

Appendix

3

(a)

```
31 class Net(nn.Module):
32     def __init__(self, d):
33         super(Net, self).__init__()
34         # TO IMPLEMENT Part (a)
35         # here you should declare functions for layers with parameters
36         self.fc1 = nn.Linear(d, 10*d, bias=True)
37         self.fc2 = nn.Linear(10*d, 1, bias=True)
38
39     def forward(self, x):
40         # TO IMPLEMENT Part (a)
41         # here you should trace the forward computation of the network
42         # using functions declared in __init__
43         x = F.relu(self.fc1(x))
44         x = self.fc2(x)
45         return x.float()
46
```

```
79 # Initializing the data generation process
80 d = 5
81 data = DataGen(d)
82
83 ### Defining network parameters by hand
84
85 # Initializing the network
86 net = Net(d)
87
88 # Defining and setting parameters for perfect approximation
89 # TO IMPLEMENT Part (a)
90 v = data.v_star
91 w = np.zeros((10*d, d))
92 for i in range(10*d):
93     if i <= 3*d/2-1:
94         w[i,:] = v
95 b = np.zeros((10*d, 1))
96 alpha = np.zeros((1, 10*d))
97 for j in range((d//2)+1):
98     b[3*j,0] = -(2*j-0.5)
99     b[3*j+1,0] = -2*j
100     b[3*j+2,0] = -(2*j+0.5)
101     alpha[0,3*j] = 4
102     alpha[0,3*j+1] = -8
103     alpha[0,3*j+2] = 4
104 beta = np.array([-1])
105 set_parameters(d, w, b, alpha, beta, net)
106
107 # Checking that network perfectly approximates 10,000 examples
108 inputs, labels = data.get_batch(10000)
109 assert sum(net(inputs).reshape(-1) - labels.float()) == 0
```

(b)

```
117 # Defining the loss
118 ▼ def myLoss(outputs, labels):
119     # TO IMPLEMENT Part (b)
120     # should return  $1/b * \sum_{i=1}^b \text{loss}(f_p(x_i), y_i)$ 
121     outputs = outputs.view(-1,1)
122     labels = labels.view(-1,1)
123     n = outputs.size()[0]
124     ones = torch.ones(n, 1, dtype=float, requires_grad=True)
125     arg = torch.mul(outputs, labels)
126     arg = torch.mul(arg, -1)
127     arg = torch.add(arg, 1)
128     l = F.relu(arg)
129     ml = torch.sum(l)
130     ml = torch.div(ml, n)
131     return ml
```

```
162 d_list = [5, 10, 30]
163 for d1 in d_list:
164     net1 = Net(d1)
165     data1 = DataGen(d1)
166     loss_function1 = myLoss
167     optimizer1 = optim.SGD(net1.parameters(), lr=0.001, momentum=0.9)
168     batch_size1 = 100
169     num_iter1 = 5*10**4
170     loss_over_iteration1 = []
171     for m in range(num_iter1):
172         inputs1, labels1 = data1.get_batch(batch_size1)
173         optimizer1.zero_grad()
174         outputs1 = net1(inputs1)
175         loss1 = loss_function1(outputs1, labels1)
176         loss1.backward()
177         optimizer1.step()
178         loss_over_iteration1.append(loss1.item())
179         if m%200 == 0:
180             print('[%5d] loss: %.3f' % (m, loss1.item()))
181     print('Finished Training d=%2d' % d1)
182     plt.plot(np.array(loss_over_iteration1))
183 plt.legend(('d=5', 'd=10', 'd=30'))
184 plt.xlabel('Number of iteration')
185 plt.ylabel('Loss')
186 plt.show()
```

Plot

