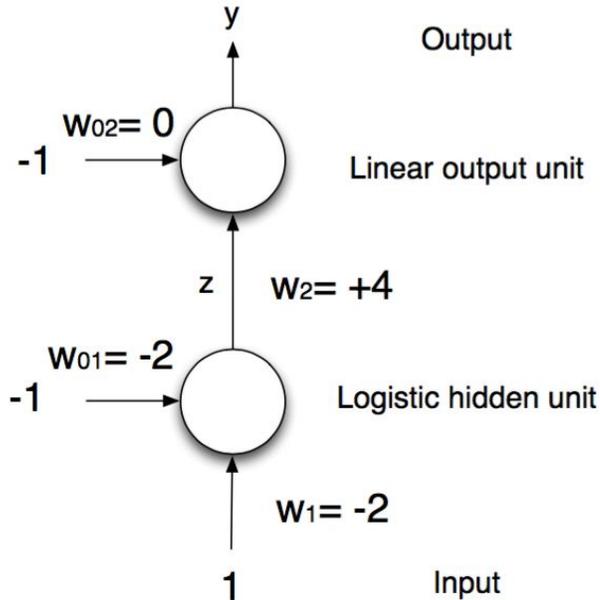


PROBLEM 23

14 Backpropagation

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.

- What is the output of the hidden unit for this input?
- What is the output of the output unit for this input?
- What is the loss, for this training case?
- What is the derivative of the loss with respect to w_2 , for this training case?
- What is the derivative of the loss with respect to w_1 , for this training case?
- 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 $z = \max(0, w_1x + w_{01})$. 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.