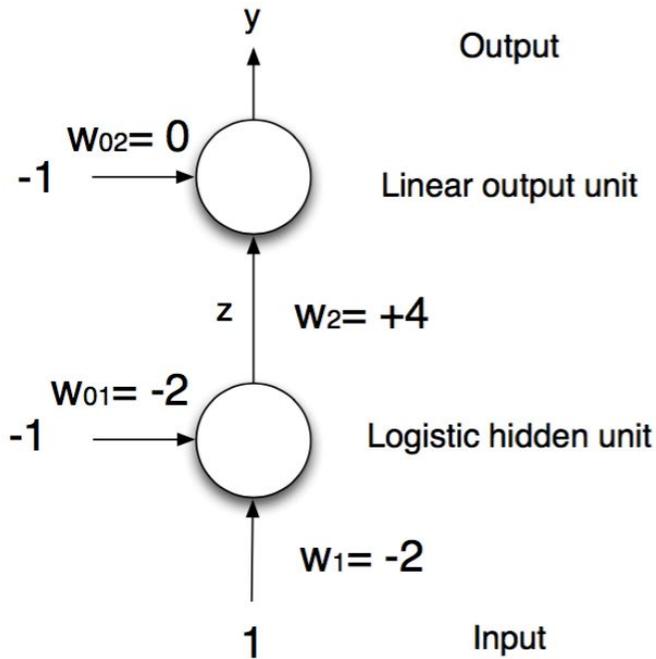


PROBLEM 23

Here you see a very small neural network: it has one input unit, one hidden unit (logistic), and one output unit (linear).



Let's consider one training case. For that training case, the input value is 1 (as shown in the diagram), and the target output value $t = 1$. We're using the following loss function:

$$E = \frac{1}{2}(t - y)^2$$

Please supply numeric answers; the numbers in this question have been constructed in such a way that you don't need a calculator. Show your work in case of mis-calculation in earlier steps.

- (a) What is the output of the hidden unit for this input?

Solution: 1/2

- (b) What is the output of the output unit for this input?

Solution: 2

- (c) What is the loss, for this training case?

Solution: 1/2

(d) What is the derivative of the loss with respect to w_2 , for this training case?

Solution: Let z be the output of the hidden unit

$$\begin{aligned}\frac{\partial E}{\partial w_2} &= (1-y) \frac{\partial(-y)}{\partial w_2} \\ &= (1-2) \cdot -z \\ &= (1-2) \cdot -(1/2) \\ &= (1/2)\end{aligned}$$

(e) What is the derivative of the loss with respect to w_1 , for this training case?

Solution:

$$\begin{aligned}\frac{\partial E}{\partial w_1} &= \frac{\partial E}{\partial z} \frac{\partial z}{\partial w_1} \\ &= (t-y) \frac{\partial(-y)}{\partial z} \cdot z \cdot (1-z) \cdot x \\ &= (t-y) \cdot -w_2 \cdot z \cdot (1-z) \cdot x \\ &= (1-2) \cdot -4 \cdot (1/2) \cdot (1/2) \cdot 1 \\ &= 1\end{aligned}$$

(f) With sigmoidal activation, the derivative with respect to w_1 and w_2 are

$$\frac{\partial E}{\partial w_2} = -(t-y)z, \text{ and } \frac{\partial E}{\partial w_1} = -(t-y) \cdot w_2 \cdot z \cdot (1-z) \cdot x.$$

Assume that we now use the rectified linear unit (ReLU) as our activation (or a *ramp* function). This means that $z = \max(0, w_1 x + w_{01})$. What is the derivative of the loss with respect to w_1 and w_2 at *differentiable points* with ReLU? Don't use numerical value for this question.

Solution: It is the same for w_2 as sigmoidal activation case:

$$\frac{\partial E}{\partial w_2} = -(t-y)z$$

For w_1 ,

$$\begin{aligned}\frac{\partial E}{\partial w_1} &= \frac{\partial E}{\partial z} \frac{\partial z}{\partial w_1} \\ &= (t-y) \frac{\partial(-y)}{\partial z} \cdot z' \cdot x \\ &= -(t-y) \cdot w_2 \cdot z' \cdot x,\end{aligned}$$

where,

$$z' = \begin{cases} 1 & \text{if } z > 0 \\ 0 & \text{if } z < 0. \end{cases}$$

At $z = 0$, it is not differentiable. For such points, we need to consider subdifferential (or subgradient), but it is not required in this question.

Additional Explanations

a) $(-2 * 1) + (-2 * -1) = 0$, and $\text{sigmoid}(0) = 0.5$.

b) $(0.5 * 4) + (0 * -1) = 2$.

c) $E = 0.5 * (2 - 1)^2 = 0.5$.