DEEP LEARNING LAB ASSIGNMENT – 1

Introduction: The given task is to implement Logistic Regression using Tensorflow and display the graph on Tensorboard using Python. Firstly Logistic Regression is similar to the linear regression but we use logistic regression for categorical data. Logistic Regression will be best fit if the predictors are either Qualitative or Quantitative and response is Qualitative. Logistic regression produces linear boundaries similar to the linear regression. No assumption on the predictors is made by logistic regression.

Objectives: In this task, we implement the logistic regression on the MNIST dataset which is a collection of the 28X28 pixel bounding black and white images. We implement the logistic regression on this dataset using tensorflow and write the graph on tensorboard using python. We change the hyperparameters to check the accuracy and costs of the model.

Approaches:

To implement the logistic regression to our dataset, i have chosen softmax function. We used softmax to build our model as we have multiple classes in our data. Softmax is a generalized version of logisite regression used for multi-class classification. Then we use the reduce_mean function to obtain the cost for each epochs for the whole batch

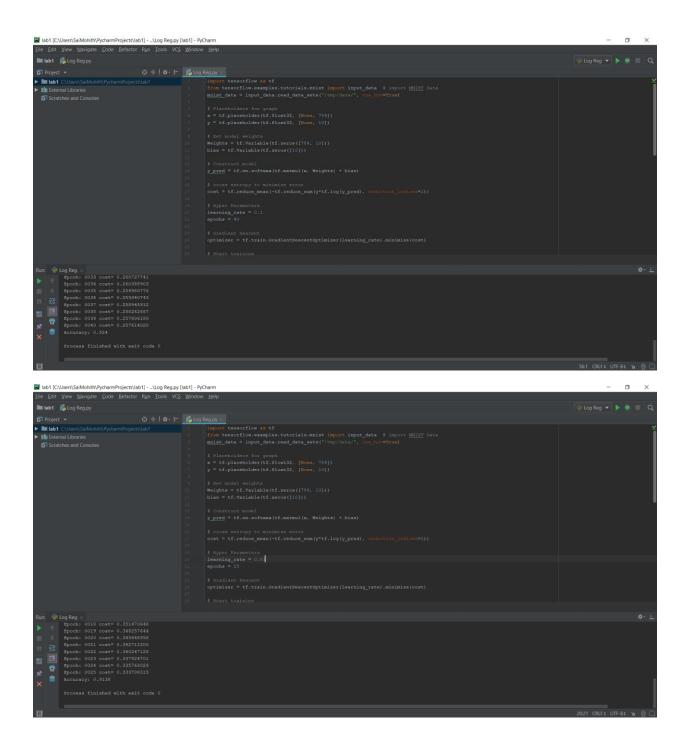
Workflow:

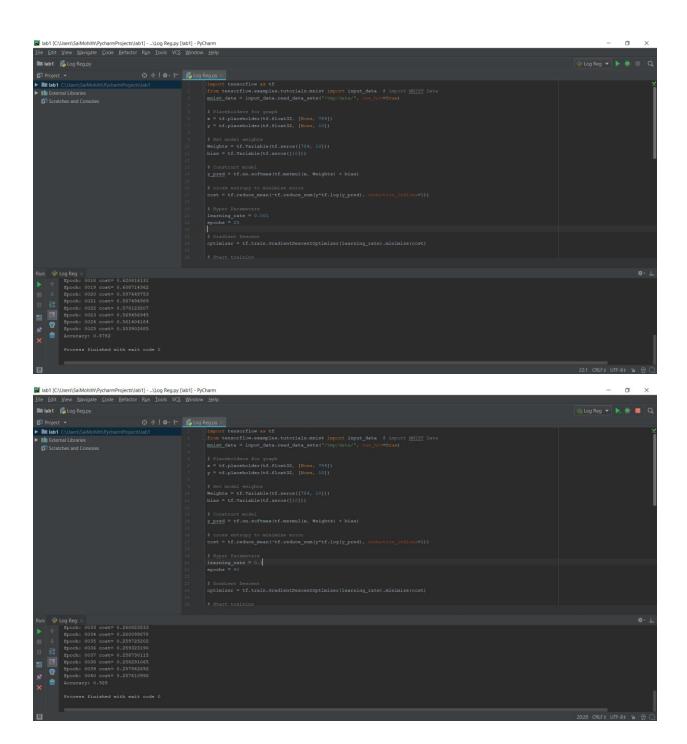
- 1. First import the data from the tensorflow.examples.tutorials.mnist and read using read_input_data function.
- 2. We then use the placeholders for the graph on the tensorboard.
- 3. Construct our model using softmax function.
- 4. We use Cross entropy to minimize the error.

