

# HW 7

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## 1

As two nodes ( $i_1$  and  $i_2$ ) are provided in the network, and the number of nodes equals the feature dimension, we have that feature dimension = **2**

## 2

We will find  $o1$  and  $o2$  as follows:

We find  $net_{h1}$  and  $net_{h2}$  below, following the approach outlined in the article:

$$net_{h1} = .15 * .05 + .2 * .1 + .35 * 1 = .3775$$

$$net_{h2} = .25 * .05 + .3 * .1 + .35 * 1 = .3925$$

Note that we can also compute these equations as matrix multiplications, done below

$$i = \begin{bmatrix} .05 \\ .10 \end{bmatrix}$$

$$h_1 = \begin{bmatrix} .15 & .2 \end{bmatrix}$$

$$h_2 = \begin{bmatrix} .25 & .3 \end{bmatrix}$$

Therefore we have

$$net_{h1} = h_1 \cdot i + b_1$$

$$net_{h2} = h_2 \cdot i + b_1$$

Now we find  $out_{h1}$  and  $out_{h2}$  below, following the approach outlined in the article with the logistics function:

$$out_{h1} = \frac{1}{1 + e^{-.3775}} = 0.593269992$$

$$out_{h2} = \frac{1}{1 + e^{-.3925}} = 0.596884378$$

Now we find  $net_{o1}$  and  $net_{o2}$  below, following the approach outlined in the article:

$$net_{o1} = .4 * 0.593269992 + .45 * 0.596884378 + .6 * 1 = 1.105905967$$

$$net_{o2} = .5 * 0.593269992 + .55 * 0.596884378 + .6 * 1 = 1.2249214039$$

Note that again we can also compute these equations as matrix multiplications, done below

$$h = \begin{bmatrix} .593269992 \\ .596884378 \end{bmatrix}$$

$$o_1 = \begin{bmatrix} .4 & .45 \end{bmatrix}$$

$$o_2 = \begin{bmatrix} .5 & .55 \end{bmatrix}$$

Therefore we have

$$net_{o1} = o_1 \cdot h + b_2$$

$$net_{o2} = o_2 \cdot h + b_2$$

Now we find  $out_{o1}$  and  $out_{o2}$  below, following the approach outlined in the article with the logistics function:

$$out_{o1} = \frac{1}{1 + e^{-1.105905967}} = 0.751365069541$$

$$out_{o2} = \frac{1}{1 + e^{-1.2249214039}} = 0.772928465$$

From these equations, we have our outputs for  $o1, o2$ , which are  $out_{o1}, out_{o2}$

### 3

We can calculate the total squared error of the output of the forward pass as given by the article, using the output values we calculated in **section 2**

$$E_{total} = \sum \frac{1}{2} (target - output)^2$$

$$E_{o1} = \frac{1}{2} (.01 - 0.751365069541)^2 = 0.274811083168$$

$$E_{o2} = \frac{1}{2} (.99 - 0.772928465)^2 = 0.0235600256536$$

$$E_{total} = 0.274811083168 + 0.0235600256536 = 0.298371108822$$

Therefore we have

$$E_{total} = 0.298371108822$$

## 4

Following backward propagation, we will first compute  $\frac{df}{dw_k}$  for  $k = 5 \dots 8$

From the chain rule, we have that

$$\frac{df}{dw_5} = \frac{dE_{total}}{dOut_{o1}} \cdot \frac{dOut_{o1}}{dNet_{o1}} \cdot \frac{dNet_{o1}}{dw_5}$$

First we find

$$\frac{dE_{total}}{dOut_{o1}}$$

and as we have the original equation of  $E_{total}$  as:

$$\frac{1}{2}(target_{o1} - output_{o1})^2 + \frac{1}{2}(target_{o2} - output_{o2})^2$$

we have

$$\frac{dE_{total}}{dOut_{o1}} = (target_{o1} - output_{o1})^1 * -1 = output_{o1} - target_{o1}$$

