None, None), name='input') #
Perhaps inputs would have been better name??
    targets = tf.placeholder(tf.int32, shape=(None, None))
    learning_rate = tf.placeholder(tf.float32)
    keep_prob = tf.placeholder(tf.float32, name='keep_prob')
    return (inputs, targets, learning_rate, keep_prob)

"""
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.

```
def process_decoding_input(target_data, target_vocab_to_int, batch_size):
    Preprocess target data for dencoding
    :param target_data: Target Placehoder
    :param target_vocab_to_int: Dictionary to go from the target words to an id
    :param batch_size: Batch Size
    :return: Preprocessed target data
    # TODO: Implement Function
    # Remove last word id
    stride_slice = tf.strided_slice(target_data,
                                     begin=[0, 0],
                                     end=[batch_size, -1],
                                     strides=[1, 1])
     # Concat GO id
    concat = tf.concat([tf.fill([batch_size, 1],
                                 target_vocab_to_int['<GO>']),
                        stride_slice], 1)
    return concat
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
tests.test_process_decoding_input(process_decoding_input)
Tests Passed
Encoding
Implement encoding layer() to create a Encoder RNN layer using tf.nn.dynamic rnn().
                                                                           In [19]:
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
    cell =
tf.contrib.rnn.MultiRNNCell([tf.contrib.rnn.BasicLSTMCell(rnn_size)]*num_layers)
    drop = tf.contrib.rnn.DropoutWrapper(cell, input_keep_prob=1,
output_keep_prob=keep_prob)
    rnn_out, rnn_state = tf.nn.dynamic_rnn(cell=drop, inputs=rnn_inputs,
dtype=tf.float32)
    return rnn_state
```

```
Language Translation
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
tests.test_encoding_layer(encoding_layer)
Tests Passed
Decoding - Training
Create training logits
using tf.contrib.seq2seq.simple decoder fn train() and tf.contrib.seq2seq.dynamic rn
n decoder(). Apply the output fn to the tf.contrib.seq2seq.dynamic rnn decoder() outputs.
                                                                            In [22]:
def decoding_layer_train(encoder_state, dec_cell, dec_embed_input,
sequence_length, decoding_scope,
                          output_fn, keep_prob):
    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
    111111
    # TODO: Implement Function
    train_decoder_fn = tf.contrib.seq2seq.simple_decoder_fn_train(encoder_state)
    train_pred, _, _ = tf.contrib.seq2seq.dynamic_rnn_decoder(cell=dec_cell,
decoder_fn=train_decoder_fn,
inputs=dec_embed_input,
sequence_length=sequence_length,
scope=decoding_scope)
    train_logits = output_fn(train_pred)
    return train_logits
,,,,,,,
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
tests.test_decoding_layer_train(decoding_layer_train)
Tests Passed
Decoding - Inference
Create inference logits
using tf.contrib.seq2seq.simple decoder fn inference() and tf.contrib.seq2seq.dynami
c rnn decoder().
```

```
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):
    ,,,,,,
    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
    infer_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)
    inference_logits, _, _ = tf.contrib.seq2seq.dynamic_rnn_decoder(dec_cell,
infer_decoder_fn,
scope=decoding_scope)
    return inference_logits
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
tests.test_decoding_layer_infer(decoding_layer_infer)
Tests Passed
```

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 lambda 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 tf.variable_scope 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):
    ,,,,,,,
    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
    cell = tf.contrib.rnn.BasicLSTMCell(rnn_size)
    dec_cell = tf.contrib.rnn.MultiRNNCell([cell] * num_layers)
    with tf.variable_scope("decoding") as decoding_scope:
        output_fn = lambda x: tf.contrib.layers.fully_connected(x,
                                                                 vocab_size,
                                                                 None,
scope=decoding_scope)
        train_output = decoding_layer_train(encoder_state,
                                             dec_cell,
                                             dec_embed_input,
                                             sequence_length,
                                             decoding_scope,
                                             output_fn,
                                             keep_prob)
```

with tf.variable_scope("decoding", reuse=True) as decoding_scope:

inf_output = decoding_layer_infer(encoder_state,

