CSEE 5590 - Special Topics Deep Learning – Lab 2

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Configuration:

IDE: pycharm python: version 3

Introduction: By using the CNN model, we need to implement the text classification with the new data sets.

Objective: The main objective is to make us familiar with TensorFlow and by implementing the Test classification with the CNN model.

Approaches: To achieve Text classification with CNN Model, I used 5 new data sets. Then I have written the code that accepts the data sets after which I run it to get the accuracy and the loss. Then I checked the accuracy and the loss continuously by changing the hyper-Parameters, i.e., to know how these factors impact the loss and accuracy.

Workflow:

- To get the 5 different data sets.
- To write the code that accepts the data or the input data files.
- To train the system or model with the new data sets.
- Then we need to evaluate the model by checking the accuracy and loss.
- To modify the factors and check how it effects loss and accuracy.

Datasets: The datasets I used are as follows

fashion_7000, finance_7000, law_7000, lifestyle_7000 and test_data these are the data sets given in the class.

Parameters:

• Initially I have taken the embedding_dim to the default value of 128 and later I made it to 64 to check the accuracy and loss.

- The default num_filter is taken as 128 and then it is reduced to half to check how it effects the accuracy and loss.
- To evaluate Model on Dev set after number of steps, the default is 100. Also checked with 200 steps.

Evaluation:

I have modified the code of data_helpers.py by giving the new data sets and the below code accepts the new data.

```
👸 add.py × 🎁 Linear Regression.py × 🎼 ICP.py × 👸 lab1.py × 🎁 word2vec.py × 🎁 rnn_words.py × 🎁 train.py × 👸 eval.py × 👸 data_helpers.py ×
        def load_data_and_labels(data_file_1, data_file_2, data_file_3, data_file_4, data_file_5):
             Loads MR polarity data from files, splits the data into words and generates labels.
            Returns split sentences and labels.
            # Load data from files
            data_1 = list(open(data_file_1, "r", encoding='UTF8').readlines())
            data_1 = [s.strip() for s in data_1]
            data_2 = list(open(data_file_2, "r", encoding='UTF8').readlines())
            data_2 = [s.strip() for s in data_2]
            data_3 = list(open(data_file_3, "r", encoding='UTF8').readlines())
            data 3 = [s.strip() for s in data 3]
            data 4 = list(open(data file 4, "r", encoding='UTF8').readlines())
            data_4 = [s.strip() for s in data_4]
            data_5 = list(open(data_file_5, "r", encoding='UTF8').readlines())
            data_5 = [s.strip() for s in data_5]
            # Split by words
            x_text = data_1 + data_2 + data_3 + data_4 + data_5
            x text = [clean str(sent) for sent in x text]
            # Generate labels
            data_labels_1 = [[1, 0, 0, 0, 0] for _ in data_1]
            data_labels_2 = [[0, 1, 0, 0, 0] for _ in data_1]
data_labels_3 = [[0, 0, 1, 0, 0] for _ in data_1]
            data_labels_4 = [[0, 0, 0, 1, 0] for _ in data_1]
            data_labels_5 = [[0, 0, 0, 0, 1] for
                                                    in data 1]
            y = np.concatenate([data_labels_1, data_labels_2, data_labels_3, data_labels_4, data_labels_5], 0)
             return [x text, y]
```

The data loading parameters are given as below that takes 5 new data sets.

```
tf.flags.DEFINE_string("fashion1_7000", "./ICP_data/fashion_7000.txt",
    "fashion_7000.txt")
tf.flags.DEFINE_string("finance2_7000", "./ICP_data/finance_7000.txt",
    "finance_7000.txt")
tf.flags.DEFINE_string("law3_7000", "./ICP_data/law_7000.txt", "law_7000.txt")
tf.flags.DEFINE_string("lifestyle4_7000", "./ICP_data/lifestyle_7000.txt",
    "lifestyle_7000.txt")
tf.flags.DEFINE_string("TestData5", "./ICP_data/TestData", "TestData")
```

