## Lesson 4

#### Code

```
import tensorflow as tf
from sklearn.model selection import train test split
import numpy as np
from sklearn import datasets
from sklearn import preprocessing
import pandas as pd
import sys
class Graph:
    def init__(self):
        tf.reset default graph()
        tf.set random seed(self.random seed())
    def cost calc(self):
        print("Need to override calc cost")
    def activation calc(self):
        print("Need activaation calc")
    def variable initializer(self):
        print("Missing variable_initializer")
    def calc accuracy(self, sess, features, labels):
        print("Missing calc accuracy")
    def random seed(self):
        return 42
```

```
def initialize(self, features, labels, hidden):
    self.features = features
    self.labels = labels
    self.hidden = hidden
    input shape = self.features.shape[1]
    out shape = self.labels.shape[1]
    self.inputs = tf.placeholder(
            tf.float32,
            shape=[None, input shape])
    self.output = tf.placeholder(
            tf.float32,
            shape=[None, out shape])
    self.test outputss = tf.placeholder(
            tf.float32,
            shape=[None, out shape])
    self.learning rate = tf.placeholder(tf.float32)
    self.w1 = tf.get variable(name="w1",
            shape=[input shape, hidden],
            initializer=self.variable initializer()())
    self.b1 = tf.get variable(name="b1",
            shape=[hidden],
            initializer=tf.constant initializer(0.0))
    self.h1 = self.activation calc()(
            tf.matmul(self.inputs, self.w1) + self.b1)
    self.w2 = tf.get variable(name="w2",
            shape=[hidden, out shape],
            initializer=self.variable initializer()())
    self.b2 = tf.get variable(name="b2",
            shape=[out shape],
            initializer=tf.constant initializer(0.0))
    self.o1 = self.activation calc()(
            tf.matmul(self.h1, self.w2) + self.b2)
    correct prediction = tf.equal(
            tf.argmax(self.o1,axis=1),
            tf.argmax(self.test outputss,axis=1))
    self.accuracy = tf.reduce mean(
            tf.cast(correct prediction, tf.float32))
    self.cost = self.cost calc()(self.output, self.o1)
```

```
self.updates = tf.train.GradientDescentOptimizer(
                self.learning rate).minimize(self.cost)
    def train(self, tsize, epochs, lr):
        train features, test features, train labels, test labels =
train test split(
             self.features,
             self.labels,
             test size=tsize ,
             random state=self.random seed())
        with tf.Session() as sess:
            init = tf.global variables initializer()
            sess.run(init)
            for epoch in range(epochs+1):
                # Train with each example
                for i in range(len(train features)):
                    op,cst = sess.run([self.updates,self.cost],
                             feed dict={
                                 self.inputs: train features[i: i + 1],
                                 self.output: train labels[i: i + 1],
                                 self.learning rate:lr})
                if (epoch % (epochs/20)) == 0:
                    test accuracy = self.calc accuracy(
                             sess, test features,
                             test labels)
                    print("Epoch: %d, accuracy: %.5f, cost: %.5f" %
                             (epoch, test accuracy, cst))
            print("Weights Level 1:\n", sess.run(self.w1))
            print("Bias Level 1:\n", sess.run(self.b1))
            print("Weights Level 2:\n", sess.run(self.w2))
            print("Bias Level 2:\n", sess.run(self.b2))
            return test accuracy,cst
class Graph1 (Graph):
    def cost calc(self):
        def calculator(target,compValue):
            return tf.reduce mean(-target*tf.log(compValue) -
                     (1-target) *tf.log(1-compValue))
        return calculator
    def activation calc(self):
```

```
return tf.sigmoid
    def variable initializer(self):
        return tf.contrib.layers.xavier initializer
    def calc accuracy(self, sess, features, labels):
        return sess.run(
                self.accuracy,
                feed dict={
                    self.inputs:features,
                    self.test outputss:labels})
class Graph2 (Graph):
    def cost calc(self):
        def calculator(target,compValue):
            return tf.reduce mean(
                    tf.pow(target - compValue,2))
        return calculator
    def activation calc(self):
        return tf.nn.relu
    def variable initializer(self):
        return tf.initializers.random uniform
    def calc accuracy(self, sess, features, labels):
        xx1 = np.stack(labels,axis=0)
        xx2 = np.stack(
                sess.run(self.ol,
                    feed dict={
                         self.inputs:features,
                         self.test outputss:labels}),axis=0)
        return sess.run(
                tf.sqrt(
                    tf.reduce mean (
                         tf.square(
                             tf.subtract(xx1,xx2))))
def scale(t):
    tMin = t.min(axis=0)
    tMax = t.max(axis=0)
    return (t-tMin) / (tMax-tMin)
def problem1(epochs, learning rate, hidden):
    f = open("Admissions.csv")
```

```
f.readline()
    dataset = np.genfromtxt(fname = f, delimiter = ',')
    features = dataset[:,1:] # antecedents
    features scaled = scale(features)
    labels = dataset[:,0:1] # consequent
    one hot = np.zeros(shape=(len(labels),2))
    for i in range(0,len(labels)):
        one hot[i, int(labels[i])] = 1
    g = Graph1()
    g.initialize(features scaled, one hot, hidden)
    return g.train(0.30,epochs,learning rate)
def problem2(epochs,learning rate,hidden):
    f = open("Advertising.csv")
    f.readline()
    dataset = np.genfromtxt(fname = f, delimiter = ',')
    features = dataset[:,1:4] # antecedents
    features scaled = scale(features)
    labels = dataset[:,4:] # consequent
    labels scaled = scale(labels)
    g = Graph2()
    g.initialize(features scaled, labels scaled, hidden)
    return g.train(0.30,epochs,learning rate)
def main():
   print("Problem 1")
   print("======")
   problem1 (375, 0.1, 9)
   print("Problem 2")
   print("======")
   print("Accuracy is via RMSE")
   problem2 (500, 0.01, 5)
if __name__ == "__main__":
   main()
```

