Neural Networks

Homework 2

I Forward Pass

$$Z_1^i = \sum_{i=1}^{N} W_{2i,i} \cdot X_{2i,i} = 1 \cdot 2 + 0.5 \cdot 4 - 0.5 \cdot 4 = 2$$
 $Z_2^i = \sum_{i=1}^{N} W_{22,i} \cdot X_{22,i} = 1 \cdot (-1) + 0.5 \cdot 1 - 0.5 \cdot (-2) = 0.5$

Application of sigmoid function:

 $A_1^i = \text{Sigmoid}(2_1^i) = \frac{1}{1+e^{-2}1^i} = 0.881$
 $A_2^i = \text{Sigmoid}(2_2^i) = \frac{1}{1+e^{-2}1^i} = 0.622$
 $Z_1^i = W_{2i,i} \cdot A_1^i + W_{2i,i}^i \cdot A_2^i + 1 \cdot 2 = -1.265$
 $P(x) = e^{z_1^2} = e^{-2} \cdot A_2^i + 1 \cdot A_2^i + A_2^i \cdot A_2^i + 1 \cdot A_2^i + A_2^i \cdot A_2^i +$

Back ward Pass

First, start from $\frac{\partial J}{\partial w_{ii}^2}$; $\frac{\partial J}{\partial W_{ii}^2} = \frac{\partial J}{\partial p} \cdot \frac{\partial p}{\partial z_i^2} \cdot \frac{\partial z_i}{\partial w_{ii}^2}$ The derivative of p with respect to z_i^2 is needed,

that's simply: $\frac{\partial p}{\partial z_i^2} = e^{z_i^2}$ The derivative of J with respect to p: $\frac{\partial J}{\partial p} = p - y$ The derivatives of z_i^2 with respect to $w_i^2 : \frac{\partial z_i^2}{\partial w_{ii}^2} = a_i^2$ Then, substitute this into the expression

For $\frac{\partial J}{\partial w_{ii}^2} = \frac{\partial J}{\partial w_{ii}^2} = \frac{\partial J}{\partial w_{ii}^2} = \frac{\partial J}{\partial w_{ii}^2}$ Alter the values with the solutions from previous exercises, we have that: $\frac{\partial J}{\partial w_{ii}^2} = \frac{\partial J}{\partial w_{ii}^2} = \frac{\partial$

The values or the derivative of j to p and the derivative of p with respect to z_1^2 , so let us calculate: $\frac{\partial z_1^2}{\partial a_1^2} = N_{11}$ $\frac{\partial z_1^2}{\partial a_1^2} = C(z_1^2) \cdot (1 - \delta(z_1^2))$ $\frac{\partial z_1^2}{\partial z_1^2} = X_{11}$ $\frac{\partial z_1^2}{\partial z_1^2} = X_{11$