

$$x \in \mathbb{R}^D$$

$$w \in \mathbb{R}^D$$

$$p(y=1|x,w) = \sigma(w^T x + b)$$

$$p(y=1|x,w,b) = \sigma(w^T x + b)$$

$$X$$

$$N \times D$$

$$w$$

$$D$$

$$b$$

$$1$$

$$\hat{y}$$

$$N$$

$$X \cdot w + b$$

$$x_i^T$$

$$N$$

$$D$$

$$x_i^T w$$

$$x_i^T w + b$$

$$\pi \sqrt{2}$$

$$x_i$$

$$d(x,y) = \sqrt{\sum_{i=1}^D (x_i - y_i)^2}$$

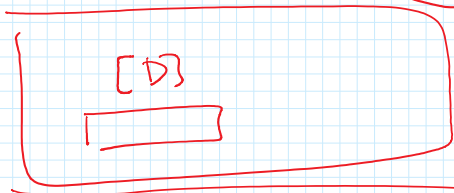
$$W$$

$$N \times D$$

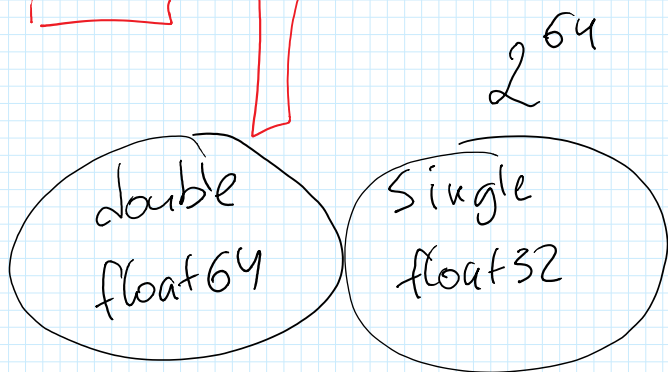
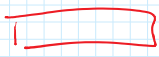
$$W_{ijk} = x_{ik} - x_{jk}$$

$$D_{ij} = \sqrt{\sum_k (x_{ik} - x_{jk})^2}$$

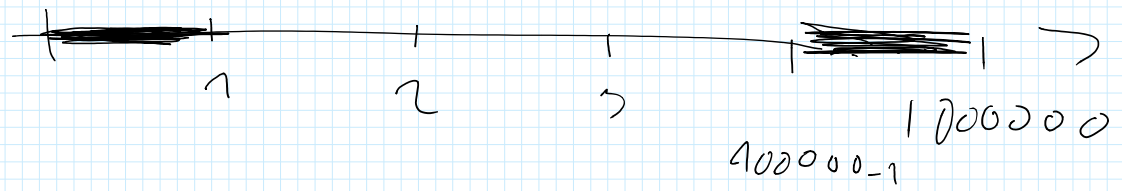
$$= \sqrt{\sum_k (w_{jk}^2)}$$



$[1, 0]$ $[0, 1]$



float16



$x = [10, 5, -2, -3.5]$

↓

$\text{softmax}(x) =$

$[0.8, 0.1, 0.06, 0.04]$

$$\text{softmax}(x) = \text{softmax}(x - a)$$

$$\frac{\exp(x_i - a)}{\sum_j \exp(x_j - a)} = \frac{\exp(x_i)}{\sum_j \exp(x_j)}$$

$$\overline{\sum_j \exp(x_j - a)} \stackrel{=}{=} \overline{\sum_j \exp(x_j)}$$

$$x = [0, 10]$$

$$e^0 = 1$$

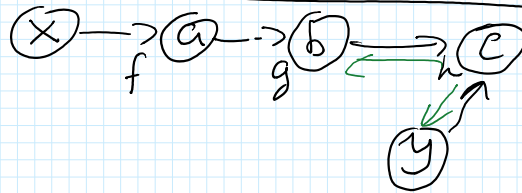
$$e^{10} =$$

$$a = f(x)$$

$$b = g(a)$$

$$c = h(b, y)$$

$$\frac{\partial c}{\partial x} = \begin{bmatrix} \frac{\partial c}{\partial b} & \frac{\partial b}{\partial a} \end{bmatrix} \frac{\partial a}{\partial x}$$



