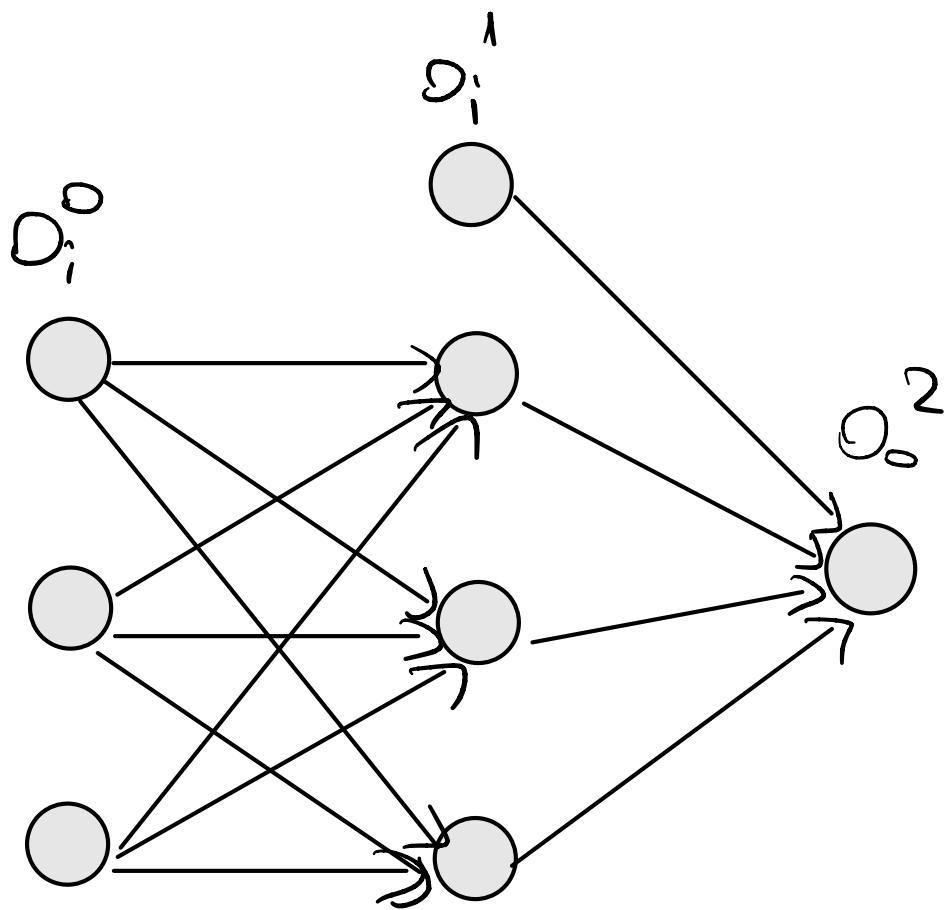


# Assignment 1



Regression Part

$$O_o^{(2)} = \sum_{k=0}^3 O_k^{(1)} \cdot a_{k0}^{(1)}$$

$$O_k^{(1)} = \sigma \left( \sum_{i=0}^n O_i^{(0)} \cdot a_{ik}^{(0)} \right) \text{ hence}$$

$$O_0^{(2)} = \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left( \sum_{i=0}^2 O_i \cdot a_{ik}^{(1)} \right)}} \right) a_{ik}^{(1)}$$

$$SE = Loss = (y - O_0^{(2)})^2 \quad \text{where } y \text{ is label}$$

Two Update Rules:

First Layer  $\rightarrow a_{ik}^{(0)} = a_{ik}^{(0)} - \alpha \frac{\partial SE(y, O_0^{(2)})}{\partial a_{ik}^{(1)}}$

Second Layer  $\rightarrow a_{k0}^{(1)} = a_{k0}^{(1)} - \alpha \frac{\partial SE(y, O_0^{(2)})}{\partial a_{k0}^{(1)}}$

I will find the derivatives for both of the layers respectively.

