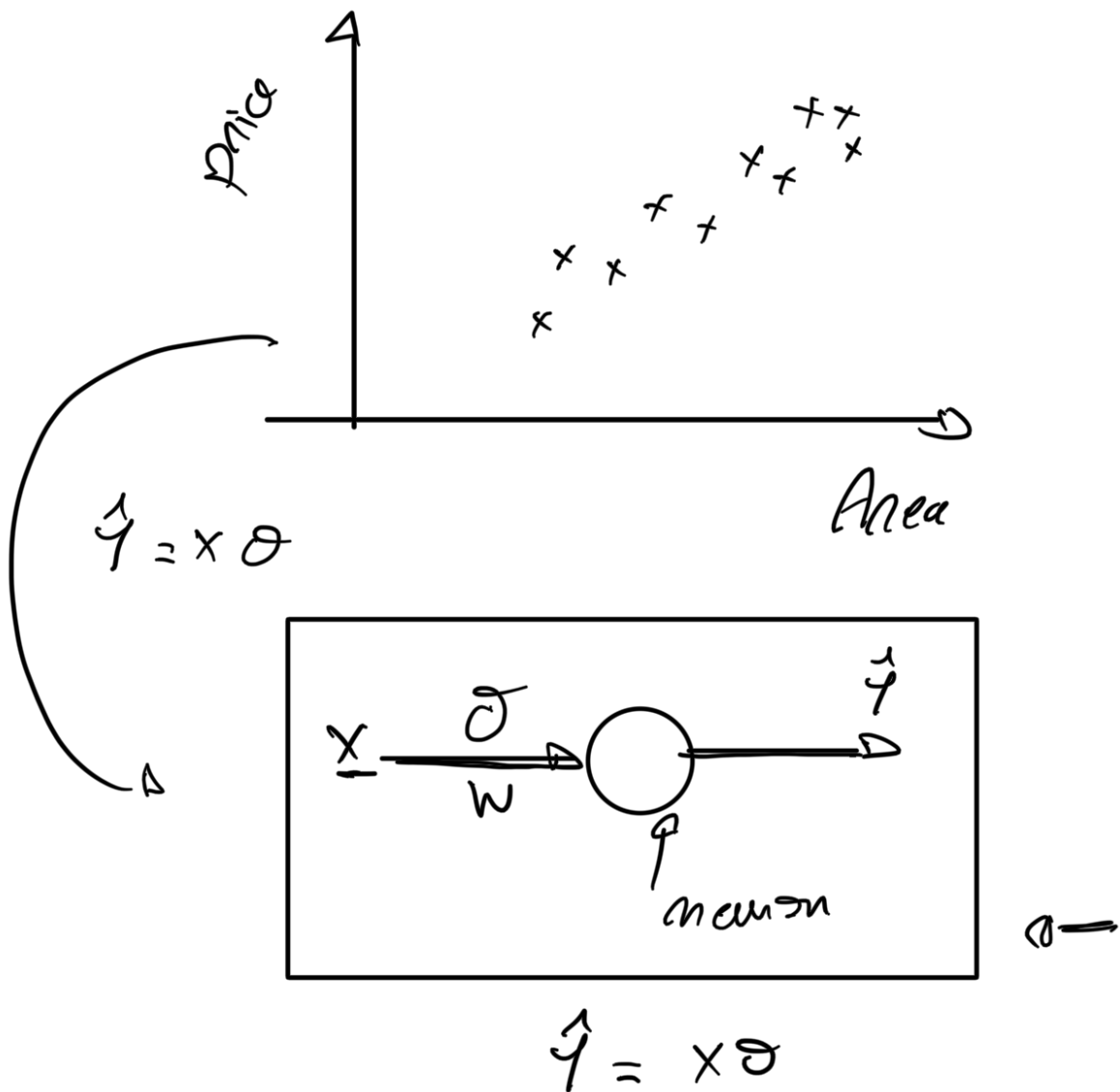


Neural Networks

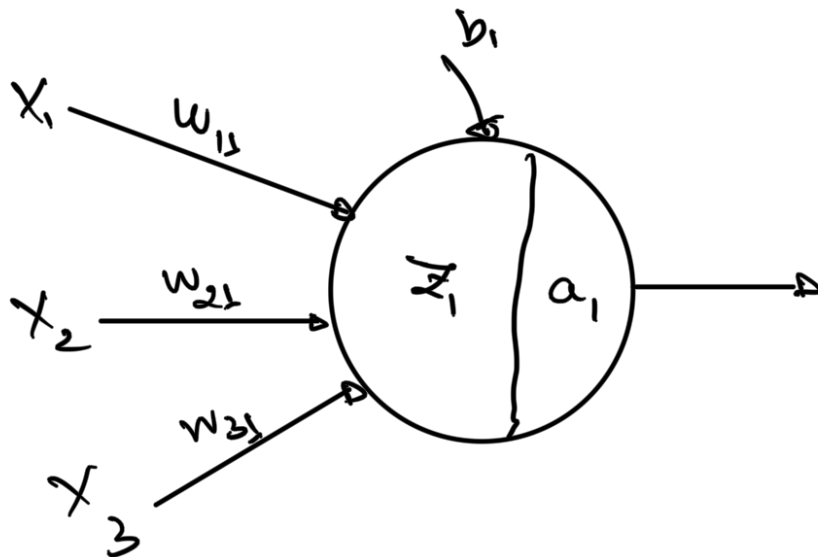
