

$$1) \ a) \ g(x) := \text{GELU}(x) = x \sigma(1.702 x)$$

$$\frac{dg(x)}{dx} = \sigma(1.702x) + x \left( \sigma(1.702x) (1 - \sigma(1.702x)) \right) \times 1.702$$

For G.D.

$$x_{t+1} = x_t - \eta \nabla g(x_t)$$

t = 0

$$x_0 = 0, \quad g(0) = 0$$

t = 1

$$\Rightarrow x_1 = x_0 - \eta (0.1) \nabla g(0)$$

$$= 0 - 0.1 \left[ \sigma(0) + 0 \right] = -0.1 \left( \frac{1}{2} \right)$$

$$\Rightarrow \boxed{x_1 = -0.05}$$

t = 2

$$x_2 = x_1 - (0.1) \nabla g(-0.05)$$

$$= -0.05 - (0.1) \left[ \sigma(-0.05) - 0.05 \left( \sigma(-0.05) (1 - \sigma(-0.05)) \right) \times 1.702 \right]$$

$$= -0.05 - 0.1 \left[ \sigma(-0.085) - 0.05 \left( \sigma(-0.085) (1 - \sigma(-0.085)) \right) \times 1.702 \right]$$

$$= -0.05 - 0.1 \left[ 0.4787 - 0.05 (0.4787 (1 - 0.4787)) \times 1.702 \right]$$

$$\boxed{x_2 = -0.0957}$$

$$t=3$$

$$x_3 = x_2 - (0.1) \nabla g(-0.0957)$$

$$\Rightarrow x_3 = -0.0957 - 0.1 \left[ \sigma(-0.1629) (1 - \sigma(-0.1629)) \right] \times 1.702$$

$$x_3 = -0.0957 - 0.1 \left[ \sigma(-0.1629) - 0.0957 (\sigma(-0.1629) (1 - \sigma(-0.1629))) \right] \times 1.702$$

$$\boxed{x_3 = -0.1376}$$

function values,

$$\begin{cases} g(x_1) = -0.0239 \\ g(x_2) = -0.0439 \\ g(x_3) = -0.0608 \end{cases}$$

b) with  $\eta = 1.0$ ,  $x_0 = 0$

$$t=1$$

$$x_1 = 0 - (1.0) \nabla g(0) = -0.5$$

$$\Rightarrow \boxed{x_1 = -0.5} \quad \text{and} \quad \boxed{g(x_1) = -0.1496}$$

$$t=2$$

$$x_2 = -0.5 - (1.0) \left[ \sigma(-0.851) - (-0.5) (\sigma(-0.851) (1 - \sigma(-0.851))) \right] \times 1.702$$

$$\boxed{x_2 = -0.6208} \quad \Rightarrow \quad \boxed{g(x_2) = -0.1601}$$

t=3

$$x_3 = \cancel{x_2}(-0.6208) - (1.0) \left[ \sigma(-1.057) - (-0.6208) / (\sigma(-1.057)(1 - \sigma(-1.057))) \right] \times 1.702$$

$$\boxed{x_3 = -0.6765} \Rightarrow \boxed{g(x_3) = -0.1625}$$

with  $\eta = 1.0$ , the function ~~moved~~ value decreased by almost an order of magnitude faster than with  $\eta = 0.1$