$$\frac{\partial \hat{S\Theta}(y, \theta^{(2)})}{\partial \theta_{i,k}^{(0)}} =$$

$$\frac{\partial \left( y - \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left( \sum_{i=0}^2 \theta_i^{(0)} \cdot \theta_{i,k}^{(0)} \right)}} \right) \theta_{i,k}^{(1)} \right)^2}{\partial \theta_{i,k}^{(0)}}$$

$$= 2 \left( y - \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left( \sum_{i=0}^2 \theta_i^{(0)} \cdot \theta_{i,k}^{(0)} \right)}} \right)^{-1} \theta_{i,k}^{(1)} \right)$$

↳ • -  $\left( -\theta_{i,0}^{(1)} \left( 1 + e^{-\sum_{i=0}^2 \theta_i^{(0)} \cdot \theta_{i,k}^{(0)}} \right)^{-2} \right) \left( e^{-\sum_{i=0}^2 \theta_i^{(0)} \cdot \theta_{i,k}^{(0)}} \right) \left( \sum_{i=0}^2 \theta_i^{(0)} \right)$   
derivative      chain 1      chain 2

Rule:

$$\theta_{i,k} = \theta_{i,k} - \alpha \left( 2 \left( y - \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left( \sum_{i=0}^2 \theta_i^{(0)} \cdot \theta_{i,k}^{(0)} \right)}} \right)^{-1} \theta_{i,k}^{(1)} \right) \right.$$

↳ •  $\left( -\theta_{i,0}^{(1)} \left( 1 + e^{-\sum_{i=0}^2 \theta_i^{(0)} \cdot \theta_{i,k}^{(0)}} \right)^{-2} \right) \left( e^{-\sum_{i=0}^2 \theta_i^{(0)} \cdot \theta_{i,k}^{(0)}} \right) \left( \sum_{i=0}^2 \theta_i^{(0)} \right)$

for 240 j

$$\partial_{\psi_0} = \partial_{\psi_0} - \lambda \frac{\partial \text{SEL}(y, \theta^{(2)})}{\partial \partial_{\psi_0}}$$

$$\frac{\partial \text{SSE}(y, \theta_0^2)}{\partial \theta_0} = \frac{\partial}{\partial \theta_0} \left( y - \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left(\sum_{i=0}^2 \theta_i \cdot x_{ik}^{(i)}\right)}} \right) \theta_k^{(i)} \right)^2$$

$$= 2 \left( y - \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left(\sum_{i=0}^2 \alpha_i^{(0)} \cdot x_{ik}^{(0)}\right)}} \right) a_{k0}^{(1)} \right) \cdot \left( \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left(\sum_{i=0}^2 \alpha_i^{(0)} \cdot x_{ik}^{(0)}\right)}} \right) \right)$$

derivative      |      chain

Hencej

$$\partial_{\epsilon_0} = \partial_{\epsilon_0} - \alpha \left( 2(y - \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left( \sum_{i=0}^2 \alpha_i^{(0)} \cdot \alpha_k^{(i)} \right)}} \right) \alpha_{\epsilon_0}^{(1)}) \cdot \left( - \sum_{k=0}^3 \left( \frac{1}{1 + e^{-\left( \sum_{i=0}^2 \alpha_i^{(0)} \cdot \alpha_k^{(i)} \right)}} \right) \right) \right)$$

## Classification Part

Update Rule:  $w = w - \alpha \frac{\partial E(x)}{\partial w}$

$$E(x) = CE(\ell, \ell') = - \sum_i \ell_i \cdot \log(\ell'_i)$$

where  $\ell = [l_0, l_1, l_2]$

$$\ell' = [o_0^2, o_1^2, o_2^2]$$

$$o_n^2 = \text{softmax}(x_n^{(2)}, x^{(2)})$$

$$o_n^2 = \frac{e^{x_n^{(2)}}}{\sum_{s=0}^S e^{x_s^{(2)}}}$$

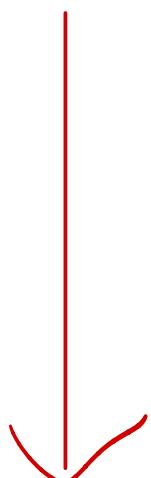
$$X_n^{(2)} = \sum_{k=0} O_k^{(1)} \cdot a_{k,n}^{(1)}$$

