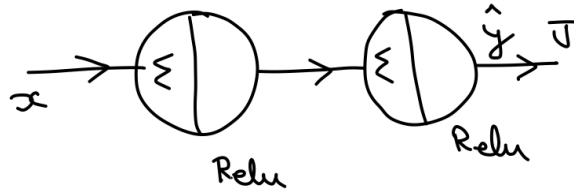


Backpropagation

Problem 1: What is being back propagated?

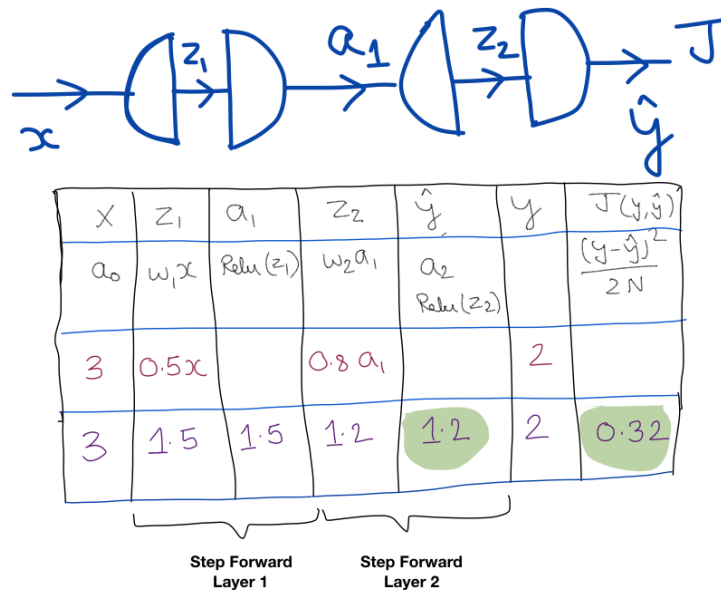
Problem 2: In the following setup



- How many functions are involved?
- How many weights are involved?
- With only one training sample $x=3$, $y=2$ and weights $w_1=0.5$, $w_2=0.8$ What's y_{hat} ?
- What's the MSE Cost?

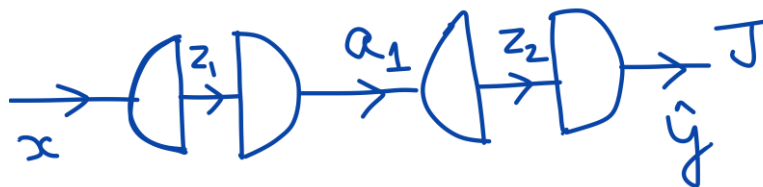
Solution 2:

- 5 Functions: z_1 , a_1 , z_2 , y_{hat} , J
- 2 Weights: w_1 , w_2
-



Problem 3:

Just expanded the above neural network to show weighted sum and activation components



To go backward, compute the following for previously given weights and training sample

$$\frac{dJ}{d\hat{y}} \quad \frac{d\hat{y}}{dz_2} \quad \frac{dz_2}{dw_2} \quad \frac{dz_2}{da_1} \quad \frac{da_1}{dz_1} \quad \frac{dz_1}{dw_1}$$

Solution 3:

| | | |
|---|--|------|
| $J(\hat{y}) = \frac{(y - \hat{y})^2}{2N}$ | $\frac{dJ}{d\hat{y}} = -(y - \hat{y})$ | -0.8 |
| $\hat{y}(z_2) = \text{Relu}(z_2)$ | $\frac{d\hat{y}}{dz_2} = \begin{cases} 1 & \text{if } z_2 > 0 \\ 0 & \text{if } z_2 < 0 \end{cases}$ | 1 |
| $z_2(a_1) = w_2 a_1$ | $\frac{dz_2}{da_1} = w_2$ | 0.8 |
| $z_2(w_2) = w_2 a_1$ | $\frac{dz_2}{dw_2} = a_1$ | 1.5 |
| $a_1(z_1) = \text{Relu}(z_1)$ | $\frac{da_1}{dz_1} = \begin{cases} 1 & \text{if } z_1 > 0 \\ 0 & \text{if } z_1 < 0 \end{cases}$ | 1 |
| $z_1(w_1) = w_1 x$ | $\frac{dz_1}{dw_1} = x$ | 3 |

Functions
Derivatives
value of derivatives

Problem 4:

Whole purpose of back propagation or this notebook is to compute **how cost changes with weights** i.e

$$\frac{dJ}{d\text{weights}}$$

What are the values of:

a) $\frac{dJ}{dw_1}$

b) $\frac{dJ}{dw_2}$

(Hint: Use Chain Rule & Problem 3 results)

Solution 4:

Propagating cost to w2

$$\begin{aligned} \frac{dJ}{d\hat{y}} &= -0.8 \\ \frac{dJ}{dz_2} &= \frac{dJ}{d\hat{y}} \cdot \frac{d\hat{y}}{dz_2} = -0.8 \\ \frac{dJ}{dw_2} &= \frac{dJ}{dz_2} \cdot \frac{dz_2}{dw_2} = -1.2 \end{aligned}$$

Propagating cost to w1

$$\begin{aligned} \frac{dJ}{d\hat{y}} &= -0.8 \\ \frac{dJ}{dz_2} &= \frac{dJ}{d\hat{y}} \cdot \frac{d\hat{y}}{dz_2} = -0.8 \\ \frac{dJ}{da_1} &= \frac{dJ}{dz_2} \cdot \frac{dz_2}{da_1} = -0.64 \\ \frac{dJ}{dz_1} &= \frac{dJ}{da_1} \cdot \frac{da_1}{dz_1} = -0.64 \\ \frac{dJ}{dw_1} &= \frac{dJ}{dz_1} \cdot \frac{dz_1}{dw_1} = -1.92 \end{aligned}$$

5. Update Weights

Problem 1: In the above problem after you have computed $\frac{dJ}{dw_1}$ and $\frac{dJ}{dw_2}$

what are new weights if the learning rate is 0.1

Solution 1:

Weight Update (Gradient Descent)

$$w_1 := w_1 - \alpha \frac{dJ}{dw_1} \quad \alpha: \text{learning rate}$$

$$w_2 := w_2 - \alpha \frac{dJ}{dw_2}$$

$$w_1 := 0.5 - 0.1(-1.92)$$

$$w_2 := 0.8 - 0.1(-1.2)$$

$$w_1 \\ 0.692$$

$$w_2 \\ 0.92$$

Forward Again

Problem 1 : Plug in those new weights and check if the cost is reduced.

Solution 1 :

Forward Epoch 2

| x | z_1 | a_1 | z_2 | \hat{y} | y | $J(y, \hat{y})$ |
|-------|---------|--------------------|------------|-----------|-----|------------------------------|
| a_0 | $w_1 x$ | $\text{ReLu}(z_1)$ | $w_2 a_1$ | a_2 | | $\frac{(y - \hat{y})^2}{2N}$ |
| 3 | 0.6922 | | 0.92 a_1 | | 2 | |
| 3 | 2.076 | 2.076 | 1.91 | 1.91 | 2 | 0.004 |

Step Forward
Layer 1

Step Forward
Layer 2