c) i)  $x_0 = -3$ ,  $\eta = 0.1$

t=1

$$x_1 = (-3) - (0.1) \nabla g(0.1)$$

$$\boxed{x_1 = -2.9975} \Rightarrow \boxed{g(x_1) = -0.0181}$$

t=2

$$x_2 = (-2.9975) - (0.1) \nabla g(-2.9975)$$

$$\Rightarrow \boxed{x_2 = -2.9951} \Rightarrow \boxed{g(x_2) = -0.0182}$$

t=3

$$x_3 = (-2.9951) - (0.1) \nabla g(-2.9951)$$

$$\Rightarrow \boxed{x_3 = -2.9926} \Rightarrow \boxed{g(x_3) = -0.0183}$$

ii) GD with momentum,  $\beta = 0.9$ ,  $\eta = 0.1$ ,  $x_0 = -3$

$$v_0 = \nabla g(x_0) = -0.0245548$$

t=1

$$v_1 = (0.9)v_0 + (0.1)\nabla g(x_0)$$

$$\Rightarrow v_1 = -0.024548$$

$$x_1 = x_0 - (0.1)v_1$$

$$\Rightarrow x_1 = -2.9975, \quad g(x_1) = -0.0181$$

t=2

$$v_2 = (0.9)v_1 + (0.1)\nabla g(x_1)$$

$$v_2 = -0.24555$$

$$x_2 = x_1 - (0.1)v_2$$

$$\Rightarrow x_2 = -2.995085, \quad g(x_2) = -0.018192$$

t=3

$$v_3 = (0.9)v_2 + (0.1)\nabla g(x_2)$$

$$v_3 = -0.02457$$

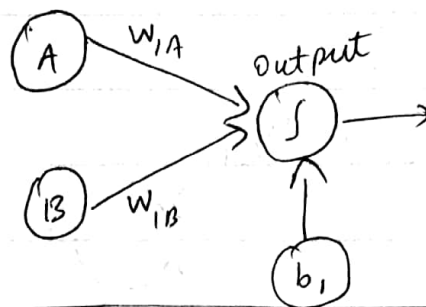
$$x_3 = x_2 - (0.1)v_3$$

$$\Rightarrow x_3 = -2.992632, \quad g(x_3) = -0.018253$$

iii) In this case both the methods perform almost the same <sup>as</sup> which could be seen ~~by~~ in the  $x_t$  and  $g(x_t)$  values.

2)

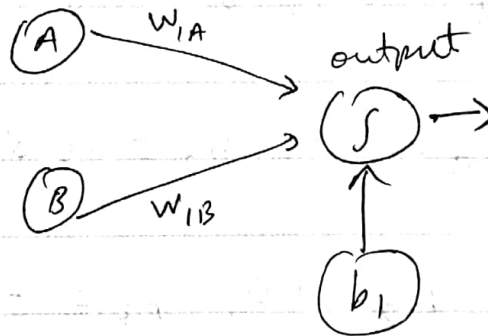
### AND gate



{ S : sigmoid }

$$w_{1A} = 1.0, w_{1B} = 1.0, b_1 = -1.5$$

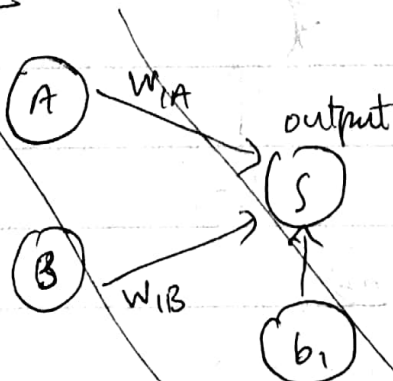
### OR gate



{ S : sigmoid }

$$w_{1A} = 1.0, w_{1B} = 1.0, b_1 = -0.5$$

### XOR gate

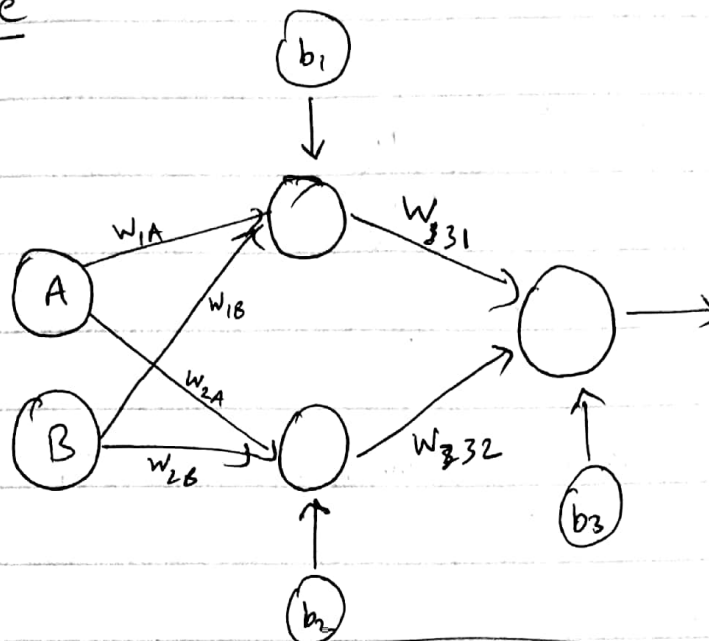


{ S : sigmoid }

$$w_{1A} = -0.4, w_{2A}$$

## XOR gate

(A)