```
Language Translation
                                             dec_cell,
                                             dec_embeddings,
                                             helper.CODES['<GO>'].
                                             helper.CODES['<EOS>'],
                                             sequence_length,
                                             vocab_size,
                                             decoding_scope,
                                             output_fn,
                                             keep_prob)
    return (train_output, inf_output)
,,,,,,,
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
tests.test_decoding_layer(decoding_layer)
Tests Passed
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_vocab_size,
                   enc_embedding_size, dec_embedding_size, rnn_size, num_layers,
target_vocab_to_int):
    111111
    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
```

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: Decoder embedding size

:param dec_embedding_size: Encoder 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)

"""

```
Language Translation
    # TODO: Implement Function
    enc_embed_input = tf.contrib.layers.embed_sequence(ids=input_data,
vocab_size=source_vocab_size.
embed_dim=enc_embedding_size)
    encode_state = encoding_layer(rnn_inputs=enc_embed_input,
                                   rnn_size=rnn_size,
                                   num_layers=num_layers,
                                   keep_prob=keep_prob)
    proc_target_data = process_decoding_input(target_data=target_data,
target_vocab_to_int=target_vocab_to_int, batch_size=batch_size)
    dec_embeddings = tf.Variable(tf.random_uniform([target_vocab_size,
dec_embedding_size]))
    dec_embed_input = tf.nn.embedding_lookup(dec_embeddings, proc_target_data)
    train_logits, infer_logits = decoding_layer(dec_embed_input=dec_embed_input,
                                                 dec_embeddings=dec_embeddings,
                                                 encoder_state=encode_state,
                                                 vocab_size=target_vocab_size,
                                                 sequence_length=sequence_length,
                                                 rnn_size=rnn_size,
                                                 num_layers=num_layers,
```

return train_logits, infer_logits

"""

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

tests.test_seq2seq_model(seq2seq_model)
Tests Passed

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.
- Set encoding_embedding_size 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

Language Translation

```
In [29]:
```

```
# Number of Epochs
epochs = 20
# Batch Size
batch_size = 256
# RNN Size
rnn_size = 30
# Number of Layers
num_layers = 2
# Embedding Size
encoding_embedding_size = 25
decoding_embedding_size = 25
# Learning Rate
learning_rate = 0.005
# Dropout Keep Probability
keep_probability = 0.7
```

Build the Graph

Build the graph using the neural network you implemented.

```
In [30]:
```

```
111111
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), len(target_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]))
```

```
Language Translation
```

```
# 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 None]
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.

```
,,,,,,,
DON'T MODIFY ANYTHING IN THIS CELL
import time
def get_accuracy(target, logits):
    Calculate accuracy
    max_seg = max(target.shape[1], logits.shape[1])
    if max_seq - target.shape[1]:
        target = np.pad(
            target,
            [(0,0),(0,\max_{e} - target.shape[1])],
            'constant')
    if max_seq - logits.shape[1]:
        logits = np.pad(
            logits,
            [(0,0),(0,\max_{e} - \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()
```

In [31]:

```
Language Translation
```

```
_, loss = sess.run(
                [train_op, cost],
                {input_data: source_batch,
                 targets: target_batch,
                 1r: 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: {:>6.3f}'
                  .format(epoch_i, batch_i, len(source_int_text) // batch_size,
train_acc, valid_acc, loss))
    # Save Mode1
    saver = tf.train.Saver()
    saver.save(sess, save_path)
    print('Model Trained and Saved')
                  0/538 - Train Accuracy: 0.209, Validation Accuracy: 0.291,
Epoch 0 Batch
Loss: 5.889
Epoch 0 Batch 1/538 - Train Accuracy: 0.231, Validation Accuracy: 0.316,
Loss: 5.828
Epoch 19 Batch 535/538 - Train Accuracy: 0.958, Validation Accuracy: 0.952,
Loss: 0.030
Epoch 19 Batch 536/538 - Train Accuracy: 0.944, Validation Accuracy: 0.953,
Loss: 0.038
Model Trained and Saved
```

Save Parameters

Save the batch size and save path parameters for inference.

```
In [32]:
```

```
"""

DON'T MODIFY ANYTHING IN THIS CELL
"""

# Save parameters for checkpoint
helper.save_params(save_path)
```

