

Homework2

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1.1

$$\frac{\delta E}{\delta w_1} = \sum_{XEX} (out_x - o_x) \frac{\delta}{\delta w_1} (out_x - (w_0 + w_1 x_1 + w_1 x_{1x}^2 + \dots + w_n x_n + w_n x_x^2))$$

$$\frac{\delta E}{\delta w_1} (out_x - o_x) (-x_{ix} - x_{ix}^2)$$

1.2

a.

Input at neuron 1 = x1

Input at neuron 2 = x2

Output at neuron 1 = f(x1) = x1

Output at neuron 2 = f(x2) = x2

Input at neuron 3 = w31.x1 + w32.x2

Input at neuron 4 = w41.x1 + w42.x2

Output at neuron 3 = h(w31.x1 + w32.x2)

Output at neuron 4 = h(w41.x1 + w42.x2)

Input at neuron 5 = w53 * (output of neuron 3) + w54 * (output of neuron 4) =
w53. h(w31.x1 + w32.x2) + w54. h(w41.x1 + w42.x2)

Output at neuron 5 = h(w53. h(w31.x1 + w32.x2) + w54. h(w41.x1 + w42.x2))
= y5

y5 = h(w53. h(w31.x1 + w32.x2) + w54. h(w41.x1 + w42.x2))

b.

Input Layer:

Output of input layer = f(X) = X[Dimensions=2*1]

Hidden Layer:

Input of hidden layer = $W[1].X$ [Dimensions = $2*2 * 2*1 = 2*1$]

Output of hidden layer = $h(W[1].X)$ [Dimensions = $2*1$]

Output Layer:

Input at output layer = $(W[2].h)(h(W[1].X))$ [Dimensions = $1*2 * 2*1 = 1*1$]

Output of output layer = $h(W[2].h)(h(W[1].X))$ [Dimensions = $1*1$]

$y5 = h(W[2].h)(h(W[1].X))$ in terms of vector X , $W[1]$ and $W[2]$.

C.

First, let's compute the relationship between the sigmoid function $hs(x)$ and the tangent function $ht(x)$:

$$hs(x) = 1/(1 + e^{(-x)})$$

$$ht(x) = (e^x - e^{(-x)})/(e^x + e^{(-x)})$$

$$y5_{sigmoid} = hs(w53 * hs(w31 * x1 + w32 * x2) + w54 * hs(w41 * x1 + w42 * x2))$$

$$y5_{tanh} = ht(w53 * ht(w31 * x1 + w32 * x2) + w54 * ht(w41 * x1 + w42 * x2))$$

Now we will substitute $ht(x)$ with the expression we found in terms of $hs(x)$:

$$y5_{tanh} = \frac{(1 - hs(-(w53 * ht(w31 * x1 + w32 * x2) + w54 * ht(w41 * x1 + w42 * x2))))}{(1 + hs(-(w53 * ht(w31 * x1 + w32 * x2) + w54 * ht(w41 * x1 + w42 * x2))))}$$

We can see that these two functions are related, with the parameters differing only by linear transformations and constants.