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1  import pickle
2
3  import matplotlib.pyplot as plt
4  import numpy as np
5
6  from functions import * # functions provided by the course for the lab
7
8
9  # ----- test class for gradient testing (EXTRACT TO SEPARATE FILE!)
10 import unittest
11
12
13 class TestModel(unittest.TestCase):
14
15     def setUp(self):
16         self.train = np.array(separate(loadBatch('data_batch_1'), 10, 10000))
17         self.W = np.random.normal(0, 0.01, (10, 3072))
18         self.b = np.random.normal(0, 0.01, (10, 1))
19         self.lmda = 0
20         self.P = evaluateClassifier(self.train[0], self.W, self.b)
21         dp = 10
22         self.gradW_n, self.gradB_n = computeGradsNumSlow(
23             self.train[0][:, :dp], self.train[1][:, :dp], self.W, self.b, self.lmda, 10**-6)
24         self.gradW_a, self.gradB_a = computeGradients(
25             self.P[:, :dp], self.train[0][:, :dp], self.train[1][:, :dp], self.W, self.lmda, dp)
26         self.checkW, self.checkB = gradientCheck(self.gradW_a, self.gradW_n, self.gradB_a, self.gradB_n, 10**-6)
27
28     def test_gradientMean(self):
29         self.assertAlmostEqual(self.gradW_n.mean(), self.gradW_a.mean(), places=7)
30         self.assertAlmostEqual(self.gradB_n.mean(), self.gradB_a.mean(), places=7)
31
32     def test_relError(self):
33         self.assertLessEqual(np.max(self.checkW), 10**-6)
34         self.assertLessEqual(self.checkW.mean(), 10**-6)
35         self.assertLessEqual(np.max(self.checkB), 10**-6)
36         self.assertLessEqual(self.checkB.mean(), 10**-6)
37
38 # ----- END TEST CLASS
39
40
41 def plotGraph(lst1, lst2, rangeX, yLabel, xLabel, lst1Label, lst2Label):
42     plt.figure()
43     plt.plot(rangeX, lst1, label=lst1Label)
44     plt.plot(rangeX, lst2, label=lst2Label)
45     plt.xlabel(xLabel)
46     plt.ylabel(yLabel)
47     plt.legend()
48
49
50 def separate(data, K):
51     """
52     Takes in dataset and seperates
53     returns: data X (dim x N), one-hot label matrix Y (KxN), labels(1xN)
54     """
55     X = np.array(data.get(b'data'), dtype=float).T
56     labels = np.array([data.get(b'labels')])
57     Y = np.zeros((K, X.shape[1]))
58     Y[labels, np.arange(labels.size)] = 1
59     return X, Y, labels
60
61
62 def normalize(X, mean, std):
63     return (X - mean) / std
64
65
66 def evaluateClassifier(X, W, b):
67     """
68     Outputs P = softmax(Wx + b) as KxDim-matrix,
69     where each column is sums to 1
70     """
71     return softmax(np.matmul(W, X) + b)
72
73
74 def computeCost(X, Y, W, b, lmda):
75     """ computes cost of loss for the network """
76     P = evaluateClassifier(X, W, b)
77     J = ((1 / np.size(X, 1)) * -np.sum(Y*np.log(P))) + (lmda * np.sum(np.square(W)))
78     return J, P
79
80
81 def computeAccuracy(P, y):
82     """ Accuracy defined as correctly classified of total datapoints """
83     P_max = np.array([np.argmax(P, axis=0)])
84     return np.array(np.where(P_max == np.array(y))).shape[1] / np.size(y)
85
86
87 def computeGradients(P, X, Y, W, lmda, bsize):
88     """ Computes gradients using chain rule """

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89     G = -(Y - P)
90     grad_W = (1 / bsize) * np.matmul(G, np.array(X).T) + 2*lmlda*W
91     grad_b = np.array((1 / bsize) * np.matmul(G, np.ones(bsize))).reshape(np.size(W, 0), 1)
92     return [grad_W, grad_b]
93
94
95 def gradientCheck(gradW_a, gradW_n, gradB_a, gradB_n, eps):
96     """ computes the relative error between analytical and numerical gradient calcs """
97
98     def check(grad_a, grad_n, eps):
99         diff = np.absolute(np.subtract(grad_a, grad_n))
100         thresh = np.full(diff.shape, eps)
101         summ = np.add(np.absolute(grad_a), np.absolute(grad_n))
102         denom = np.maximum(thresh, summ)
