TV Script Generation

In this project, you'll generate your own <u>Simpsons (https://en.wikipedia.org/wiki/The Simpsons)</u> TV scripts using RNNs. You'll be using part of the <u>Simpsons dataset (https://www.kaggle.com/wcukierski/the-simpsons-by-the-data)</u> of scripts from 27 seasons. The Neural Network you'll build will generate a new TV script for a scene at <u>Moe's Tavern (https://simpsonswiki.com/wiki/Moe's Tavern)</u>.

Get the Data

The data is already provided for you. You'll be using a subset of the original dataset. It consists of only the scenes in Moe's Tavern. This doesn't include other versions of the tavern, like "Moe's Cavern", "Flaming Moe's", "Uncle Moe's Family Feed-Bag", etc..

Explore the Data

Play around with view sentence range to view different parts of the data.

```
In [ ]: 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 text.spli
        t()})))
        scenes = text.split('\n\n')
        print('Number of scenes: {}'.format(len(scenes)))
        sentence count scene = [scene.count('\n') for scene in scenes]
        print('Average number of sentences in each scene: {}'.format(np.average(sentence_count_scen
        e)))
        sentences = [sentence for scene in scenes for sentence in scene.split('\n')]
        print('Number of lines: {}'.format(len(sentences)))
        word_count_sentence = [len(sentence.split()) for sentence in sentences]
print('Average number of words in each line: {}'.format(np.average(word_count_sentence)))
        print()
        print('The sentences {} to {}:'.format(*view_sentence_range))
```

Implement Preprocessing Functions

The first thing to do to any dataset is preprocessing. Implement the following preprocessing functions below:

- Lookup Table
- Tokenize Punctuation

Lookup Table

To create a word embedding, you first need to transform the words to ids. In this function, create two dictionaries:

- Dictionary to go from the words to an id, we'll call vocab_to_int
- Dictionary to go from the id to word, we'll call int to vocab

Return these dictionaries in the following tuple (vocab_to_int, int_to_vocab)

Tokenize Punctuation

We'll be splitting the script into a word array using spaces as delimiters. However, punctuations like periods and exclamation marks make it hard for the neural network to distinguish between the word "bye" and "bye!".

Implement the function token_lookup to return a dict that will be used to tokenize symbols like "!" into "||Exclamation_Mark||". Create a dictionary for the following symbols where the symbol is the key and value is the token:

- Period (.)
- Comma (,)
- Quotation Mark (")
- Semicolon (;)
- Exclamation mark (!)
- Question mark (?)
- Left Parentheses (()
- Right Parentheses ())
- Dash (--)
- Return (\n)

This dictionary will be used to token the symbols and add the delimiter (space) around it. This separates the symbols as it's own word, making it easier for the neural network to predict on the next word. Make sure you don't use a token that could be confused as a word. Instead of using the token "dash", try using something like "||dash||".

Preprocess all the data and save it

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

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.

```
In []:
    DON'T MODIFY ANYTHING IN THIS CELL
    import helper
    import numpy as np
    import problem_unittests as tests
    int_text, vocab_to_int, int_to_vocab, token_dict = helper.load_preprocess()
```

Build the Neural Network

You'll build the components necessary to build a RNN by implementing the following functions below:

- get_inputs
- get_init_cell
- get_embed
- build_rnn
- build_nn
- get_batches

Check the Version of TensorFlow and Access to GPU

```
In []:
    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__) >= LooseVersion('1.0'), 'Please use TensorFlow version
    1.0 or newer'
    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()))
```

Input

Implement the get_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 (https://www.tensorflow.org/api_docs/python/tf/placeholder) name parameter.
- · Targets placeholder
- Learning Rate placeholder

Return the placeholders in the following tuple (Input, Targets, LearningRate)

Build RNN Cell and Initialize

Stack one or more <u>BasicLSTMCells</u> (https://www.tensorflow.org/api docs/python/tf/contrib/rnn/BasicLSTMCell) in a <u>MultiRNNCell</u> (https://www.tensorflow.org/api docs/python/tf/contrib/rnn/MultiRNNCell).

- The Rnn size should be set using rnn_size
- Initalize Cell State using the MultiRNNCell's sero state() (https://www.tensorflow.org/api_docs/python/tf/contrib/rnn/MultiRNNCell#zero_state) function
 - Apply the name "initial_state" to the initial state using tf.identity() (https://www.tensorflow.org/api docs/python/tf/identity)

Return the cell and initial state in the following tuple (Cell, InitialState)

Word Embedding

Apply embedding to input_data using TensorFlow. Return the embedded sequence.