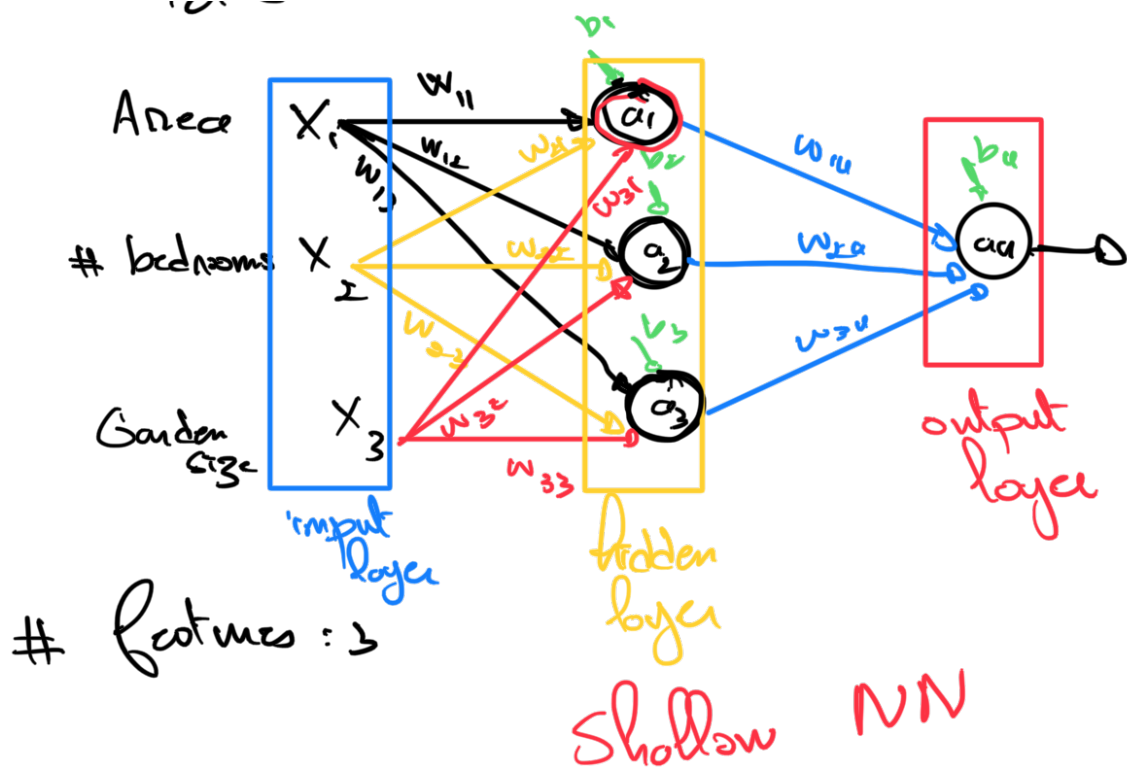
$$[\underline{x}_1, \underline{x}_2, \underline{x}_3, \underline{x}_4, \underline{x}_5]$$