$$a = f.\text{forward}(x)$$

$$b = g.\text{forward}(a)$$

$$c = h.\text{forward}(b, y)$$

$$\begin{bmatrix} \frac{\partial c}{\partial b} & \frac{\partial c}{\partial y} \end{bmatrix} = h.\text{backward}(1.)$$

$$\begin{bmatrix} \frac{\partial c}{\partial a} \end{bmatrix} = g.\text{backward}\left(\begin{bmatrix} \frac{\partial c}{\partial b} \end{bmatrix}\right)$$

$$\frac{\partial c}{\partial x} = f.\text{backward}\left(\begin{bmatrix} \frac{\partial c}{\partial a} \end{bmatrix}\right)$$

$$\frac{\partial c}{\partial a} \frac{\partial a}{\partial x} = \frac{\partial c}{\partial x}$$

$$f(x, y) = x + y = z$$

$$\frac{\partial z}{\partial x} = \underbrace{[1 \ 1 \ 1 \ 1]}_D$$

$$a = x \cdot y$$

$$\frac{\partial a}{\partial x} = y$$

$$\frac{\partial a}{\partial y} = x$$

$$f(x,y) = x+y$$

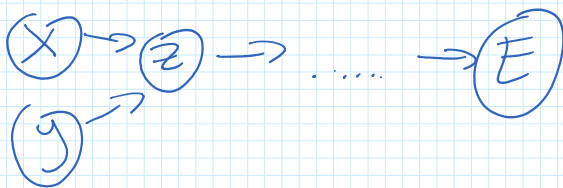
$$\mathbb{R}^D \times \mathbb{R}^D \rightarrow \mathbb{R}^D$$

$$z = f(x,y) = x+y$$

$$\frac{\partial z}{\partial x} = \begin{bmatrix} \frac{\partial z_i}{\partial x_j} \end{bmatrix}_D = \begin{bmatrix} 1 & 1 & 0 \\ 0 & 1 & 1 \\ & & \ddots \end{bmatrix}$$

$$z_i = x_i + y_i$$

$$\frac{\partial z_i}{\partial x_j} = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{if } i \neq j \end{cases}$$



$$z = f.\text{forward}(x,y)$$

$$E = \dots$$

$$\frac{\partial E}{\partial z}$$

$$\frac{\partial E}{\partial z} \quad 1 \times D$$

$$\frac{\partial E}{\partial x} = \frac{\partial E}{\partial z} \cdot \frac{\partial z}{\partial x} \quad 1 \times D \quad D \times D$$

$$\frac{\partial E}{\partial x}, \frac{\partial E}{\partial y} = \underline{f.\text{backward}} \left(\frac{\partial E}{\partial z} \right)$$

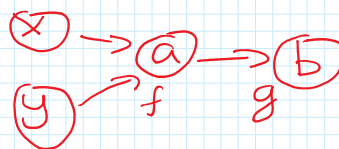
$$x, y \in \mathbb{R}^D$$

$$z = x^T y \quad \frac{\partial z}{\partial x} = y \quad \frac{\partial z}{\partial y} = x$$

$$[0] a = f(x,y) = x \odot y$$

$$[1] b = g(a) = \sum_{i=1}^D a_i$$

$$h(x,y) = x^T y \quad b = h(x,y)$$



$$\text{input} \cdot \frac{\partial b}{\partial a} \quad \text{input} \downarrow$$

$$\frac{\partial b}{\partial a} = g.\text{backward}(1.0)$$

$$\frac{\partial b}{\partial x}, \frac{\partial b}{\partial y} = f.\text{backward} \left(\frac{\partial b}{\partial a} \right)$$

$$\frac{\partial b}{\partial x} \quad [1, D]$$

$$x, y$$

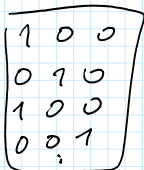
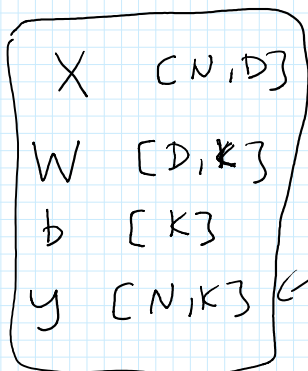
$$a = f.\text{forward}(x,y)$$

$$b = g.\text{forward}(a)$$

$$\frac{\partial b}{\partial x} = \frac{\partial b}{\partial a} \frac{\partial a}{\partial x}$$

f.backward(q):