Build RNN

You created a RNN Cell in the get_init_cell() function. Time to use the cell to create a RNN.

- Build the RNN using the tf.nn.dynamic rnn() (https://www.tensorflow.org/api_docs/python/tf/nn/dynamic rnn)
 - Apply the name "final_state" to the final state using tf.identity() (https://www.tensorflow.org/api_docs/python/tf/identity)

Return the outputs and final_state state in the following tuple (Outputs, FinalState)

Build the Neural Network

Apply the functions you implemented above to:

- Apply embedding to input_data using your get_embed(input_data, vocab_size, embed_dim) function.
- Build RNN using cell and your build_rnn(cell, inputs) function.
- Apply a fully connected layer with a linear activation and vocab_size as the number of outputs.

Return the logits and final state in the following tuple (Logits, FinalState)

```
In []: def build_nn(cell, rnn_size, input_data, vocab_size, embed_dim):
    """
    Build part of the neural network
    :param cell: RNN cell
    :param rnn_size: Size of rnns
    :param input_data: Input data
    :param vocab_size: Vocabulary size
    :param embed_dim: Number of embedding dimensions
    :return: Tuple (Logits, FinalState)
    """
    # TODO: Implement Function
    return None, None

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

Batches

Implement get_batches to create batches of input and targets using int_text. The batches should be a Numpy array with the shape (number of batches, 2, batch size, sequence length). Each batch contains two elements:

- The first element is a single batch of **input** with the shape [batch size, sequence length]
- The second element is a single batch of targets with the shape [batch size, sequence length]

If you can't fill the last batch with enough data, drop the last batch.

For example, get_batches([1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20], 3, 2) would return a Numpy array of the following:

```
# First Batch
   # Batch of Input
   [[ 1 2], [ 7 8], [13 14]]
   # Batch of targets
    [[ 2 3], [ 8 9], [14 15]]
 1
 # Second Batch
   # Batch of Input
   [[ 3 4], [ 9 10], [15 16]]
   # Batch of targets
   [[ 4 5], [10 11], [16 17]]
 1
 # Third Batch
   # Batch of Input
   [[ 5 6], [11 12], [17 18]]
   # Batch of targets
   [[ 6 7], [12 13], [18 1]]
 ]
]
```

Notice that the last target value in the last batch is the first input value of the first batch. In this case, 1. This is a common technique used when creating sequence batches, although it is rather unintuitive.

Neural Network Training

Hyperparameters

Tune the following parameters:

- Set num_epochs to the number of epochs.
- Set batch_size to the batch size.
- \bullet Set rnn_size to the size of the RNNs.
- Set embed_dim to the size of the embedding.
- Set seq_length to the length of sequence.Set learning_rate to the learning rate.
- Set show_every_n_batches to the number of batches the neural network should print progress.

```
In []: # Number of Epochs
    num_epochs = None
    # Batch Size
    batch_size = None
    # RNN Size
    rnn_size = None
    # Embedding Dimension Size
    embed_dim = None
    # Sequence Length
    seq_length = None
    # Learning Rate
    learning_rate = None
    # Show stats for every n number of batches
    show_every_n_batches = None

"""

DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
"""
save_dir = './save'
```

Build the Graph

Build the graph using the neural network you implemented.

