NEURAL NETWORK AND FUZZY LOGIC

ASSIGNMENT-3

PROBLEM 1

```
import numpy as np
   from scipy.io import loadmat
  import tensorflow as tf
4 from matplotlib import pyplot as plt
  m = n = 1000
   def to one hot(Y):
       mm = Y.shape[0]
       Y one hot = np.zeros((mm, 2))
       for i in range(mm):
           Y one hot[i, Y[i]] = 1
       return Y one hot
   def normalize features(X train, X test):
       mu = np.mean(X train, axis=0, keepdims=True)
       sigma = np.std(X train, axis=0, keepdims=True)
       X train = (X train - mu) / sigma
       X test = (X test - mu) / sigma
  def import data():
       """Import data files and divide into training and test sets."""
       X = loadmat('data for cnn.mat')['ecq in window']
       Y = loadmat('class label.mat')['label']
       random indices = np.random.permutation(m)
       X = X[random indices]
       Y = Y[random indices]
       train set size = 768
       test set size = m - train set size
       X train, X test = X[0:train set size], X[train set size:]
       Y train, Y test = Y[0:train set size], Y[train set size:]
       X train, X test = normalize features(X train, X test)
       Y train = to one hot(Y train)
       Y test = to one hot(Y test)
       return X train, X test, Y train, Y test
```

```
def convld(x, W, b, s=1):
    """Conv1D wrapper, with bias and relu activation"""
    x = tf.nn.convld(x, W, stride=s, padding='SAME')
    x = tf.nn.bias add(x, b)
    return tf.nn.relu(x)
def maxpool1d(x, k=2):
    """MaxPool1D wrapper"""
    return tf.layers.max poolingld(x, pool size=k, strides=k, padding='SAME')
def conv_net(x, weights, biases, dropout):
    x = tf.reshape(x, shape=[-1, m, 1])
    convl = convld(x, weights['wcl'], biases['bcl'])
    conv1 = maxpoolld(conv1, k=2)
    fcl = tf.reshape(conv1, [-1, weights['wd1'].get shape().as list()[0]])
    fcl = tf.add(tf.matmul(fcl, weights['wdl']), biases['bdl'])
    fc1 = tf.nn.relu(fc1)
    fc2 = tf.add(tf.matmul(fc1, weights['wd2']), biases['bd2'])
    fc2 = tf.nn.relu(fc2)
    fc2 = tf.nn.dropout(fc2, dropout)
    out = tf.add(tf.matmul(fc2, weights['out']), biases['out'])
    return out
    X train, X test, Y train, Y test = import data()
```

```
if name == '_main__':
    X_train, X_test, Y_train, Y_test = import_data()
    num steps = 400
    learning rate = 0.00001
    batch size = 64
    display step = 10
    num input = 1000
    num classes = 2
    dropout = 0.70
    X = tf.placeholder(tf.float32, [None, num input])
    Y = tf.placeholder(tf.float32, [None, num classes])
    keep prob = tf.placeholder(tf.float32)
    weights = {
    'wcl': tf.Variable(tf.random_normal([ , 1, 64])),
    'wd1': tf.Variable(tf.random normal([ 00*64, 1024])),
    'wd2': tf.Variable(tf.random normal([1024, 20])),
    'out': tf.Variable(tf.random_normal([20, num_classes]))
    biases = {
    'bcl': tf.Variable(tf.random normal([64])),
    'bd1': tf.Variable(tf.random normal([1024])),
    'bd2': tf.Variable(tf.random_normal([20])),
    'out': tf.Variable(tf.random normal([num classes]))
    logits = conv net(X, weights, biases, keep prob)
```

```
prediction = tf.nm.softmax(logits)

# Define loss and optimizer
loss.op = tf.reduce_mean(tf.mn.softmax_cross_entropy_with_logits(logits=logits, labels=Y))
optimizer = tf.train.AdamOptimizer(learning_rate=learning_rate)

train_op = optimizer.minimize(loss_op)

# Evaluate_model
correct_pred = tf.equal(tf.argmax(prediction, 1), tf.argmax(Y, 1))
accuracy = tf.reduce_mean(tf.cast(correct_pred, tf.float32))

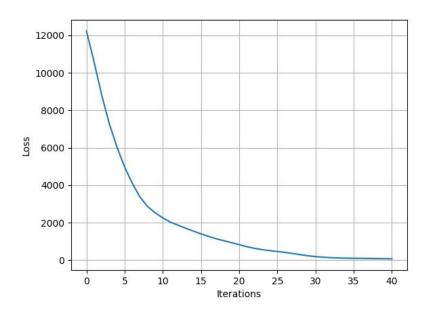
# Initialize the variables (i.e. assign their default value)
init = tf.global_variables_initializer()

losses = []

# Start training
with tf.Session() as sess:

sess.run(init)
batch_start = 0
for step in range(1, num_steps+1):
batch_start = 0
for step in range(1, num_steps+1):
batch_start = (batch_start + batch_size) % batch_size
sess.run(train_op, feed_dict=(X: batch_x, Y: batch_y, X: batch_y, X: batch_y, Y: batch_y, X: batch_y, Y: batch_y, X: batch_y, Y: batch_y, X: batch_y, Y: batch_y, Y: batch_y, X: batch_y, Y: batch_y, K: aff'.format(loss) + ", Training Accuracy" + \
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Training Plot:



Test Evaluation:

Accuracy = 87.92%