

## PROJECT

## Generate TV Scripts

A part of the Deep Learning Nanodegree Program

## PROJECT REVIEW

## CODE REVIEW

## NOTES

SHARE YOUR ACCOMPLISHMENT!  

## Meets Specifications

Congratulations you passed this project.

Now you should implement your own RNN from scratch for better understanding. Following links would help you (You can any time take help from mentors)

<http://www.wildml.com/2015/09/recurrent-neural-networks-tutorial-part-1-introduction-to-rnns/>

<https://www.youtube.com/watch?v=cO0a0QYmFm8&index=10&list=PLIjy-eBtNFt6EuMxFYRiNR507MCWN5UIA>

Good Luck for next project.

## Required Files and Tests



The project submission contains the project notebook, called "dln\_tv\_script\_generation.ipynb".



All the unit tests in project have passed.

Great work, All unit test are working fine.

Unit testing is very good practice to ensure that your code is free from all bugs without getting confused and

prevent you from wasting a lot of time while debugging minor things. Unit test also help in improving our code standards. For more details read this(<https://cgoldberg.github.io/python-unittest-tutorial/>) and I really hope that you will continue to use unit testing in every module that you write to keep it clean and speed up your development and for quality development. Unit testing is highly motivated in industries.

It is not always that if you passed unit test your code is okay, there can be some errors.

## Preprocessing



The function `create_lookup_tables` 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`

The function `create_lookup_tables` return these dictionaries in the a tuple (`vocab_to_int`, `int_to_vocab`)



The function `token_lookup` returns a dict that can correctly tokenizes the provided symbols.

## Build the Neural Network



Implemented the `get_inputs` function to create TF Placeholders for the Neural Network with the following placeholders:

- Input text placeholder named "input" using the TF Placeholder name parameter.
- Targets placeholder
- Learning Rate placeholder

The `get_inputs` function return the placeholders in the following the tuple (Input, Targets, LearningRate)

Fantastic! You've perfectly used "None"'s to adequately create a dynamic sized placeholder variables.

`get_inputs` function is correctly implemented. Seems that you have good grasp of TF Placeholder, I appreciate your efforts.

There is a very good course offered by Stanford University, CS 20SI: Tensorflow for Deep Learning Research. This would help you in further understanding of Tensorflow and its application.



The `get_init_cell` function does the following:

- Stacks one or more BasicLSTMCells in a MultiRNNCell using the RNN size `rnn_size`.
- Initializes Cell State using the MultiRNNCell's `zero_state` function

- The name "initial\_state" is applied to the initial state.
- The `get_init_cell` function return the cell and initial state in the following tuple (Cell, InitialState)

Good Work! You've

- Stacked 1 layer, for a more complex behaviour please try to include more layers(2 layers would be sufficient for this project). (Please have a look at <https://arxiv.org/abs/1506.02078>)
- Cell state initialized correctly.
- Used the identity function to name the state. (Good Work)

Awesome Job!! initiated cell correctly.

To know more about LSTM and RNN, please follow following links (explained very clearly)

<http://karpathy.github.io/2015/05/21/rnn-effectiveness/>  
<http://colah.github.io/posts/2015-08-Understanding-LSTMs/>



The function `get_embed` applies embedding to `input_data` and returns embedded sequence.



The function `build_rnn` does the following:

- Builds the RNN using the `tf.nn.dynamic_rnn`.
- Applies the name "final\_state" to the final state.
- Returns the outputs and final\_state state in the following tuple (Outputs, FinalState)



The `build_nn` function does the following in order:

- Apply embedding to `input_data` using `get_embed` function.
- Build RNN using cell using `build_rnn` 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)

Good Work!! Taken care off activation function in fully connected layer function.

More on embedding layer:

It might be necessary to add an Embedding layer just before the RNN to embed the words into a smaller dimensional space. Since the LSTM operates linearly over its input  $x$ , then linearly embedding the 1-of-K words to some embedding dimension  $D$  first corresponds to the original case you tried to get working here, **except** the matrix **is** factored through a rank  $D$ -dimensional bottleneck. If that makes sense.

Also, it is common to keep track of frequency of all words and discard words that appear, say, less than 5 times in the dataset.  
(Words from Andrej karpathy)



The `get_batches` function create batches of input and targets using `int_text`. The batches should be a Numpy array of tuples. Each tuple is (batch of input, batch of target).

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

Good work!! Good use of numpy library.

## Neural Network Training



- Enough epochs to get near a minimum in the training loss, no real upper limit on this. Just need to make sure the training loss is low and not improving much with more training.
- Batch size is large enough to train efficiently, but small enough to fit the data in memory. No real "best" value here, depends on GPU memory usually.
- Size of the RNN cells (number of units in the hidden layers) is large enough to fit the data well. Again, no real "best" value.
- The sequence length (`seq_length`) here should be about the size of the length of sentences you want to generate. Should match the structure of the data.  
The learning rate shouldn't be too large because the training algorithm won't converge. But needs to be large enough that training doesn't take forever.  
Set `show_every_n_batches` to the number of batches the neural network should print progress.

### Some Tips

The values of hyper parameters should be power of 2. Tensorflow optimizes our computation if we do so.

The `seq_length` should be kept similar to average size of sentence or say according to the structure of our data, this would help in getting better results. Link: <http://stats.stackexchange.com/questions/158834/what-is-a-feasible-sequence-length-for-an-rnn-to-model> (for this project 12-15 is a desirable range)

Select a learning rate such the model converges well with minimum oscillations/spikes (where the training loss mostly decreases and doesn't "jitter around" in value).

Some links that would help in fine tune model.

[http://neupy.com/2016/12/17/hyperparameter\\_optimization\\_for\\_neural\\_networks.html](http://neupy.com/2016/12/17/hyperparameter_optimization_for_neural_networks.html)



The project gets a loss less than 1.0

## Generate TV Script



"input:0", "initial\_state:0", "final\_state:0", and "probs:0" are all returned by `get_tensor_by_name`, in that order, and in a tuple



The `pick_word` function predicts the next word correctly.

Good Work!! correctly implemented this function without any bug.

Really appreciate the use of randomness when choosing the next word.

If you don't have used randomness, the predictions would have fall into a loop of the same words.



The generated script looks similar to the TV script in the dataset.

It doesn't have to be grammatically correct or make sense.

You have some mighty fine generated text here! If you are interested in doing another project about generating text, I encourage you to take a look at Project Gutenberg: <http://www.gutenberg.org>. They have thousands of free books that you can use to build a recurrent neural network with much more training data. You will also notice, as you use more training data, the quality of your generated text will improve!

To make script that make more sense. (More involved)

- We need a good amount of data and large network to learn it.
- You an follow text generation by GAN, this can also lead to good results.  
([https://www.reddit.com/r/MachineLearning/comments/40ldq6/generative\\_adversarial\\_networks\\_for\\_text/](https://www.reddit.com/r/MachineLearning/comments/40ldq6/generative_adversarial_networks_for_text/))

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