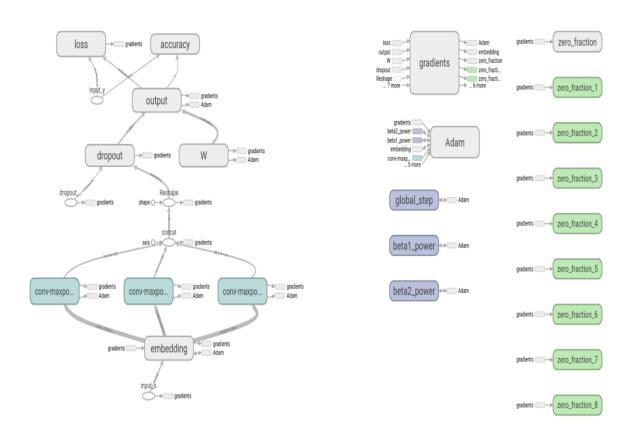
The data is then loaded or passed to the data helpers class as shown below.

The data is loaded here as shown below in the eval.py in order to evaluate the result or check the loss and accuracy.

Output:

Here is the output graph from the Tensor Flow.

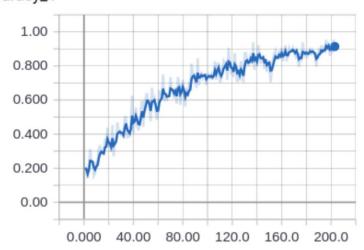
Main Graph Auxiliary Nodes



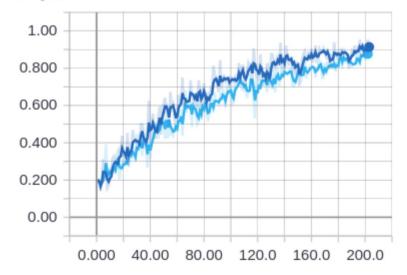
Accuracy Graph:

The accuracy for the default values that are provided.

accuracy_1

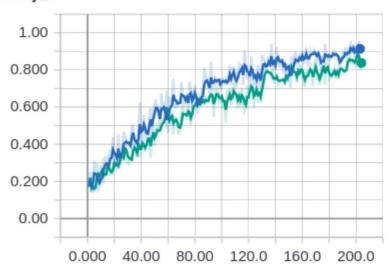


The accuracy when the embedding_dim is reduced to half i.e., 64 accuracy_1



The accuracy when the num_filters is reduced to half of the default value.

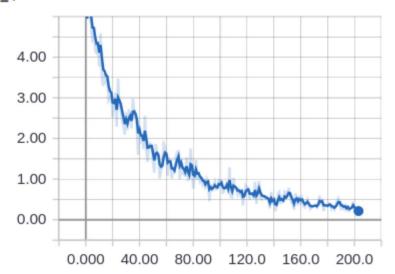
accuracy_1



Loss Graph:

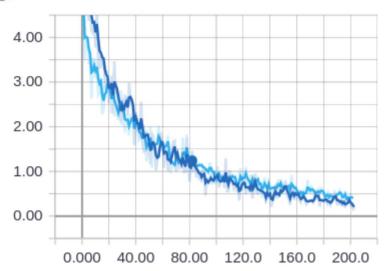
The loss graph with the default values provided for the hyper-parameters.

loss_1



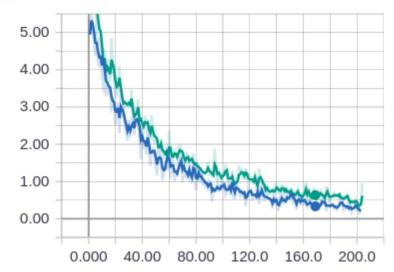
The loss when embedding_dim is reduced to half.

loss_1



The loss when num_filters is reduced to half when compared to the default value.

loss_1



Conclusion:

The loss reduces and the accuracy increases as we increase the number of iterations. By more iterations the system or the model will be more accurate.