Next,

$$out_{o1} = \frac{1}{1 + e^{-net_{o1}}}$$

Therefore

$$\frac{dOut_{o1}}{dNet_{o1}} = out_{o1}(1 - out_{o1})$$

Finally, we find

$$\frac{dNet_{o1}}{dw_5}$$

We have

$$Net_{o1} = w_5 \cdot out_{h1} + w_6 \cdot out_{h2} + b_2$$

Giving us:

$$\frac{dNet_{o1}}{dw_5} = out_{h1}$$

Therefore we can calculate

$$\frac{df}{dw_5} = (output_{o1} - target_{o1}) \cdot out_{o1}(1 - out_{o1}) \cdot out_{h1}$$

Now with substitution (based on our previous work) we have:

$$\begin{aligned} \frac{df}{dw_5} &= (0.751365069541 - .01) \cdot 0.751365069541(1 - 0.751365069541) \cdot 0.593269992 \\ &= 0.0821670405506 \end{aligned}$$

I will now compute  $\frac{df}{dw_k}$  in similar fashion for  $k = 6$ , noting when differences occur:

$$\frac{df}{dw_6} = \frac{dE_{total}}{dOut_{o1}} \cdot \frac{dOut_{o1}}{dNet_{o1}} \cdot \frac{dNet_{o1}}{dw_6}$$

This gives the formula below (note that because of the proximity of  $w_5$  and  $w_6$  the only difference in the derivative is  $out_{h2}$  for the third term, which follows from how that term is computed):

$$\frac{df}{dw_6} = (output_{o1} - target_{o1}) \cdot out_{o1}(1 - out_{o1}) \cdot out_{h2}$$

Now with substitution (based on our previous work) we have:

$$\begin{aligned} \frac{df}{dw_6} &= (0.751365069541 - .01) \cdot 0.751365069541(1 - 0.751365069541) \cdot 0.596884378 \\ &= 0.0826676278128 \end{aligned}$$

I will now compute  $\frac{df}{dw_k}$  in similar fashion for  $k = 7$ , noting when differences occur. Here we will switch the derivatives for  $o2$  instead of  $o1$ , but otherwise we can follow a similar approach. Note that:

$$\frac{df}{dw_7} = \frac{dE_{total}}{dOut_{o2}} \cdot \frac{dOut_{o2}}{dNet_{o2}} \cdot \frac{dNet_{o2}}{dw_6}$$

So therefore:

$$\frac{df}{dw_7} = (output_{o2} - target_{o2}) \cdot out_{o2}(1 - out_{o2}) \cdot out_{h1}$$

Now with substitution (based on our previous work) we have:

$$\begin{aligned} \frac{df}{dw_7} &= (0.772928465 - .99) \cdot 0.772928465(1 - 0.772928465) \cdot 0.593269992 \\ &= -0.0226025404533 \end{aligned}$$

I will now compute  $\frac{df}{dw_k}$  in similar fashion for  $k = 8$ , noting when differences occur. In this case, only the final term (like the difference between  $k = 5$  and  $k = 6$ )

$$\frac{df}{dw_8} = (output_{o2} - target_{o2}) \cdot out_{o2}(1 - out_{o2}) \cdot out_{h2}$$

Therefore:

$$\begin{aligned} \frac{df}{dw_8} &= (0.772928465 - .99) \cdot 0.772928465(1 - 0.772928465) \cdot 0.596884378 \\ &= -0.0227402422625 \end{aligned}$$

For  $1 \dots 4$ , we will change our approach as we are computing an inner layer. We start with the following equation:

$$\frac{df}{dw_1} = \frac{dE_{total}}{dOut_{h1}} \cdot \frac{dOut_{h1}}{dNet_{h1}} \cdot \frac{dNet_{h1}}{dw_1}$$

Note that we must consider  $h1$  through the inputs of  $E_{o1}$  and  $E_{o2}$  as well. Therefore we have

$$\frac{dE_{total}}{dOut_{h1}} = \frac{dE_{o1}}{dOut_{h1}} + \frac{dE_{o2}}{dOut_{h1}}$$

Which then gives us:

$$\frac{dE_{total}}{dOut_{h1}} = \left( \frac{dE_{o1}}{dOut_{o1}} \cdot \frac{dOut_{o1}}{dNet_{o1}} \cdot \frac{dNet_{o1}}{dOut_{h1}} \right) + \left( \frac{dE_{o2}}{dOut_{o2}} \cdot \frac{dOut_{o2}}{dNet_{o2}} \cdot \frac{dNet_{o2}}{dOut_{h1}} \right)$$