$$O_k^{(1)} = g\left(\sum_{i=0} O_i^{(0)} \cdot a_{i,k}^{(0)}\right)$$

$$O_k^{(0)} = \frac{1}{1 + e^{-\left(\sum_{i=0} O_i^{(0)} \cdot a_{i,k}^{(0)}\right)}}$$

$$X_n^{(2)} = \sum_{k=0} \left( \frac{1}{1 + e^{-\left(\sum_{i=0} O_i^{(0)} \cdot a_{i,k}^{(0)}\right)}} \right) a_{k,n}^{(1)}$$

So now, time for  $S_n^2$



$$\sum_{k=0}^l \left( \frac{1}{1 + e^{-\left(\sum_{i=0}^{l-1} o_i^{(0)} \cdot a_{ik}^{(0)}\right)}} \right)^2 \quad \text{--- } (1)$$

$$\sum_{s=0}^S \left( \sum_{k=0}^l \left( \frac{1}{1 + e^{-\left(\sum_{i=0}^{l-1} o_i^{(s)} \cdot a_{ik}^{(s)}\right)}} \right)^2 \right) \quad \text{--- } (2)$$

difference

$$= \sigma^2$$

For the sake of simplicity

I will call A and B for the nominator and denominator respectively.

Hence;

$$\sigma_n^2 = \frac{A_n}{B_n}$$

$$CE(\ell, \ell') = \sum_{i=0} p_i \cdot \log(\ell'_i)$$

For  $\partial_{\ell_n}$ :

$$\frac{\partial CE(\ell, \ell')}{\partial \ell_n} = \frac{\partial \sum_{i=0} p_i \cdot \log(\ell'_i)}{\partial (\ell_n)}$$

$$\log(\ell'_i) = \log\left(\frac{A_i}{B}\right)$$

derivative for this is:

by  $\left( \frac{A_i}{B} \right) \rightarrow$  we need to find  $A'_i$  for chain rule

no  $a_{kn}$  inside

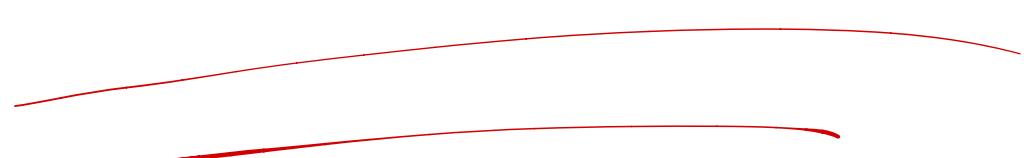
$$= \sum_{i=0}^l f_i \cdot \frac{1}{\frac{A_i}{B} \cdot \ln(2)} \cdot \left( \frac{A'_i}{B_i} \right)$$

(no  $a_{kn}$  inside)

$$A'_i = A_i \cdot \sum_{k=0}^o \sigma \left( \sum_{i=0}^o a_i^o \cdot a_{ik} \right)$$

(no  $a_{kn}$  inside)

$$\frac{\partial CE(l, l')}{\partial a_{kn}} = \sum_{i=0}^l f_i \cdot B_i \cdot \frac{\left( A_i \cdot \sum_{k=0}^o \sigma \left( \sum_{p=0}^o a_p^o \cdot a_{pk} \right) \right)}{A_i \cdot \ln(2) \cdot B_i}$$