$$\rightarrow [x_1, x_2, x_1 x_2 x_3, x_5 x_2]$$

$$h^5 = 1024$$



Let's consider more features



$$z_1 = w_{11}X_1 + w_{21}X_2 + w_{31}X_3 + b_1$$

$$a_1 = f(z_1) = \sigma(z_1)$$

$$z_2 = w_{12}X_1 + w_{22}X_2 + w_{32}X_3 + b_2$$

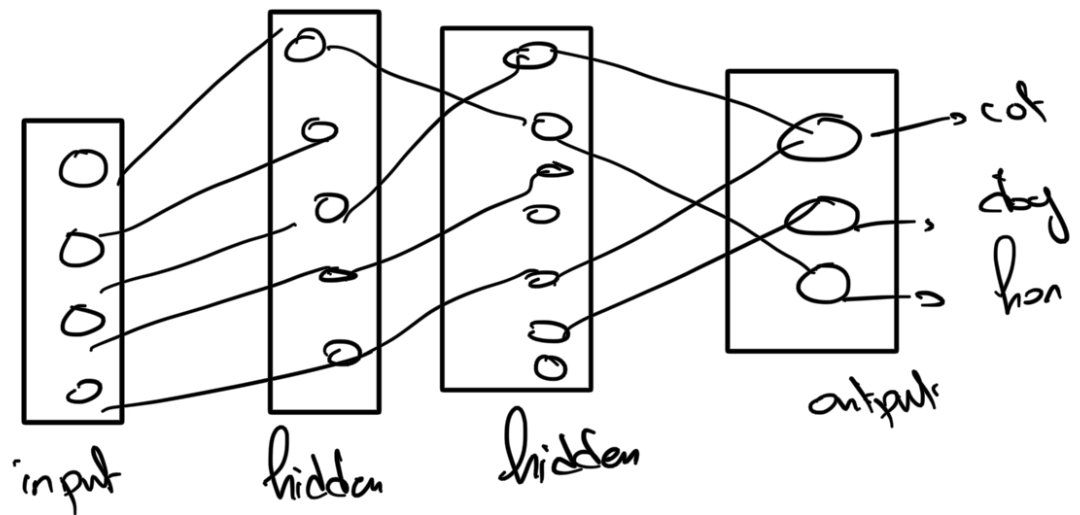
$$a_2 = \sigma(z_2)$$

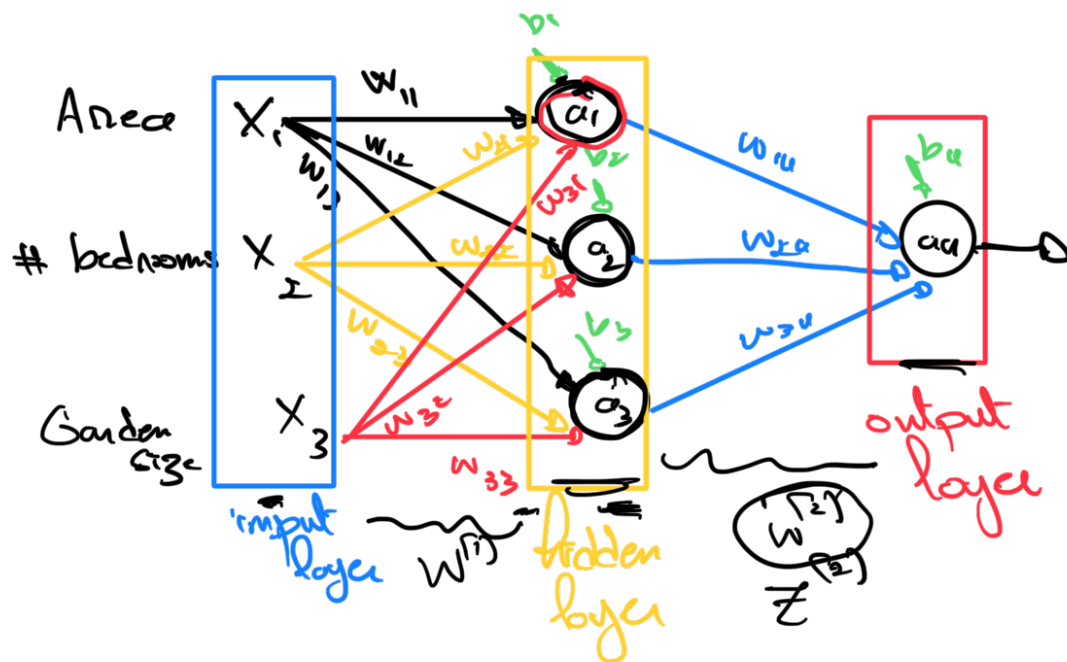
$$z_3 = w_{13} x_1 + w_{23} x_2 + w_{33} x_3 + b_3$$