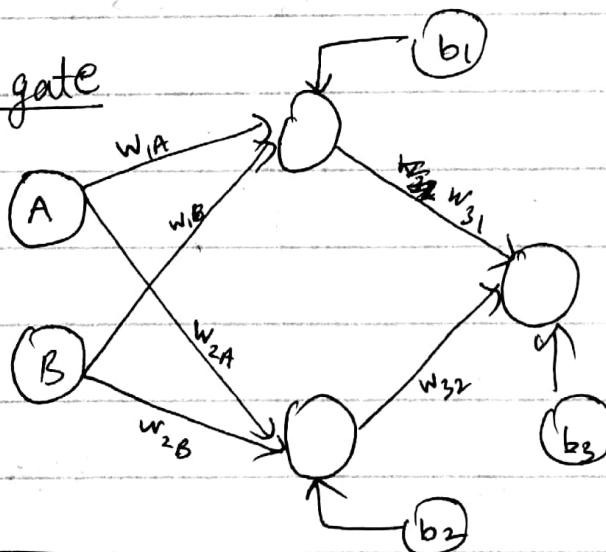


$$w_{1A} = -0.4, w_{1B} = -0.4, b_1 = 0.6$$

$$w_{2A} = 1.0, w_{2B} = 1.0, b_2 = -0.5$$

$$w_{31} = 1.0, w_{32} = 1.0, b_3 = -1.5$$

## XNOR gate



$$w_{1A} = 1.0, w_{1B} = 1.0, b_1 = -1.5$$

$$w_{2A} = -0.7, w_{2B} = -0.7, b_2 = 0.6$$

$$w_{31} = 1.0, w_{32} = 1.0, b_3 = -0.5$$

$$3) \ a) \quad E(\theta) = -\sum_{i=1}^K y_i \log \theta_i$$

$$\nabla_{f_2} E = \left( \nabla_{f_2} \theta \right) \left( \nabla_{\theta} E \right)$$

Now,  $\nabla_{\theta} E = \begin{bmatrix} \partial E / \partial \theta_1 \\ \vdots \\ \partial E / \partial \theta_i \\ \vdots \\ \partial E / \partial \theta_K \end{bmatrix} = \begin{bmatrix} -y_1 / \theta_1 \\ \vdots \\ -y_K / \theta_K \end{bmatrix}$

$$\Rightarrow \nabla_{\theta} E_i = -y_i / \theta_i$$

and  $\nabla_{f_2} \theta = \begin{bmatrix} \frac{\partial S(f_1)}{\partial f_{21}} & \dots & \frac{\partial S(f_{2K})}{\partial f_{21}} \\ \vdots & & \vdots \\ \frac{\partial S(f_1)}{\partial f_{2K}} & \dots & \frac{\partial S(f_{2K})}{\partial f_{2K}} \end{bmatrix}$

$$\nabla_{f_{2j}} \theta_i = \frac{\partial S(f_{2i})}{\partial f_{2j}} = \frac{e^{f_{2i}} \left( \frac{\partial f_{2i}}{\partial f_{2j}} \right) e^{f_{2j}}}{\left( \sum_{p=1}^K e^{f_{2p}} \right)^2}$$

$$= \frac{e^{f_{2i}}}{\sum_{p=1}^K e^{f_{2p}}} \left( \frac{\partial f_{2i}}{\partial f_{2j}} \right) - \frac{e^{f_{2i}} e^{f_{2j}}}{\left( \sum_{p=1}^K e^{f_{2p}} \right)^2}$$

if  $i=j$ ,  $\Rightarrow \frac{\partial f_{2i}}{\partial f_{2j}} = 1 \Rightarrow \nabla_{f_{2j}} \theta_i = \frac{e^{f_{2i}}}{\sum_{p=1}^K e^{f_{2p}}} \left[ 1 - \left( \frac{e^{f_{2i}}}{\sum_{p=1}^K e^{f_{2p}}} \right) \right]$

~~$\Rightarrow \frac{\partial f_{2i}}{\partial f_{2i}} =$~~

$$\Rightarrow \nabla_{f_{2i}} O_i = O_i (1 - O_i)$$

$$\text{if } i \neq j, \Rightarrow \frac{\partial f_{2i}}{\partial f_{2j}} = 0 \Rightarrow \nabla_{f_{2j}} O_i = -O_i O_j$$

$$\Rightarrow \nabla_{f_2} O = \begin{bmatrix} O_1(1-O_1) & \dots & -O_1 O_k \\ \vdots & & \vdots \\ -O_1 O_k & \dots & O_k(1-O_k) \end{bmatrix}$$

$$\text{or } \nabla_{f_{2j}} O_i = \begin{cases} O_i(1-O_i) & ; \text{ if } i=j \\ -O_i O_j & ; \text{ if } i \neq j \end{cases}$$

$$\Rightarrow \nabla_{f_2} E = \begin{bmatrix} O_1(1-O_1) & \dots & -O_1 O_k \\ \vdots & & \vdots \\ -O_1 O_k & \dots & O_k(1-O_k) \end{bmatrix} \begin{bmatrix} -y_1/O_1 \\ \vdots \\ -y_k/O_k \end{bmatrix}$$