For  $\alpha_{ik}$ :

$$\frac{\partial CE(\ell, \ell')}{\partial \alpha_{ik}} = \frac{\partial \sum_{i=0}^l p_i \cdot \log(\ell'_i)}{\partial (\alpha_{ik})}$$

$$\log(\ell'_i) = \log\left(\frac{A_i}{B_i}\right)$$

$$\frac{\partial CE(\ell, \ell')}{\partial \alpha_{ik}} = \sum_{i=0}^l p_i \cdot \left( \frac{1}{\frac{A_i}{B_i} \cdot \ln(2)} \cdot \frac{A'_i \cdot B_i - A_i \cdot B'_i}{B_i^2} \right)$$

So, we need  $A'_i$  and  $B'_i$  with respect to  $\alpha_{ik}$ . where

$$A_n = e^{\left( \sum_{k=0}^l \left( \frac{1}{1 + e^{(\sum_{i=0}^l \alpha_{ik})}} \right) \alpha_{kn} \right)}$$

$$\frac{\partial A_n}{\partial a_k} = \sum_{k=0} \left( -2 \left( \sum_{i=0} o_i^{(0)} \cdot a_i^{(0)} \right) \cdot a_k^{(1)} \right) \cdot \sum_{i=0} o_i^{(0)}$$

$$= A_n \cdot \sum_{k=0} - \left( \sum_{i=0} o_i^{(0)} \cdot a_i^{(0)} \right) \cdot \sum_{i=0} o_i^{(0)}$$

Time for  $B_n$  where;

$$B = \sum_{s=0} \left( e^{\left( \sum_{k=0} \left( \frac{1}{1 + e^{- \left( \sum_{i=0} o_i^{(0)} \cdot a_i^{(0)} \right)}} \right) a_k^{(1)} \right)} \right)$$

we can see that  
this part is as

This knowledge will make our derivative much simpler

We know that  $\frac{\partial A_s}{\partial \alpha_{ik}}$  which is

$$A_s \cdot \sum_{k=0}^q - \left( \sum_{i=0}^{n_s} o_i^{(0)} \cdot q_{ik}^{(0)} \right) \cdot \sum_{i=0}^{n_s} o_i^{(0)}$$

which makes  $\frac{\partial B}{\partial \alpha_{ik}}$  to be  $b_{kj}$

$$\sum_{s=0}^S A_s \cdot \sum_{k=0}^q - \left( \sum_{i=0}^{n_s} o_i^{(0)} \cdot q_{ik}^{(0)} \right) \cdot \sum_{i=0}^{n_s} o_i^{(0)}$$

Let's remind the loss function derivation.

$$\sum_{i=0}^I p_i \cdot \left( \frac{1}{\frac{A_i}{B} \cdot \ln(2)} \cdot \frac{A_i \cdot B - A_i \cdot \tilde{B}}{B^2} \right)$$

We know all the elements of this equation.

$$A_n = \left( \sum_{k=0}^{\infty} \left( \frac{1}{1 + e^{(\sum_{i=0}^{n-1} o_i^{(0)}, a_{ik}^{(0)})}} \right) a_{nk}^{(1)} \right)$$

$$B = \sum_{s=0}^{\infty} \left( \left( \sum_{k=0}^{\infty} \left( \frac{1}{1 + e^{(\sum_{i=0}^{n-1} o_i^{(0)}, a_{ik}^{(0)})}} \right) a_{ks}^{(1)} \right) \right)$$

$$\frac{\partial A_n}{\partial a_{ik}^{(0)}} = A_n \cdot \sum_{k=0}^{\infty} - \left( \sum_{i=0}^{n-1} o_i^{(0)} \cdot a_{ik}^{(0)} \right) \cdot \sum_{i=0}^{n-1} o_i^{(0)}$$

$$\frac{\partial B}{\partial a_{ik}^{(0)}} = \sum_{s=0}^{\infty} A_s \cdot \sum_{k=0}^{\infty} - \left( \sum_{i=0}^{n-1} o_i^{(0)} \cdot a_{ik}^{(0)} \right) \cdot \sum_{i=0}^{n-1} o_i^{(0)}$$