$$a_3 = 5 \quad (x_3)$$

$$z_u = w_{1u} a_1 + w_{2u} a_2 + w_{3u} a_3 + b_u$$

$$a_u = z_u = 1$$





$$X = \begin{pmatrix} X_1 \\ X_2 \\ X_3 \end{pmatrix}$$

$$W = \begin{pmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & w_{22} & w_{23} \\ w_{31} & w_{32} & w_{33} \end{pmatrix}$$

$$b = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$

$$Z = W X + b \quad (3,1)$$

$$A = \text{Relu}(Z) \quad (3,1)$$

$$W^{(2)} = (w_{1u} \quad w_{2u} \quad w_{3u})$$

$$b^{(2)} = b_u$$

$$Z^{(2)} = W^{(2)} A + b^{(2)} \quad (1,1)$$

$$A^{(2)} = Z^{(1)} (1,1)$$



General Cox



features : M

$$\begin{pmatrix} w^{(1)} \\ b \end{pmatrix} \begin{pmatrix} w^{(2)} \\ b \end{pmatrix} \begin{pmatrix} w^{(3)} \\ b \end{pmatrix}$$

$$\begin{aligned} w^{(1)} &: (N, M) \\ b^{(1)} &: (N, 1) \\ w^{(2)} &: (H, N) \\ b^{(2)} &: (H, 1) \end{aligned}$$

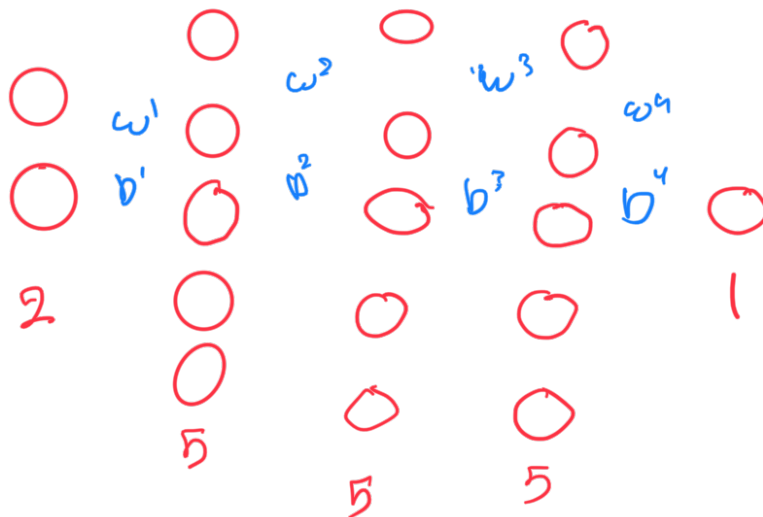
m, n, H, L

$w^{(3)} : (L, H)$
 $b^{(3)} : (L, 1)$

Exercise

Neural Network with 3 hidden layers
 each hidden layer has 5 neurons
 with input 2 neurons
 with output 1 neuron.