~~In classification, let  $y_q = 1$  and  $y_{i \neq q} = 0$~~

$$\Rightarrow \nabla_{f_2} E =$$

$$\Rightarrow \nabla_{f_{2j}} E = \left[ \sum_{\substack{p=1 \\ p \neq j}}^K O_p (1-O_p) \left( -\frac{y_p}{O_p} \right) \right] - O_j O$$



(9)

$$\Rightarrow \nabla_{f_2 i} E_{\frac{1}{2}} = \left[ \sum_{\substack{j=1, \\ j \neq i}}^K (-o_i o_j) \left( -\frac{y_j}{o_j} \right) \right] + o_i (1 - o_i) \left( -\frac{y_i}{o_i} \right)$$

$$\Rightarrow \nabla_{f_2 i} E = \sum_{\substack{j=1, \\ j \neq i}}^K y_j o_i + y_i o_i - y_i$$

$$\Rightarrow \nabla_{f_2 i} E = \left[ \sum_{j=1}^K y_j o_i \right] - y_i$$

$$b) \nabla_x E = (\nabla_x f_1) (\nabla_{f_1} a) (\nabla_a f_2) (\nabla_{f_2} E)$$

$$\underline{\text{Now}} \rightarrow \nabla_x f_1 = \nabla_x (xw_1 + b_1) = w_1 \quad - (1)$$

$$\rightarrow \nabla_{f_1} a = \nabla_{f_1} \sigma(f_1) = \cancel{\sigma(f_1) (1 - \sigma(f_1))} \sigma(f_1) \mathbf{I} (\mathbf{I} - \sigma(f_1))$$

$$\Rightarrow \nabla_{f_1} a = a \mathbf{I} (\mathbf{I} - a \mathbf{I}) \quad - (2) \quad \left\{ \mathbf{I} = \underbrace{[1 \dots 1]^T}_{n \text{ times}} \right\}$$

$$\rightarrow \nabla_a f_2 = \nabla_a (aw_2 + b_2)$$

$$\Rightarrow \nabla_a f_2 = w_2$$

$$\Rightarrow \nabla_x E = w_1 \quad a \mathbf{I} (\mathbf{I} - a \mathbf{I}) w_2 \quad \nabla_{f_2} E$$

4) ~~at~~ Given the feature map has same Height and width ( $H_{in}$ )

$$H_{out} = \frac{H_{in} + \cancel{2p} 2p - d(k-1) - 1}{s} + 1 \quad \text{--- (1)}$$

a) given  $s = 1$ ,  $d = \text{dilation} = 1$ , and  $H_{out} = H_{in}$ ,  
and  $k = 3$

$$s(H_{in}) = \frac{H_{in} + 2p - (3-1) - 1}{1} + 1$$

$$2p = 2 \Rightarrow \boxed{p = 1}$$

$$F_{out} = \begin{bmatrix} 13.5 & -18.5 & 9 & -12.0 \\ 4 & 21.5 & 5 & 11.5 \\ 20.5 & 10.5 & 24.5 & 17 \\ 6 & 24 & 17 & 13.5 \end{bmatrix}$$

b) for filter 2,  $k = 2$

Using (1) with  $H_{out} = H_{in} = 4$ ,  $p = 1$ ,  $d = 1$

$$4 \text{ } \cancel{H_{out}} = \frac{\cancel{H_{in}} + 2 - 1(2-1) - 1}{s} + 1$$

$$3 = \frac{4}{s} \Rightarrow \boxed{s = \frac{4}{3}}$$

$\Rightarrow$   $s$  is not an integer  
so it is not possible to  
get a feature map with  $H_{out} = H_{in}$

$$c) \quad F' = \begin{bmatrix} 13.5 & 18.5 & 9 & -12 \\ 4 & 21.5 & 5 & 11.5 \\ 20.5 & 10.5 & 24.5 & 17 \\ 6 & 24 & 17.0 & 13.5 \end{bmatrix}$$

$$\text{Avg pool}(F') = \begin{bmatrix} 5.125 & 3.375 \\ 15.25 & 18.0 \end{bmatrix}$$

d) convolution filter for average pooling,

$$\text{Filter} = \begin{bmatrix} 0.25 & 0.25 \\ 0.25 & 0.25 \end{bmatrix}$$

$$\text{padding} = 0, \text{ stride} = 2$$