### Results

Run > python lesson3.py

# Problem 1

```
Epoch: 0, accuracy: 0.67500, cost: 0.36755
Epoch: 75, accuracy: 0.73333, cost: 0.11847
Epoch: 150, accuracy: 0.75000, cost: 0.11610
Epoch: 225, accuracy: 0.75000, cost: 0.11407
Epoch: 300, accuracy: 0.75000, cost: 0.11176
Epoch: 375, accuracy: 0.75000, cost: 0.11009
Weights Level 1:
-1.0863187
 -0.44966298 -3.2398062 -1.6787095
[-2.2877874 -0.10465505 -0.60933286 1.7770392 2.6080217
1.8389196
  0.84129864 -2.1970472 -1.8251772 |
[1.5223188 -1.5600401 -1.6974384 -1.916973 -3.2841446]
-1.8059342
 -1.7172989 1.4502791 1.2959756 ]]
Bias Level 1:
[-0.2882644 -1.414716 -1.4287888 -2.0827897 -1.8387002]
-1.741329
-1.6675537 0.19879904 -0.83083284]
Weights Level 2:
[[ 1.6850691 -1.2131442 ]
 [-0.47689864 \quad 0.21940975]
 [-0.55776393 0.91958785]
 [-1.1342436 0.9533675]
 [-0.61192983 0.9399046]
[-0.9596164 0.05750658]
[ 1.711513 -2.077902 ]
 [ 1.031563 -1.2603813 ]]
Bias Level 2:
 [ 0.9256766 -0.93577576]
```

```
Problem 2
=======
Accuracy is via RMSE
Epoch: 0, accuracy: 0.27416, cost: 0.04632
Epoch: 25, accuracy: 0.07419, cost: 0.00622
Epoch: 50, accuracy: 0.06980, cost: 0.00545
Epoch: 75, accuracy: 0.06594, cost: 0.00475
Epoch: 100, accuracy: 0.06244, cost: 0.00422
Epoch: 125, accuracy: 0.05942, cost: 0.00382
Epoch: 150, accuracy: 0.05682, cost: 0.00328
Epoch: 175, accuracy: 0.05418, cost: 0.00243
Epoch: 200, accuracy: 0.05206, cost: 0.00169
Epoch: 225, accuracy: 0.04999, cost: 0.00121
Epoch: 250, accuracy: 0.04825, cost: 0.00098
Epoch: 275, accuracy: 0.04659, cost: 0.00080
Epoch: 300, accuracy: 0.04518, cost: 0.00064
Epoch: 325, accuracy: 0.04409, cost: 0.00055
Epoch: 350, accuracy: 0.04278, cost: 0.00047
Epoch: 375, accuracy: 0.04168, cost: 0.00043
Epoch: 400, accuracy: 0.04064, cost: 0.00039
Epoch: 425, accuracy: 0.03966, cost: 0.00036
Epoch: 450, accuracy: 0.03885, cost: 0.00031
Epoch: 475, accuracy: 0.03819, cost: 0.00031
Epoch: 500, accuracy: 0.03744, cost: 0.00032
Weights Level 1:
 [0.48358107 \quad 0.37768707 \quad 0.96805984 \quad 0.8175911
                                                  0.865153131
             0.4416568 -0.09218649 0.5522364 -0.385495931
 [-0.117202]
 Bias Level 1:
 [-0.46927968 -0.3457328 \quad 0.04907916 -0.08450499 -0.24562386]
Weights Level 2:
 [[ 0.7695015 ]
 [ 0.57346714]
 [ 0.4843294 ]
 [ 0.28136104]
 [-0.5615065]
```

Bias Level 2:

#### [0.07882915]

```
[[0.0, 0.0, 0, 0.04137863, 0.0017121908934420033], [1.0, 0.0, 1, 0.97319275, 0.000718628577093483], [0.0, 1.0, 1, 0.9920135, 6.378395795891834e-05], [1.0, 1.0, 0, 0.017914694, 0.00032093625404074066]]
```