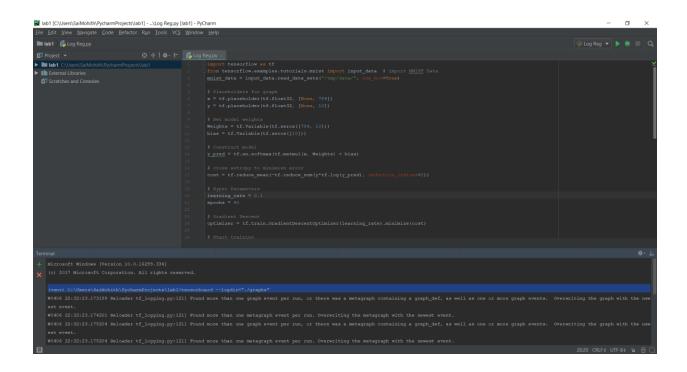
5. Declare the hyper parameters Learning rate and Epochs.
6. Using the Gradient Descent Optimizer to optimize the cost obtained.
7. We then train our model. In this section we use the epochs also can be referred as iterations. We create a batch and train our model based on that batch for iterations of the size of epochs.
8. We write the graph in the graph directory using the tf.summary.filewriter fucntion to get the graph from tensorboard.
9. Then we test the data using the tf.equal function and compare the real output with the predicted output. We use Accuracy function to get the accuracy of the obtained output in the final.
Dataset:
MNIST Dataset is a 28X28 pixel bounding box with black and white images. It is normalize data from the NSIT database. It has 60,000 training and 10,000 testing images. It has the gray scale images. It is widely used for training and testing purposes in the Machine Learning.
Parameters:
We used softmax in the data to construct our model. We get accuracy using the accuracy function. Batch_size is chosen low to get the cost minimized. Overfitting of the data should be avoided. We use one_hot for converting the binary classes to the categorical values.
Evaluation:
Code:
import tensorflow as tf

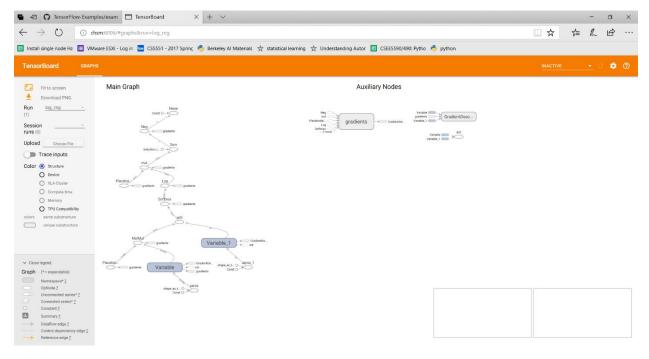
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from tensorflow.examples.tutorials.mnist import input data # import MNIST Data
  mnist_data = input_data.read_data_sets("/tmp/data/", one_hot=True)
  # Placeholders for graph
  x = tf.placeholder(tf.float32, [None, 784])
  y = tf.placeholder(tf.float32, [None, 10])
  # Set model weights
  Weights = tf.Variable(tf.zeros([784, 10]))
  bias = tf.Variable(tf.zeros([10]))
  # Construct model
  y pred = tf.nn.softmax(tf.matmul(x, Weights) + bias)
  # cross entropy to minimize error
  cost = tf.reduce_mean(-tf.reduce_sum(y*tf.log(y_pred), reduction_indices=1))
# Hyper Parameters
learning_rate = 0.1
epochs = 40
# Gradient Descent
optimizer = tf.train.GradientDescentOptimizer(learning_rate).minimize(cost)
# Start training
with tf.Session() as sess:
  # Run the Global Initializer to initialize all the values
```

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sess.run(tf.global variables initializer())
# Write the graph in the log reg directory
writer = tf.summary.FileWriter('./graphs/log reg', sess.graph)
# Start training cycle
for epoch in range(epochs):
  AverageCost = 0.
  Batch_tot = int(mnist_data.train.num_examples/100)
  # This loops over all batches
  for i in range(Batch tot):
    batch xs, batch ys = mnist data.train.next batch(batch size=100)
    # Run optimization op (backprop) and cost op (to get loss value)
    _, c = sess.run([optimizer, cost], feed_dict={x: batch_xs, y: batch_ys})
    # Compute average loss
    AverageCost += c / Batch tot
  # Display logs per epoch step
  if (epoch+1) \% 1 == 0:
    print("Epoch:", '%04d' % (epoch+1), "cost=", "{:.9f}".format(AverageCost))
writer.close()
w, b = sess.run([Weights, bias])
# Run the test model
correct_prediction = tf.equal(tf.argmax(y_pred, 1), tf.argmax(y, 1))
# Calculate accuracy
accu = tf.reduce mean(tf.cast(correct prediction, tf.float32))
print("Accuracy:", accu.eval({x: mnist data.test.images, y: mnist data.test.labels}))
```









First one shows the accuracy of 92.4 when learning rate is 0.1 but as we decrease the learning rate, we see the accuracy drops to 91.3 and 87.82. We see the accuracy is high when the learning rate is high and epochs are high. The accuracy falls down as the learning rate and epochs decrease. Learning rate have the major impact on the accuracy.

Therefore the accuracy is seems to be 92.4% when the learning rate is 0.1. This seems to be a good fit the data.
References:
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https://github.com/niyamatalmass/Logistic-regression-using-iris-and-mnist-data-set/blob/master/mnist_logistic_regression.ipynb
https://www.tensorflow.org/programmers_guide/summaries_and_tensorboard
https://www.tensorflow.org/api_docs/python/tf/nn/softmax

Conclusion: