COEN 240 Machine Learning

Homework #1

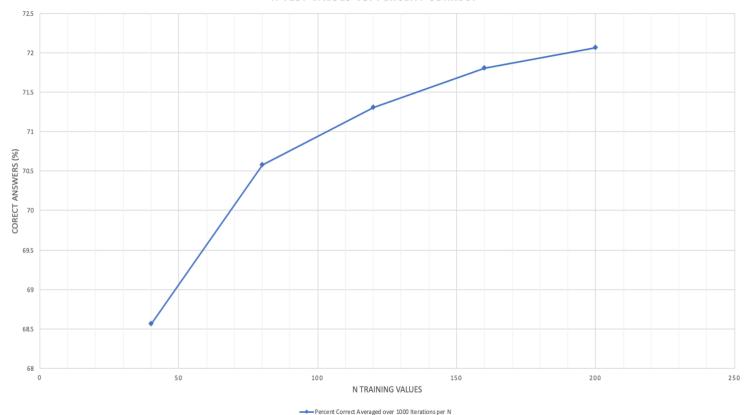
Name: Alex Cherekdjian Student ID: 00001083236

Problem 1

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& \overset{-}{$$

Problem 2

N TEST VALUES VS. PERCENT CORRECT



Comment: Through plotting all these points and averaging all 1000 iterations of each n, we can see that as the number of training values increases, the accuracy rating of the algorithm goes up slightly with a maximum increase from n=40 to n=200 of around four percent.

Results:

n=40, 68.566

n=80, 70.583

n=120, 71.311

n=160, 71.807

n=200, 72.069

Attachment

Problem 2 Code (in zip file):

```
import tensorflow.compat.v1 as tf

tf.disable_v2_behavior()

import pandas as pd
```

```
# Print iterations progress
def printProgressBar (iteration, total, prefix = '', suffix = '', decimals =
1, length = 100, fill = ''', printEnd = "\r"):
    Call in a loop to create terminal progress bar
    @params:
       iteration - Required : current iteration (Int)
       total - Required : total iterations (Int)
       prefix
                  - Optional : prefix string (Str)
       suffix - Optional : suffix string (Str)
       decimals
                  - Optional : positive number of decimals in percent
complete (Int)
       length
                  - Optional : character length of bar (Int)
                  - Optional : bar fill character (Str)
       printEnd - Optional : end character (e.g. "\r", "\r\n") (Str)
   percent = ("{0:." + str(decimals) + "f}").format(100 * (iteration /
float(total)))
    filledLength = int(length * iteration // total)
   bar = fill * filledLength + '-' * (length - filledLength)
    print('\r%s |%s| %s%% %s' % (prefix, bar, percent, suffix), end =
printEnd)
    # Print New Line on Complete
    if iteration == total:
       print()
# importing data into pandas
diabetes data = pd.read csv('pima-indians-diabetes.csv', \
                           names=['Pregnancies', 'Glucose', 'Blood
Pressure', 'Skin Thickness', \
                                  'Insulin', 'BMI', 'Diabetes Pedigree
Function', 'Age', 'Outcome'])
# getting all positive and negative rows for test selection
outcome positive = diabetes data[diabetes data['Outcome'] == 1]
outcome negative = diabetes data[diabetes data['Outcome'] == 0]
N = 768 \# number of samples
m = 8 \# number of attributes
n train test values = [40, 80, 120, 160, 200]
percent total list = [] # list of averaged totals from 1000 iterations of
each 40, 80, 120, 160, 200
n test total = 0 # total number of test questions for the 1000 iterations
correct total = 0 # total number of correct answers for 1000 iterations
```

```
printProgressBar(0, 1000, prefix = 'Progress:', suffix = 'Complete', length =
50)
for value in n train test_values:
    n train test = value
   print("n value: " + str(n_train_test) + "\n")
    n test total = 0 # init to zero
   correct total = 0 # init to zero
    for i in range (0,1000):
       # getting training data
       X train = outcome negative.sample(n=n train test) # get n train test
negative rows
       positive = outcome positive.sample(n=n train test) # get n train test
positive rows
       X train = X train.append(positive, ignore index = True) # append
positive rows to negative rows
        X train = X train.sample(frac=1) # shuffle training rows
        # getting testing data
        ans = pd.merge(diabetes data, X train, how='outer', indicator=True) #
get the rest of the data as training data
       X test = ans[ans[' merge'] == 'left only'] # get all values not in
both sets
        del X test[' merge'] # remove last column
        # assign t values to outcomes of X
        t train = pd.DataFrame(X train['Outcome'], columns=['Outcome'])
        t test = pd.DataFrame(X test['Outcome'], columns=['Outcome'])
        # creating list of correct answers to allow for interation when
checking for correct answers
        t test list = X test['Outcome']
        del X test['Outcome'] # remove last column
        del X train['Outcome'] # remove last column
        n train,m = X train.shape
        n \text{ test,m} = X \text{ test.shape}
       # define the tensors
```

```
X = tf.placeholder(tf.float64, shape=(None, m), name='X') # input
features vector
       t = tf.placeholder(tf.float64, shape=(None, 1), name='t') # target
values
       n = tf.placeholder(tf.float64, name='n') # number of samples
       XT = tf.transpose(X)
       w = tf.matmul(tf.matmul(tf.matrix inverse(tf.matmul(XT,X)), XT), t) #
w = inv(X'*X)*X'*t
        # predicted value
       y = tf.matmul(X, w)
        # mean squared error of the prediction training set
       MSE = tf.div(tf.matmul(tf.transpose(y-t), y-t), n)
       w star = tf.placeholder(tf.float64, shape=(m, 1), name='w star')
       y test = tf.matmul(X, w star)
       with tf.Session() as sess:
            # running tensorflow sessions
            MSE train val, w val = \
            sess.run([MSE, w], feed dict={X : X train, t : t train, n :
n train})
            y test val = \
            sess.run([y test], feed dict={X : X test, t : t test, n : n test,
w star : w val})
            correct = 0
            # testing all the values
            for prediction, actual in zip(y_test_val[0], t_test_list):
                if (prediction[0] >= 0.5 and actual == 1):
                    correct+=1
                elif (prediction[0] < 0.5 and actual == 0):</pre>
                    correct.+=1
        # update progress bar on terminal screen
       printProgressBar(i+1, 1000, prefix = 'Progress:', suffix =
'Complete', length = 50)
        # increment total values tested and correct for n iteration
        n test total += n test
```

```
correct_total += correct

# calculate total for n percent correct

percent_correct = (correct_total/n_test_total)*100

percent_total_list.append(percent_correct)

# print final list

print(percent_total_list)
```