

 Return to "Artificial Intelligence Nanodegree and Specializations" in the classroom

DISCUSS ON STUDENT HUB

DNN Speech Recognizer

REVIEW
CODE REVIEW
HISTORY

Meets Specifications

Dear Excellent Student,

Congratulations on meeting all the specifications for this project.

The hard work demonstrated in this submission is very commendable and shows a good understanding of the concepts from the lessons. Keep up the hard work and no obstacle will be above you. Have a great day celebrating your success and good luck in your future endeavours! ()

Pro Tips

Here are some resources that can help you learn further.

- Deep learning: from speech recognition to language and multimodal processing
- Dense LSTMs for Speech Recognition
- Machine Learning is Fun Part 6: How to do Speech Recognition with Deep Learning
- Deep neural network training for whispered speech recognition using small databases and generative model sampling

STEP 2: Model 0: RNN

The submission trained the model for at least 20 epochs, and none of the loss values in <code>model_0.pickle</code> are undefined. The trained weights for the model specified in <code>simple_rnn_model</code> are stored in <code>model_0.h5</code>.

Well done!

This submission trained the model with 20 epochs and none of the loss values in <code>model_0.pickle</code> are undefined. The weights for the <code>simple_rnn_model</code> are stored in <code>model_0.h5</code> in the results folder as required. We observe that this simple model doesn't train the data very well and thus the loss is very high. The validation loss for the last epoch is 754.7416.

STEP 2: Model 1: RNN + TimeDistributed Dense

The submission includes a sample_models.py file with a completed rnn_model module containing the correct architecture.

sample_models.py contains the rnn_model module with the correct architecture. Good job!

The submission trained the model for at least 20 epochs, and none of the loss values in <code>model_1.pickle</code> are undefined. The trained weights for the model specified in <code>rnn_model</code> are stored in <code>model_1.h5</code>.

Well done!

This submission trained the model with 20 epochs and none of the loss values in model_1.pickle are defined. The weights for the rnn_model are stored in model_1.h5 in the results folder as required. We can see that adding batch normalization and a time distributed layer improves the loss. The validation loss for the last epoch is 132.2031.

STEP 2: Model 2: CNN + RNN + TimeDistributed Dense

The submission includes a sample_models.py file with a completed cnn_rnn_model module containing the correct architecture.

sample_models.py contains the cnn_rnn_model module with the correct architecture. Good job!

The submission trained the model for at least 20 epochs, and none of the loss values in model_2.pickle are undefined. The trained weights for the model specified in cnn_rnn_model are stored in model_2.h5.

Well done!

This submission trained the model with 20 epochs and none of the loss values in <code>model_2.pickle</code> are undefined. The weights for the <code>cnn_rnn_model</code> are stored in <code>model_2.h5</code> in the results folder as required. These models are very powerful but have a tendency to severely overfit the data. You can add dropout to combat this. The validation loss for the last epoch is 128.8138.

The submission includes a sample_models.py file with a completed deep_rnn_model module containing the correct architecture.

sample_models.py contains the deep_rnn_model module with the correct architecture. Good job!

The submission trained the model for at least 20 epochs, and none of the loss values in <code>model_3.pickle</code> are undefined. The trained weights for the model specified in <code>deep_rnn_model</code> are stored in <code>model_3.h5</code>.

Well done!

This submission trained the model with 20 epochs and none of the loss values in <code>model_3.pickle</code> are undefined. The weights for the <code>deep_rnn_model</code> are stored in <code>model_3.h5</code> in the results folder as required. We can see that adding additional layers allows your network to capture more complex sequence representations, but this also makes it more prone to overfitting. As mentioned earlier, dropouts can help here greatly too. The validation loss for the last epoch is 130.0743.

STEP 2: Model 4: Bidirectional RNN + TimeDistributed Dense

The submission includes a sample_models.py file with a completed bidirectional_rnn_model module containing the correct architecture.

sample_models.py contains the bidirectional_rnn_model module with the correct architecture. Good job!

The submission trained the model for at least 20 epochs, and none of the loss values in model_4.pickle are undefined. The trained weights for the model specified in bidirectional_rnn_model are stored in model_4.h5.

Well done!

✓

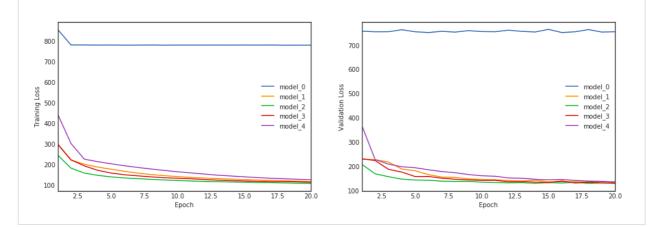
This submission trained the model with 20 epochs and none of the loss values in <code>model_4.pickle</code> are undefined. The weights for the <code>bidirectional_rnn_model</code> are stored in <code>model_4.h5</code> in the results folder as required. These models tend to converge quickly as they take advantage of future information through the forward and backward processing of data. The validation loss for the last epoch is 135.0070.

STEP 2: Compare the Models

The submission includes a detailed analysis of why different models might perform better than others.

Good job comparing the trained models, take particular observation of the trend of change in validation loss

that indicates whether the model overfits or not. It is also great to see that you have a good understanding of the models and recognize the fact that the deeper ones are able to detect more complex patterns in the data, giving a much better performance than the other models. Keep up the good work!



STEP 2: Final Model

The submission trained the model for at least 20 epochs, and none of the loss values in model_end.pickle are undefined. The trained weights for the model specified in final_model are stored in model_end.h5.

Well done!

This submission trained the model with 20 epochs and none of the loss values in model_end.pickle are undefined. The weights for the final_model are stored in model_end.h5 in the results folder as required. We can see that the loss decreases greatly to 105.2059.

The submission includes a sample_models.py file with a completed final_model module containing a final architecture that is not identical to any of the previous architectures.

sample_models.py contains the final_model module with a CNN + Bidirectional RNN + TimeDistributed

Dense architecture. Good job building this new architecture from components of the trained models above.

Nicely done!

The submission includes a detailed description of how the final model architecture was designed.

Excellent work here, presenting a detailed account of the final model architecture in your answer to question 2.

RETURN TO PATH

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