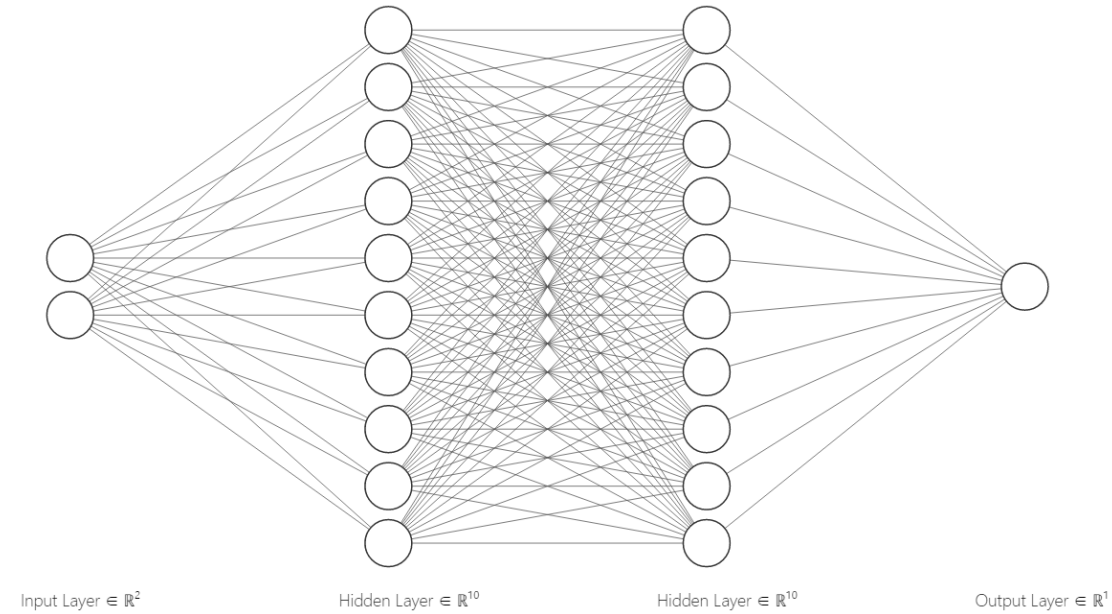


The structured fully connected neural network shows below, input layer contains 2 nodes, the first hidden layer contains 10 nodes, the second hidden layer contains 10 nodes, and the last output layer contains one node.



In this case the shape of b_1 , w_1 between input and first hidden layer are $(1,10)$, $(10,2)$, and shape of b_2 , w_2 between first and second hidden layer are $(1,10)$, $(10,10)$, respectively. And shape of b_3 , w_3 between second and output hidden layer are $(1,1)$ and $(1,10)$, respectively.

Forward propagation:

$$z_1 = xw_1^T + b_1$$

$$y_1 = \text{sigmoid}(z_1)$$

$$z_2 = y_1w_2^T + b_2$$

$$y_2 = \text{sigmoid}(z_2)$$

$$z_3 = y_2w_3^T + b_3$$

$$y_3 = z_3$$

Loss Function:

$$\text{loss} = (y_3 - y)^2$$

Backward Propagation:

$$\frac{\partial \text{loss}}{\partial w_3} = \frac{\partial \text{loss}}{\partial y_3} \frac{\partial y_3}{\partial w_3} = 2(y_3 - y) \cdot y_2$$

$$\frac{\partial \text{loss}}{\partial b_3} = \frac{\partial \text{loss}}{\partial y_3} \frac{\partial y_3}{\partial b_3} = 2(y_3 - y)$$

$$\frac{\partial \text{loss}}{\partial w_2} = \frac{\partial \text{loss}}{\partial y_3} \frac{\partial y_3}{\partial y_2} \frac{\partial y_2}{\partial z_2} \frac{\partial z_2}{\partial w_2} = 2(y_3 - y) \cdot w_3 \cdot \text{sigmoid}(z_2)(1 - \text{sigmoid}(z_2)) \cdot y_1$$

$$\frac{\partial \text{loss}}{\partial b_2} = \frac{\partial \text{loss}}{\partial y_3} \frac{\partial y_3}{\partial y_2} \frac{\partial y_2}{\partial z_2} \frac{\partial z_2}{\partial b_2} = 2(y_3 - y) \cdot w_3 \cdot \text{sigmoid}(z_2)(1 - \text{sigmoid}(z_2))$$

$$\begin{aligned} \frac{\partial \text{loss}}{\partial w_1} &= \frac{\partial \text{loss}}{\partial y_3} \frac{\partial y_3}{\partial y_2} \frac{\partial y_2}{\partial z_2} \frac{\partial z_2}{\partial y_1} \frac{\partial y_1}{\partial z_1} \frac{\partial z_1}{\partial w_1} \\ &= 2(y_3 - y) \cdot w_3 \cdot \text{sigmoid}(z_2)(1 - \text{sigmoid}(z_2)) \cdot w_2 \cdot \text{sigmoid}(z_1)(1 - \text{sigmoid}(z_1)) \cdot x \end{aligned}$$

$$\frac{\partial loss}{\partial b_1} = \frac{\partial loss}{\partial y_3} \frac{\partial y_3}{\partial y_2} \frac{\partial y_2}{\partial z_2} \frac{\partial z_2}{\partial y_1} \frac{\partial y_1}{\partial z_1} \frac{\partial z_1}{\partial b_1} =$$

$$= 2(y_3 - y) \cdot w_3 \cdot \text{sigmoid}(z_2)(1 - \text{sigmoid}(z_2)) \cdot w_2 \cdot \text{sigmoid}(z_1)(1 - \text{sigmoid}(z_1))$$

After this backward propagation, I use 100 samples and batch-size as 1 to generate the grad for w1,b1,w2,b2,w3 and b3. The file can be generated by torch.autograd() and by my self-implemented-backward-propagation code. Please see the all numbers in "my_autograd.dat" and "torch_autograd.dat" files.

Herein is the first sample of both files, the number of them are totally same.

my_autograd.dat - 记事本	torch_autograd.dat - 记事本
文件(F) 编辑(E) 格式(O) 查看(V) 帮助(H)	文件(F) 编辑(E) 格式(O) 查看(V) 帮助(H)
0:	0:
W1_grad: tensor([[-8.5401e-04, -1.8109e-04], [9.5278e-04, 2.0204e-04], [1.1784e-03, 2.4988e-04], [9.0200e-04, 1.9127e-04], [-9.4664e-04, -2.0073e-04], [-2.1553e-04, -4.5703e-05], [-2.0535e-04, -4.3545e-05], [2.7569e-04, 5.8460e-05], [-1.4033e-04, -2.9756e-05], [2.7065e-03, 5.7391e-04]])	W1_grad: tensor([[-8.5401e-04, -1.8109e-04], [9.5278e-04, 2.0204e-04], [1.1784e-03, 2.4988e-04], [9.0200e-04, 1.9127e-04], [-9.4664e-04, -2.0073e-04], [-2.1553e-04, -4.5703e-05], [-2.0535e-04, -4.3545e-05], [2.7569e-04, 5.8460e-05], [-1.4033e-04, -2.9756e-05], [2.7065e-03, 5.7391e-04]])
b1_grad: tensor([-0.0024, 0.0026, 0.0033, 0.0025, -0.0026, -0.0006, -0.0006, 0.0008, -0.0004, 0.0075])	b1_grad: tensor([-0.0024, 0.0026, 0.0033, 0.0025, -0.0026, -0.0006, -0.0006, 0.0008, -0.0004, 0.0075])
W2_grad: tensor([[0.0160, 0.0148, 0.0160, 0.0121, 0.0082, 0.0159, 0.0112, 0.0119, 0.0162, 0.0115], [0.0041, 0.0038, 0.0041, 0.0031, 0.0021, 0.0040, 0.0029, 0.0030, 0.0041, 0.0029], [0.0178, 0.0165, 0.0178, 0.0135, 0.0091, 0.0176, 0.0124, 0.0132, 0.0180, 0.0128], [0.0154, 0.0143, 0.0154, 0.0117, 0.0079, 0.0153, 0.0108, 0.0114, 0.0156, 0.0111], [-0.0039, -0.0037, -0.0039, -0.0030, -0.0020, -0.0039, -0.0028, -0.0029, -0.0040, -0.0028], [-0.0134, -0.0125, -0.0135, -0.0102, -0.0069, -0.0133, -0.0094, -0.0100, -0.0136, -0.0096], [-0.0083, -0.0077, -0.0083, -0.0063, -0.0042, -0.0082, -0.0058, -0.0061, -0.0084, -0.0059], [0.0230, 0.0213, 0.0231, 0.0174, 0.0118, 0.0228, 0.0161, 0.0171, 0.0233, 0.0165], [0.0079, 0.0073, 0.0079, 0.0060, 0.0040, 0.0078, 0.0055, 0.0058, 0.0080, 0.0057], [-0.0215, -0.0200, -0.0216, -0.0163, -0.0110, -0.0213, -0.0151, -0.0160, -0.0218, -0.0155]])	W2_grad: tensor([[0.0160, 0.0148, 0.0160, 0.0121, 0.0082, 0.0159, 0.0112, 0.0119, 0.0162, 0.0115], [0.0041, 0.0038, 0.0041, 0.0031, 0.0021, 0.0040, 0.0029, 0.0030, 0.0041, 0.0029], [0.0178, 0.0165, 0.0178, 0.0135, 0.0091, 0.0176, 0.0124, 0.0132, 0.0180, 0.0128], [0.0154, 0.0143, 0.0154, 0.0117, 0.0079, 0.0153, 0.0108, 0.0114, 0.0156, 0.0111], [-0.0039, -0.0037, -0.0039, -0.0030, -0.0020, -0.0039, -0.0028, -0.0029, -0.0040, -0.0028], [-0.0134, -0.0125, -0.0135, -0.0102, -0.0069, -0.0133, -0.0094, -0.0100, -0.0136, -0.0096], [-0.0083, -0.0077, -0.0083, -0.0063, -0.0042, -0.0082, -0.0058, -0.0061, -0.0084, -0.0059], [0.0230, 0.0213, 0.0231, 0.0174, 0.0118, 0.0228, 0.0161, 0.0171, 0.0233, 0.0165], [0.0079, 0.0073, 0.0079, 0.0060, 0.0040, 0.0078, 0.0055, 0.0058, 0.0080, 0.0057], [-0.0215, -0.0200, -0.0216, -0.0163, -0.0110, -0.0213, -0.0151, -0.0160, -0.0218, -0.0155]])
b2_grad: tensor([0.0257, 0.0066, 0.0285, 0.0247, -0.0063, -0.0216, -0.0133, 0.0370, 0.0127, -0.0346])	b2_grad: tensor([0.0257, 0.0066, 0.0285, 0.0247, -0.0063, -0.0216, -0.0133, 0.0370, 0.0127, -0.0346])
W3_grad: tensor([[-0.3461, -0.2203, -0.3023, -0.2546, -0.2992, -0.1894, -0.3107, -0.2192, -0.2811, -0.2055]])	W3_grad: tensor([[-0.3461, -0.2203, -0.3023, -0.2546, -0.2992, -0.1894, -0.3107, -0.2192, -0.2811, -0.2055]])
b3_grad: tensor([-0.5181])	b3_grad: tensor([-0.5181])