103         return np.divide(diff, denom)
104
105     resW = check(gradW_a, gradW_n, eps)
106     resB = check(gradB_a, gradB_n, eps)
107     return resW, resB
108
109
110 def updateParameters(W, b, grad_W, grad_b, eta):
111     W = W - eta * grad_W
112     b = b - eta * grad_b
113     return W, b
114
115
116 def miniBatch(X, Y, y, W, b, lmlda, bsize, eta):
117     """ bsize'ed batches evaluated """
118     for i in range(int(np.size(X, 1)/bsize)):
119         n = i*bsize
120         P = evaluateClassifier(X[:, n:n+bsize], W, b)
121         grad = computeGradients(P, X[:, n:n+bsize], Y[:, n:n+bsize], W, lmlda, bsize)
122         W, b = updateParameters(W, b, grad[0], grad[1], eta)
123     return W, b
124
125
126 def main():
127     """ loading of data, initilisation of parameters and main script """
128
129     K = 10 # num of classes
130
131     # load data
132     train = loadBatch('data_batch_1')
133     validation = loadBatch('data_batch_2')
134     test = loadBatch('data_batch_3')
135
136     # separate data
137     trainX, trainY, train_y = separate(train, K)
138     valX, valY, val_y = separate(validation, K)
139     testX, testY, test_y = separate(test, K)
140
141     # pre-process data
142     trainXmean = np.array([np.mean(trainX, 1)]).T
143     trainXstd = np.array([np.std(trainX, 1)]).T
144     trainX = normalize(trainX, trainXmean, trainXstd)
145     valX = normalize(valX, trainXmean, trainXstd)
146     testX = normalize(testX, trainXmean, trainXstd)
147
148     # initialize parameters
149     W_start = np.random.normal(0, 0.01, (K, np.size(trainX, 0)))
150     b_start = np.random.normal(0, 0.01, (K, 1))
151     lmlda = [0, 0, 0.1, 1]
152     bsize = 100
153     eta = [0.1, 0.001, 0.001, 0.001]
154     epochs = 40
155     accuracy = []
156     loss = []
157     weight_layers = []
158
159     for i in range(4):
160
161         W = W_start
162         b = b_start
163
164         accEpochsTrain = []
165         accEpochsVal = []
166         lossTrain = []
167         lossVal = []
168
169         # training the network
170         for epoch in range(epochs):
171             # minibatch returning W_star, b_star
172             W, b = miniBatch(trainX, trainY, train_y, W, b, lmlda[i], bsize, eta[i])
173
174             # compute training loss and accuracy for each epoch
175             J_train, P_train = computeCost(trainX, trainY, W, b, lmlda[i])
176             accTrain = computeAccuracy(P_train, train_y)
177             accEpochsTrain.append(accTrain)

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178         lossTrain.append(J_train)
179
180         # compute validation loss and accuracy for each epoch
181         J_val, P_val = computeCost(valX, valY, W, b, lmda[i])
182         accVal = computeAccuracy(P_val, val_y)
183         accEpochsVal.append(accVal)
184         lossVal.append(J_val)
185
186         # compute test loss and accuracy
187         J_test, P_test = computeCost(testX, testY, W, b, lmda[i])
188         accTest = computeAccuracy(P_test, test_y)
189
190         # collect results
191         accuracy.append([accEpochsTrain, accEpochsVal, accTest])
192         loss.append([lossTrain, lossVal, J_test])
193         weight_layers.append(W)
194
195     # plotting
196     for j in range(4):
197         print("Test accuracy:", accuracy[j][2], "Test loss:", loss[j][2], "lamda:", lmda[j], "eta:", eta[j])
198         plotGraph(accuracy[j][0], accuracy[j][1], range(epochs), "Accuracy",
199                  "Epochs", "Training accuracy", "Validation accuracy")
200         plotGraph(loss[j][0], loss[j][1], range(epochs), "Loss", "Epochs", "Training loss", "Validation loss")
201         montage(weight_layers[j])
202     plt.show()
203
204
205 if __name__ == "__main__":
206     main()
207

```