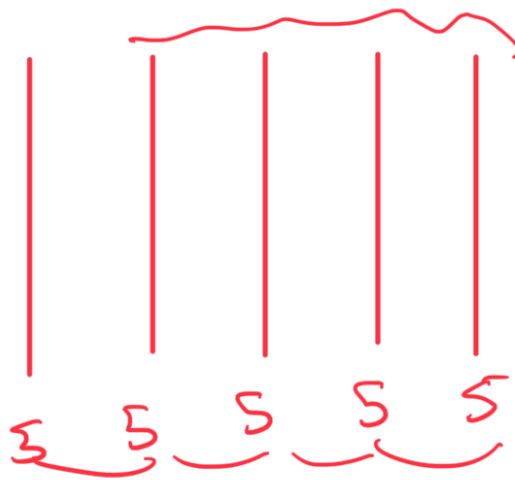
- ① Represent the NN
- ② Count the learnable parameters.



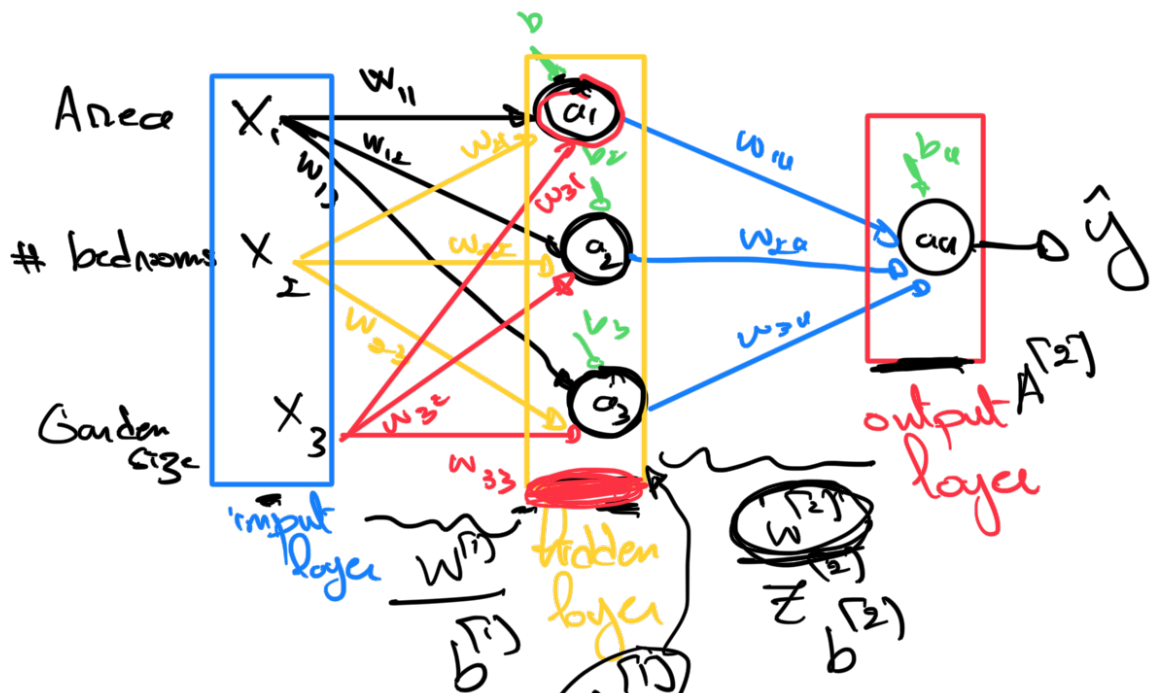
Exercise 2

11. Common each?

5 layer with 1 neuron - bias

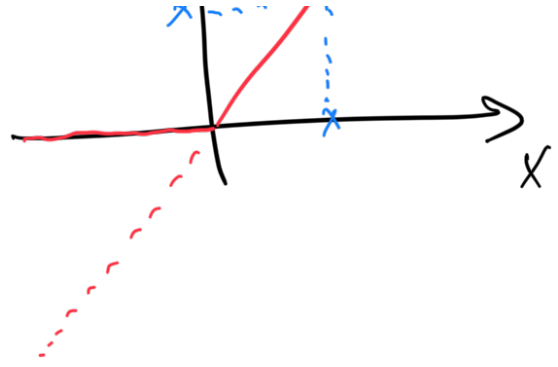


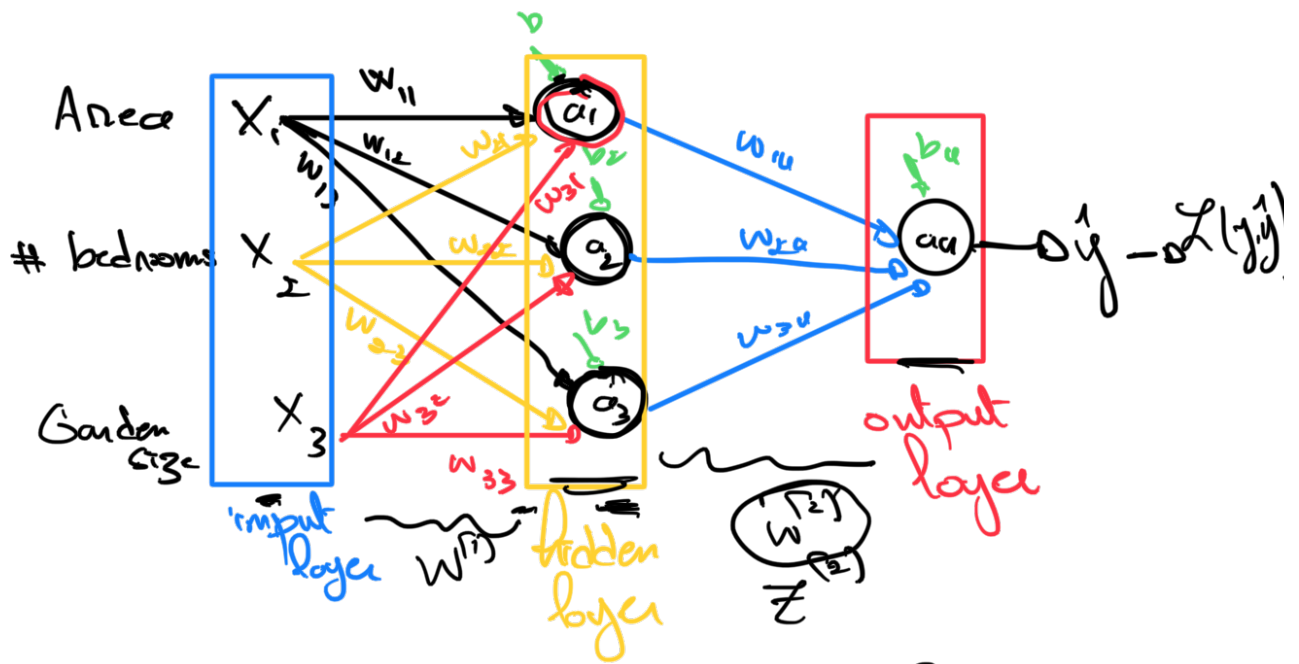
$$\begin{aligned}
 LP: & 5 \times 5 + 5 \times 5 + 5 \times 5 + 5 \times 5 \\
 & + \underbrace{5 + 5 + 5 + 5}_{\text{bias}} = 120
 \end{aligned}$$



$$\begin{aligned}
 \textcircled{a} \quad A^{[2]} = \hat{y} &= W^{[2]} A^{[1]} + b^{[2]} \\
 &= W^{[2]} \left(W^{[1]} X + b^{[1]} \right) + b^{[2]} \\
 &= \underbrace{W^{[2]} W^{[1]}}_{\theta} X + \underbrace{W^{[2]} b^{[1]} + b^{[2]}}_B \\
 \hat{y} &= \theta X + B \\
 A^{[2]} = \hat{y} &= \sigma \left(W^{[2]} A^{[1]} + b^{[2]} \right) \\
 &= \sigma \left(W^{[2]} \text{Relu} \left(W^{[1]} X + b^{[1]} \right) + b^{[2]} \right)
 \end{aligned}$$

$z \uparrow$ $\text{Relu}(x)$





GD:

