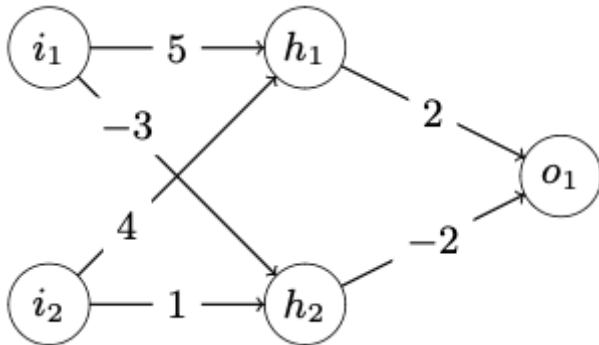


Task 2. Backpropagation (6 points)

Given the following neural network:



a) Calculate the output for the following inputs $i_1 = 5$ and $i_2 = 2$. The output function is the identity function. (0.5

P.)

$$net_{h_1} = i_1 * 5 + i_2 * 4 = 25 + 8 = 33$$

$$net_{h_2} = i_1 * -3 + i_2 * 1 = -15 + 2 = -13$$

$$out_{o_1} = h_1 * 2 + h_2 * -2 = 66 + 26 = 92$$

b) Calculate the error (MSE) for the input from part a). The actual output for the input is 100 (0.5 P.)

$$MSE = (92 - 100)^2 = 64$$

c) Apply the Backpropagation algorithm once and update all weights of the network. Use a learning rate of 0.0005. Calculate the output and error again for the input from part a). (4 P.)

w_i is the i -th weight reading left to right, top to bottom

Initial:

$$w_1 = 5, w_2 = -3, w_3 = 4, w_4 = 1, w_5 = 2, w_6 = -2$$

Calculation:

$$w'_i = w_i - \eta \frac{\delta E}{\delta w_i} = w_i - \eta \frac{\delta E}{\delta p} * \frac{\delta p}{\delta w_i}$$

$$\eta = 0.0005, p = 92, a = 100, E = \frac{1}{2}(p - a)^2$$

Calculate w'_i layer by layer backwards

Output Node

$$\begin{aligned} \frac{\delta}{\delta p} E * \frac{\delta}{\delta w_6} p &= \left[\frac{\delta}{\delta p} (p - a)^2 \right] * \left[\frac{\delta}{\delta w_6} h_1 * w_5 + h_2 * w_6 \right] = 2 * (p - a) * \frac{\delta}{\delta p} (p - a) * h_2 = 2 * (p - a) * i \\ \implies w'_6 &= w_6 - \eta * 2\Delta h_2 = -2 - .0005 * 2 * -8 * -13 = -2.104 \end{aligned}$$

Analogue w_5

$$\implies w'_5 = w_5 - \eta * \Delta * h_1 = 2 - .0005 * -8 * 33 = 2.264$$

Hidden Nodes

$$\begin{aligned} \frac{\delta}{\delta p} E * \frac{\delta}{\delta w_4} p &= \frac{\delta}{\delta p} E * \frac{\delta}{\delta h_2} p * \frac{\delta}{\delta w_4} h_2 = \frac{\delta}{\delta p} (p - a)^2 * \frac{\delta}{\delta h_2} (h_1 * w_5 + h_2 * w_6) * \frac{\delta}{\delta w_4} (i_1 * w_2 + i_2 * w_4) \\ \implies w'_4 &= w_4 - \eta * 2\Delta w_6 i_2 = 1 - .0005 * 2 * -8 * -2 * 2 = 0.968 \end{aligned}$$

Analogue w_3 to w_1

$$\begin{aligned} \implies w'_3 &= w_3 - \eta * 2\Delta w_5 i_2 = 4 - .0005 * 2 * -8 * 2 * 2 = 4.032 \\ \implies w'_2 &= w_2 - \eta * 2\Delta w_6 i_1 = -3 - .0005 * 2 * -8 * -2 * 5 = -3.080 \\ \implies w'_1 &= w_1 - \eta * 2\Delta w_5 i_1 = 5 - .0005 * 2 * -8 * 2 * 5 = 5.080 \\ \implies w'_1 &= 5.080, w'_2 = -3.080, w'_3 = 4.032, w'_4 = 0.968, w'_5 = 2.264, w'_6 = -2.104 \end{aligned}$$

d) How would your result change if you increased or decreased the learning rate by a factor of 10? (0.5 P.)

The increase or decrease of each new weight would increase or decrease by a factor of 10.

$$\begin{aligned} \eta = .00005 : w'_1 &= 5.0080, w'_2 = -3.0080, w'_3 = 4.0032, w'_4 = 0.0968, w'_5 = 2.0264, w'_6 = -2.0104 \\ \eta = .005 : w'_1 &= 5.80, w'_2 = -3.80, w'_3 = 4.32, w'_4 = 0.78, w'_5 = 4.64, w'_6 = -3.04 \end{aligned}$$

As we can see, the smaller learning rate simply adds a 0 after the comma. The bigger one massively influenced the weights connected to the output. If we have lots of data, it would cause huge fluctuations, with an epoch.

e) What would happen if you initialized all weights with 0? (0.5 P.)

Then they would stay at 0, since each new weight is multiplicatively dependent on the weights of the previous iteration.