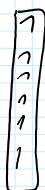
return $q \frac{\partial a}{\partial x}$, $q \cdot \frac{\partial a}{\partial y}$



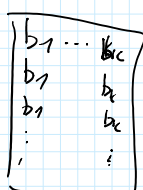
$y_{nk} = \begin{cases} 1 & \text{if sample } n \text{ belongs to class } k \\ 0 & \text{else} \end{cases}$

$$A = X \cdot W + \mathbf{1}_N \cdot b^T$$

$\begin{matrix} N \times D & D \times K & N \times 1 & 1 \times K \\ N \times K & & N \times K & \end{matrix}$



$b_1 \dots b_K$



$A = \text{Affine.forward}(X, W, b)$

A_{nk} = "unnormalized probability of n in class k "

$$P_{nk} = p(y_{nk} = 1 | X_n, W, b)$$

$$P_{nk} = \frac{\exp(A_{nk})}{\sum_c \exp(A_{nc})}$$

$$CE(y, P) = -\frac{1}{N} \sum_{n=1}^N \sum_{k=1}^K y_{nk} \cdot \log p(y_{nk} = 1 | X_n, W, b)$$

$$= -\frac{1}{N} \sum_n \sum_k y_{nk} \cdot \log P_{nk}$$

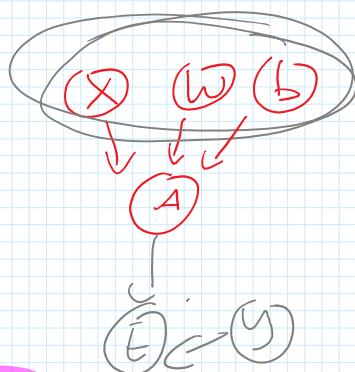
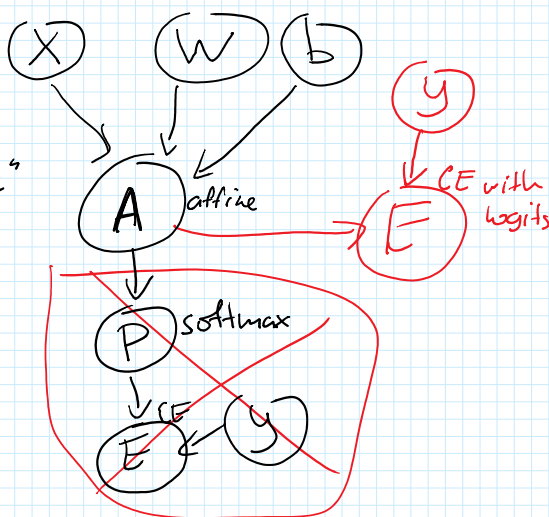
$$CE_{\text{logits}}(y, A) = \dots$$

$A = \text{Affine.forward}(X, W, b)$

$E = \text{CE.forward}(A, y)$

$$\frac{\partial E}{\partial A} = \frac{\partial E}{\partial y} = \text{CE.backward}(1.)$$

$N \times K$



Goal:
find

$$\frac{\partial E}{\partial W}, \frac{\partial E}{\partial b}$$

$$\frac{\partial A}{\partial x}$$

$$\hat{E} \leftarrow y$$

$$\frac{\partial E}{\partial x}, \frac{\partial E}{\partial w}, \frac{\partial E}{\partial b} = \text{Affine. backward} \left(\frac{\partial E}{\partial A} \right)$$

$$z = g(x) = \sum_i x_i \quad \leftarrow g.\text{back} \left(\frac{\partial E}{\partial z} \right)$$

$$g: \mathbb{R}^D \rightarrow \mathbb{R}$$

$$\frac{\partial z}{\partial x} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} \quad \frac{\partial z}{\partial x} = \underbrace{[1 \ 1 \ 1 \ 1]}_D$$

\uparrow
 $\frac{\partial z}{\partial x_i}$