

Return to "Deep Learning" in the classroom

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Generate TV Scripts

	CODE REVIEW
	HISTORY
Meet	s Specifications
Good job	o overall generating your own Simpsons TV script using Recurrent Neural Networks!
ou put	a lot of effort into this project!
Congrati	ulations and good luck with your Nanodegree!
Requii	red Files and Tests
The pr	roject submission contains the project notebook, called "dlnd_tv_script_generation.ipynb".
	correct files are included!
All the	

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, LearingRate)

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)

Nice job stacking RNNs and initializing cell state.

To find out more about RNN effectiveness please follow this link:

http://karpathy.github.io/2015/05/21/rnn-effectiveness/

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

Good job applying embedding to input_data and returning embedded sequence.

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Check out this link:

https://www.tensorflow.org/api_docs/python/tf/contrib/layers/embed_sequence

to learn more about tf.contrib.layers.embed_sequence() which you could use to do the same.

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)

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]

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 in hatches to the number of hatches the neural network should brint progress

Good job - your parameters look reasonable!

For more info about hyperparameter optimization please go here

The project gets a loss less than 1.0

Congratulations on getting the loss to 0.361

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.

Great job using randomness when choosing the next word!

Your project predicts reasonable words!

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

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

Your generated script looks structurally similar to the TV script in the dataset! Congratulations!

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