Checkpoint

```
"""
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_preprocess()
load_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 [35]:
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
    sentence = sentence.lower()
    words = sentence.split()
    word_id_list = [vocab_to_int.get(word, vocab_to_int['<UNK>'])
                    for word in words]
    return word_id_list
DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
tests.test_sentence_to_seq(sentence_to_seq)
Tests Passed
Translate
This will translate translate sentence from English to French.
                                                                           In [37]:
translate_sentence = 'he saw a old yellow truck .'
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]))
print(' English Words: {}'.format([source_int_to_vocab[i] for i in
translate_sentence]))
print('\nPrediction')
print('
        Word Ids:
                        {}'.format([i for i in np.argmax(translate_logits, 1)]))
         French words: {}'.format([target_int_to_vocab[i] for i in
print('
np.argmax(translate_logits, 1)]))
Input
  Word Ids:
                 [63, 168, 187, 146, 175, 148, 116]
 English Words: ['he', 'saw', 'a', 'old', 'yellow', 'truck', '.']
Prediction
 Word Ids:
                 [70, 332, 350, 228, 30, 180, 234, 126, 1]
  French Words: ['il', 'a', 'vu', 'un', 'vieille', 'voiture', 'jaune', '.',
'<EOS>'1
```

Imperfect Translation

You might notice that some sentences translate better than others. Since the dataset you're using only has a vocabulary of 227 English words of the thousands that you use, you're only going to see good results using these words. For this project, you don't need a perfect translation. However, if you want to create a better translation model, you'll need better data.

You can train on the <u>WMT10 French-English corpus</u>. This dataset has more vocabulary and richer in topics discussed. However, this will take you days to train, so make sure you've a GPU and the neural network is performing well on dataset we provided. Just make sure you play with the WMT10 corpus after you've submitted this project.

Helper.py

```
Language Translation
import os
import pickle import copy
import numpy as np
CODES = {'<PAD>': 0, '<EOS>': 1, '<UNK>': 2, '<GO>': 3 }
def load_data(path):
    Load Dataset from File
    input_file = os.path.join(path)
    with open(input_file, 'r', encoding='utf-8') as f:
        data = f.read()
    return data
def preprocess_and_save_data(source_path, target_path, text_to_ids):
    Preprocess Text Data. Save to to file.
    # Preprocess
    source_text = load_data(source_path)
    target_text = load_data(target_path)
    source_text = source_text.lower()
    target_text = target_text.lower()
    source_vocab_to_int, source_int_to_vocab = create_lookup_tables(source_text)
    target_vocab_to_int, target_int_to_vocab = create_lookup_tables(target_text)
    source_text, target_text = text_to_ids(source_text, target_text,
source_vocab_to_int, target_vocab_to_int)
    # Save Data
    pickle.dump((
         (source_text, target_text),
        (source_vocab_to_int, target_vocab_to_int),
(source_int_to_vocab, target_int_to_vocab)), open('preprocess.p','wb'))
def load_preprocess():
    Load the Preprocessed Training data and return them in batches of
<batch_size> or less
    return pickle.load(open('preprocess.p', mode='rb'))
def create_lookup_tables(text):
    Create lookup tables for vocabulary
    vocab = set(text.split())
    vocab_to_int = copy.copy(CODES)
    for v_i, v in enumerate(vocab, len(CODES)):
        vocab\_to\_int[v] = v\_i
    int_to_vocab = {v_i: v for v, v_i in vocab_to_int.items()}
    return vocab_to_int, int_to_vocab
def save_params(params):
    Save parameters to file
    pickle.dump(params, open('params.p', 'wb'))
def load_params():
    Load parameters from file
```

```
return pickle.load(open('params.p', mode='rb'))

def batch_data(source, target, batch_size):
    Batch source and target together
    for batch_i in range(0, len(source)//batch_size):
        start_i = batch_i * batch_size
        source_batch = source[start_i:start_i + batch_size]
        target_batch = target[start_i:start_i + batch_size]
        yield np.array(pad_sentence_batch(source_batch)),

np.array(pad_sentence_batch(target_batch))

def pad_sentence_batch(sentence_batch):
    Pad sentence with <PAD> id
    max_sentence = max([len(sentence) for sentence in sentence_batch])
    return [sentence + [CODES['<PAD>']] * (max_sentence - len(sentence))
        for sentence in sentence_batch]
```