```
In [ ]:
        DON'T MODIFY ANYTHING IN THIS CELL
        from tensorflow.contrib import seq2seq
        train_graph = tf.Graph()
        with train graph.as default():
            vocab_size = len(int_to_vocab)
            input_text, targets, lr = get_inputs()
            input data shape = tf.shape(input text)
            cell, initial_state = get_init_cell(input_data_shape[0], rnn_size)
            logits, final_state = build_nn(cell, rnn_size, input_text, vocab_size, embed_dim)
            # Probabilities for generating words
            probs = tf.nn.softmax(logits, name='probs')
            # Loss function
            cost = seq2seq.sequence_loss(
                logits,
                targets
                tf.ones([input_data_shape[0], input_data_shape[1]]))
            # 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 i
        f 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 <u>forums (https://discussions.udacity.com/)</u> to see if anyone is having the same problem.

```
In [ ]: """
        DON'T MODIFY ANYTHING IN THIS CELL
        batches = get_batches(int_text, batch_size, seq_length)
        with tf.Session(graph=train graph) as sess:
            sess.run(tf.global_variables_initializer())
            for epoch i in range(num_epochs):
                 state = sess.run(initial_state, {input_text: batches[0][0]})
                 for batch_i, (x, y) in enumerate(batches):
                     feed = {
                         input text: x,
                         targets: y,
                         initial_state: state,
                         lr: learning_rate}
                     train_loss, state, _ = sess.run([cost, final_state, train_op], feed)
                     # Show every <show_every_n_batches> batches
                     if (epoch_i * len(batches) + batch_i) % show_every_n_batches == 0:
                         print('Epoch {:>3} Batch {:>4}/{} train loss = {:.3f}'.format(
                            epoch_i,
                             batch i,
                             len(batches),
                             train_loss))
            # Save Model
            saver = tf.train.Saver()
            saver.save(sess, save dir)
            print('Model Trained and Saved')
```

Save Parameters

Save seq_length and save_dir for generating a new TV script.

Checkpoint

Implement Generate Functions

Get Tensors

Get tensors from loaded_graph using the function <u>get_tensor_by_name()</u> (https://www.tensorflow.org/api_docs/python_tt/Graph#get_tensor_by_name). Get the tensors using the following names:

- "input:0"
- "initial state:0"
- "final_state:0"
- "probs:0"

Return the tensors in the following tuple (InputTensor, InitialStateTensor, FinalStateTensor, ProbsTensor)

Choose Word

Implement the pick_word() function to select the next word using probabilities.

```
In []: def pick_word(probabilities, int_to_vocab):
    """
    Pick the next word in the generated text
    :param probabilities: Probabilites of the next word
    :param int_to_vocab: Dictionary of word ids as the keys and words as the values
    :return: String of the predicted word
    """
    # TODO: Implement Function
    return None

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

Generate TV Script

This will generate the TV script for you. Set gen_length to the length of TV script you want to generate.

```
In [ ]: gen length = 200
         # homer_simpson, moe_szyslak, or Barney_Gumble
         prime_word = 'moe_szyslak'
         DON'T MODIFY ANYTHING IN THIS CELL THAT IS BELOW THIS LINE
         loaded_graph = tf.Graph()
         with tf.Session(graph=loaded graph) as sess:
              # Load saved model
              loader = tf.train.import_meta_graph(load_dir + '.meta')
              loader.restore(sess, load dir)
              # Get Tensors from loaded model
              input_text, initial_state, final_state, probs = get_tensors(loaded_graph)
             # Sentences generation setup
              gen_sentences = [prime_word + ':']
              prev_state = sess.run(initial_state, {input_text: np.array([[1]])})
              # Generate sentences
              for n in range(gen length):
                  # Dynamic Input
                  dyn_input = [[vocab_to_int[word] for word in gen_sentences[-seq_length:]]]
                  dyn_seq_length = len(dyn_input[0])
                  # Get Prediction
                  probabilities, prev_state = sess.run(
                       [probs, final_state],
                       {input text: dyn input, initial state: prev state})
                  pred_word = pick_word(probabilities[dyn_seq_length-1], int_to_vocab)
                  gen_sentences.append(pred_word)
              # Remove tokens
              tv_script = ' '.join(gen_sentences)
             for key, token in token_dict.items():
    ending = ' ' if key in ['\n', '(', '"'] else ''
    tv_script = tv_script.replace(' ' + token.lower(), key)
              tv_script = tv_script.replace('\n', '\n')
tv_script = tv_script.replace('(', '('))
              print(tv_script)
```

The TV Script is Nonsensical

It's ok if the TV script doesn't make any sense. We trained on less than a megabyte of text. In order to get good results, you'll have to use a smaller vocabulary or get more data. Luckily there's more data! As we mentioned in the beggining of this project, this is a subset of <u>another dataset</u> (https://www.kaggle.com/wcukierski/the-simpsons-by-the-data). We didn't have you train on all the data, because that would take too long. However, you are free to train your neural network on all the data. After you complete the project, of course.

Submitting This Project

When submitting this project, make sure to run all the cells before saving the notebook. Save the notebook file as "dlnd_tv_script_generation.ipynb" and save it as a HTML file under "File" -> "Download as". Include the "helper.py" and "problem_unittests.py" files in your submission.