We now substitute from previous computations:

$$\frac{dE_{total}}{dOut_{h1}} = (.74136507 \cdot 0.751365069541 \cdot (1 - 0.751365069541) \cdot .4) + \left( \frac{dE_{o2}}{dOut_{o2}} \cdot \frac{dOut_{o2}}{dNet_{o2}} \cdot .45 \right)$$

$$\frac{dE_{total}}{dOut_{h1}} = 0.0553994246866 + (-.24119 \cdot 0.772928465(1 - 0.772928465) \cdot .45)$$

$$\frac{dE_{total}}{dOut_{h1}} = 0.0553994246866 + -0.019049119$$

$$\frac{dE_{total}}{dOut_{h1}} = 0.0363503056866$$

As we have computed the other derivatives previously, we have

$$\frac{df}{dw_1} = 0.0363503056866 * .241300709 * .05 = 0.000438567726727$$

Now that we have

$$\frac{dE_{total}}{dOut_{h1}}$$

we can easily calculate

$$\frac{df}{dw_3} = \frac{dE_{total}}{dOut_{h1}} \cdot \frac{dOut_{h1}}{dNet_{h1}} \cdot \frac{dNet_{h1}}{dw_3}$$

We have

$$\frac{df}{dw_3} = 0.0363503056866 * .136918 * .1 = 0.0004977$$

We will follow the same approach as above to calculate for the  $w_2, w_4$

$$\frac{dE_{total}}{dOut_{h2}} = \left( \frac{dE_{o1}}{dOut_{o1}} \cdot \frac{dOut_{o1}}{dNet_{o1}} \cdot \frac{dNet_{o1}}{dOut_{h2}} \right) + \left( \frac{dE_{o2}}{dOut_{o2}} \cdot \frac{dOut_{o2}}{dNet_{o2}} \cdot \frac{dNet_{o2}}{dOut_{h2}} \right)$$

We can substitute our previous work:

$$\frac{dE_{total}}{dOut_{h2}} = (.74136507 \cdot 0.751365069541 \cdot (1 - 0.751365069541) \cdot .50$$

$$+ (-.15505 \cdot 0.772928465(1 - 0.772928465) \cdot .55$$

$$\frac{dE_{total}}{dOut_{h2}} = 0.0692492808582 - 0.0149670585441$$

$$\frac{dE_{total}}{dOut_{h2}} = .0567158$$

Now we have

$$\frac{df}{dw_2} = .0567158 \cdot (0.772928465(1 - 0.772928465)) * .05 = .000877714$$

Again, finding  $w_4$  is easy given we have  $w_2$ :

$$\frac{df}{dw_4} = .0567158 \cdot (0.772928465(1 - 0.772928465)) * .1 = 0.00099542$$

Finally we have  $df/dw_k$   $1 \cdots 8$  :

$$\frac{df}{dw_1} = 0.000438567726727$$

$$\frac{df}{dw_2} = 0.000877714$$

$$\frac{df}{dw_3} = 0.0004977$$

$$\frac{df}{dw_4} = 0.00099542$$

$$\frac{df}{dw_5} = 0.0821670405506$$

$$\frac{df}{dw_6} = 0.0826676278128$$

$$\frac{df}{dw_7} = -0.0226025404533$$

$$\frac{df}{dw_8} = -0.0227402422625$$

## 5

Now we will update each of the given  $1 \cdots 8$  to find  $w_k^+$  in terms of  $\tau$

$$w_1^+ = .15 - \tau * 0.000438567726727$$

$$w_2^+ = .20 - \tau * 0.000877714$$

$$w_3^+ = .25 - \tau * 0.0004977$$

$$w_4^+ = .30 - \tau * 0.00099542$$

$$w_5^+ = .40 - \tau * 0.0821670405506$$

$$w_6^+ = .45 - \tau * 0.0826676278128$$

$$w_7^+ = .50 - \tau * -0.0226025404533$$

$$w_8^+ = .55 - \tau * -0.0227402422625$$