

02/07/2023 Neural Networks

1) Review of SVD

$$A = \begin{bmatrix} | & | & & | \\ a_1 & a_2 & \dots & a_n \\ | & | & & | \end{bmatrix} = U \Sigma V^T$$

Data Matrix

columns of  $U$  provide a good model for columns of  $A$

$$a^* = U\alpha$$

Side note: Linear Operator Definition

$$A(\alpha_1 x_1 + \alpha_2 x_2) = \alpha_1 A x_1 + \alpha_2 A x_2$$

- SVD is linear because  $W$  is definitely separated from  $U$  and  $W$  is the parameters of  $A$ ;  $U$  and  $W$  are in product form

2) Neural Networks

- Sklearn's Iris dataset

- independent variables

- petal length, petal width, sepal length, sepal width

- classes (1, 2, 3)

- Multilayer Perceptron / Artificial Neural Networks

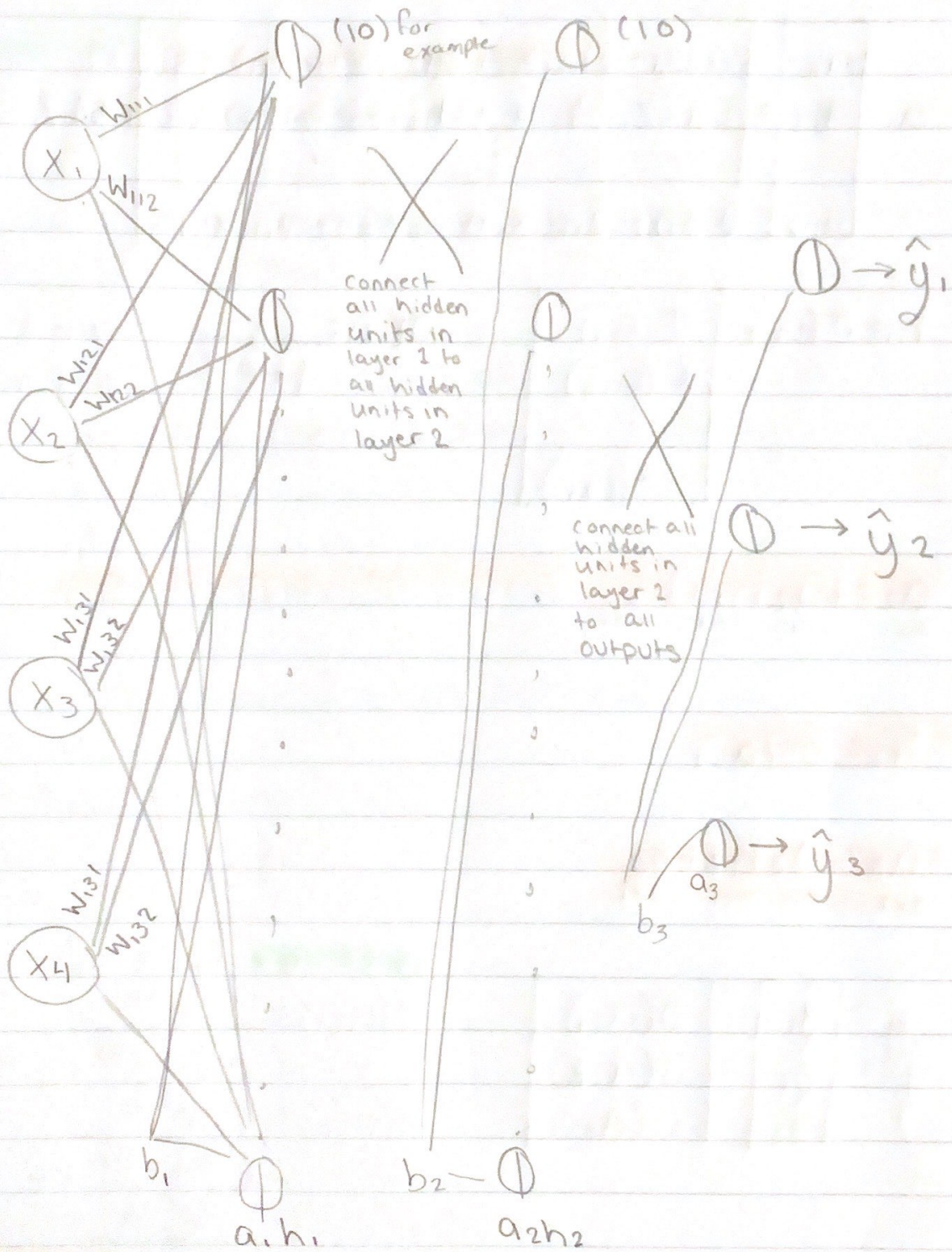
- weights

$W_{abc}$

where  $a$  = hidden layer number

where  $b$  = input variable number

where  $c$  = hidden unit number



$a_1 = W_1 X$   
 $10 \times 4 \quad 10 \times 4 \quad 4 \times 1$

$$a_1 = \begin{bmatrix} a_{11} \\ a_{12} \\ \vdots \\ a_{1n} \end{bmatrix} = \begin{bmatrix} W_{111}X_1 + W_{121}X_2 + W_{131}X_3 + W_{141}X_4 + b_1 \\ W_{112}X_1 + W_{122}X_2 + W_{132}X_3 + W_{142}X_4 + b_2 \\ \vdots \\ W_{11n}X_1 + W_{12n}X_2 + W_{13n}X_3 + W_{14n}X_4 + b_n \end{bmatrix} = W_1 X + b_1$$

$$h_1 = \sigma(a_1) = \begin{bmatrix} \sigma(a_{11}) \\ \sigma(a_{12}) \\ \vdots \\ \sigma(a_{1n}) \end{bmatrix} \quad \sigma(v) = \frac{1}{1+e^{-v}} \text{ (sigmoid function)}$$

$$a_2 = W_2 h_1 + b_2$$

$10 \times 1 \quad 10 \times 10 \quad 10 \times 1 \quad 10 \times 1$

$$h_2 = \sigma(a_2)$$

$$a_3 = W_3 h_2 + b_3$$

$3 \times 1 \quad 3 \times 10 \quad 10 \times 1 \quad 3 \times 1$

$$\hat{y} = \begin{bmatrix} \hat{y}_1 \\ \hat{y}_2 \\ \hat{y}_3 \end{bmatrix} = \begin{bmatrix} \sigma(a_{31}) \\ \sigma(a_{32}) \\ \sigma(a_{33}) \end{bmatrix}$$

soft max function

$$\sigma(a_{31}) = \frac{e^{a_{31}}}{e^{a_{31}} + e^{a_{32}} + e^{a_{33}}}$$

$$\sigma(a_{32}) = \frac{e^{a_{32}}}{e^{a_{31}} + e^{a_{32}} + e^{a_{33}}}$$

$$\sigma(a_{33}) = \frac{e^{a_{33}}}{e^{a_{31}} + e^{a_{32}} + e^{a_{33}}}$$

parameters:  $w_1, w_2, w_3, b_1, b_2, b_3$   
 $\begin{matrix} 10 \times 4 & 10 \times 10 & 3 \times 10 \\ 40 & 100 & 30 \end{matrix} + 3 = 173 \text{ parameters}$

$a$  represents the intermediate steps with the weights and bias

$h$  represents the result of putting  $a$  in a function, this gives us the non-linearity aspect

$$y = \sigma(w_3(\sigma(w_2(\sigma(w_1x + b_1)) + b_2)) + b_3)$$

- Other non-linear function options

- tanh

-  $\text{ReLU}(v) = \max(0, v)$