

1. When  $\tanh(x)=T$  and  $\text{sigmoid}(x)=S$  show that:

[2]

$$T = \frac{2S-1}{2S^2-2S+1}$$

$$S = \frac{1}{1+e^{-x}} = \frac{e^x}{1+e^x}$$

$$S + Se^x = e^x \Rightarrow e^x = \frac{S}{1-S}$$

$$\tanh(x) = \frac{e^{2x}-1}{e^{2x}+1} = \frac{(e^x)^2-1}{(e^x)^2+1}$$

Substituting for  $e^x$

$$\Rightarrow \frac{\left(\frac{S}{1-S}\right)^2-1}{\left(\frac{S}{1-S}\right)^2+1}$$

$$\Rightarrow \frac{S^2-(1-S)^2}{S^2+(1-S)^2}$$

$$\therefore \tanh(x) = \frac{2S-1}{2S^2-2S+1}$$

2. A deep neural network for a 4-class classification problem has, for a particular input, the pre-activation vector at output layer as  $[-1 \ 0 \ 5 \ 3]$  corresponding to classes [A B C D]. Which class will the model predict for the input? [2]

$$z = [-1 \ 0 \ 5 \ 3]$$

$$e^z = [0.368 \ 1 \ 148.41 \ 20.09]$$

$$\text{sum} = 169.868$$

$$\text{Softmax}(z) = e^z / \text{sum}(e^z) = [0.002 \ 0.006 \ 0.873 \ 0.118]$$

Model will predict Class C.

3. Consider the following neural network for binary classification  $[0 \ 1]$ . The input is  $[0.1 \ 0.5]^T$  and belongs to Class 1. Use an appropriate loss function.

- a) Compute the loss at the output node assuming hidden layer uses sigmoid activation function. [2]

$$z_1^1 = 0.1x_1 + 0.3x_2 + 0.25 = 0.41$$

$$a_1^1 = \text{sigmoid}(z_1^1) = 0.601$$

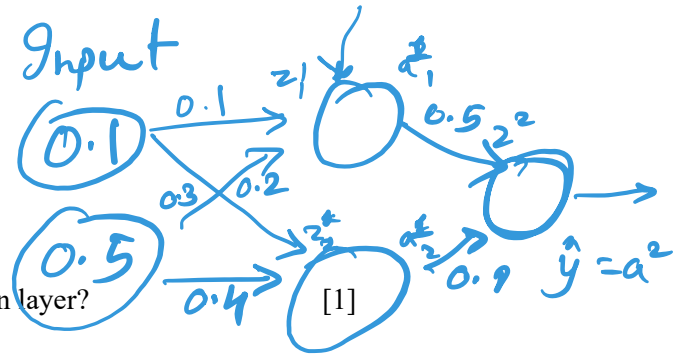
$$z_2^1 = 0.2x_1 + 0.4x_2 + 0.25 = 0.47$$

$$a_2^1 = \text{sigmoid}(z_2^1) = 0.615$$

$$z^2 = 0.5 a_1^1 + 0.8 a_2^1 + 0.35 = 1.1425$$

$$y' = \text{sigmoid}(z^2) = 0.758$$

$$\text{Loss} = -\log y' = 0.277$$



- b) What will be the loss if ReLU is used in the hidden layer?

$$z_1^1 = 0.41$$

$$a_1^1 = \text{ReLU}(z_1^1) = 0.41$$

$$z_2^1 = 0.47$$

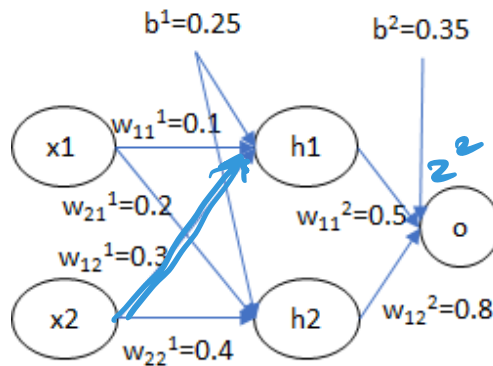
$$a_2^1 = \text{ReLU}(z_2^1) = 0.47$$

$$z^2 = 0.5 a_1^1 + 0.8 a_2^1 + 0.35 = 0.931$$

$$y' = \text{sigmoid}(z^2) = 0.717$$

$$\text{Loss} = -\log y' = 0.3327$$

- c) Derive the updated weight  $w_{12}^1$  after one iteration.  $w_{ij}^k$  refers to weight of connection between  $i$ th neuron in layer  $k$  with  $j$ th neuron of layer  $k-1$ . Assuming ReLU activation in hidden layer and a learning rate of  $\alpha = 0.1$ . [3]



$$\begin{aligned}\frac{\partial L}{\partial w_{12}^1} &= \frac{\partial L}{\partial \hat{y}} * \frac{\partial \hat{y}}{\partial z^2} * \frac{\partial z^2}{\partial a_1^1} * \frac{\partial a_1^1}{\partial z_1^1} * \frac{\partial z_1^1}{\partial w_{12}^1} \\ &= \frac{-1}{\hat{y}} * \hat{y}(1 - \hat{y}) * w_{11}^2 * 1 * x_2 \\ &= -(1 - \hat{y}) * w_{11}^2 * x_2\end{aligned}$$

$$w_{11}^2(\text{new}) = w_{11}^2(\text{old}) - 0.1 * \frac{\partial L}{\partial w_{12}^1}$$

$$\begin{aligned}w_{11}^2(\text{new}) &= 0.3 - 0.1 * [-(1 - 0.7402) * 0.5 * 0.5] \\ w_{11}^2(\text{new}) &= 0.306\end{aligned}$$

4. Give brief answers to the following questions:
- A deep neural network has 100 hidden layers. How can the depth affect the learning and performance of the network?
  - How does dropout help in increasing performance of deep neural networks?
  - "Using L1 loss enforces sparsity on the weights of the network." Do you agree with this statement? Why/Why not?
  - You train a deep neural network with a two hidden layers and observe that training and validation accuracy is low. You increase the number of hidden layers. Will this solve the issue or will there be a problem? Explain.
  - Consider the Loss vs. Iterations plot given below. Will early stopping technique be useful in this case? Justify.