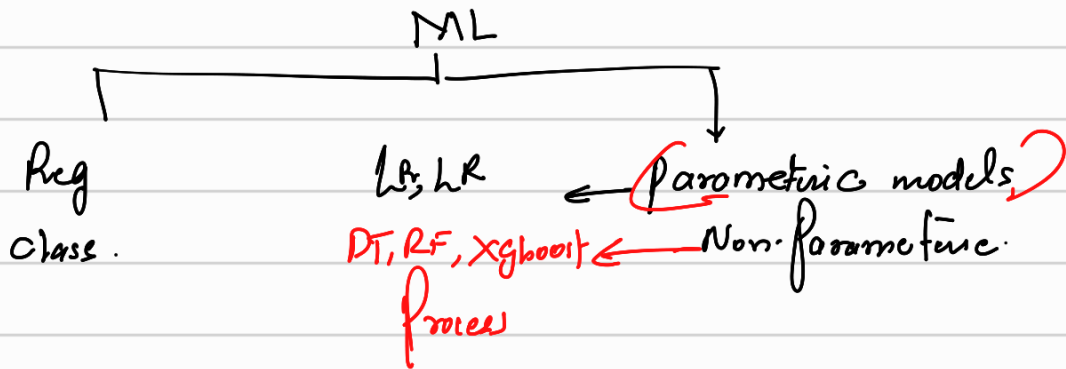
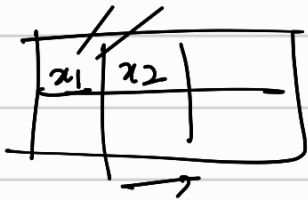


Deep learning

$$y = f(x) + \epsilon$$

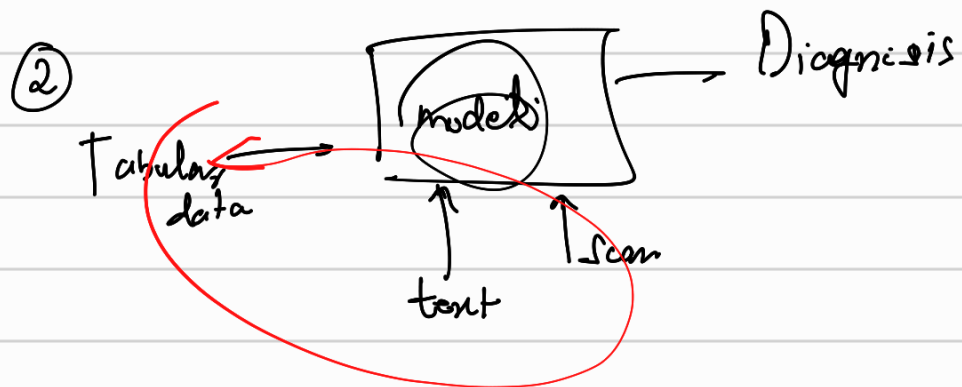


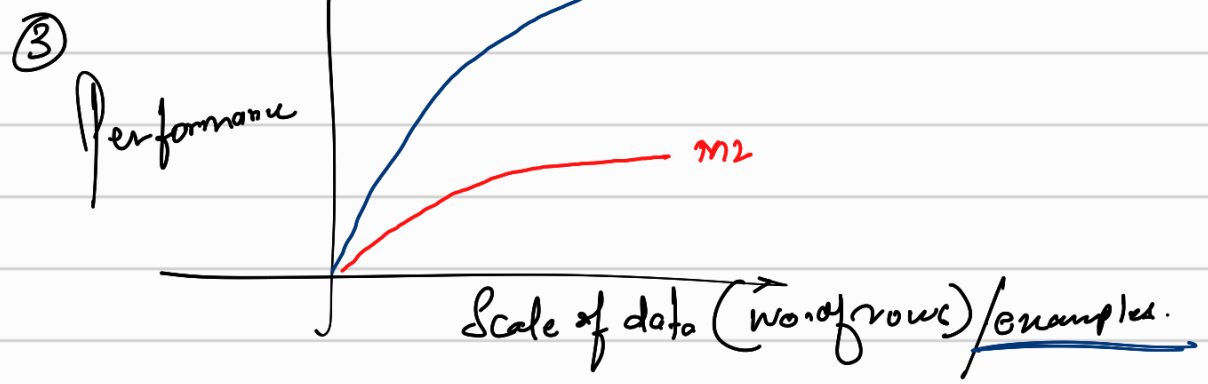
$$y \sim f(x)$$



① $x \rightarrow$ (video, audio, text, tabular data, graph).
architecture

multimodality





$$ML$$

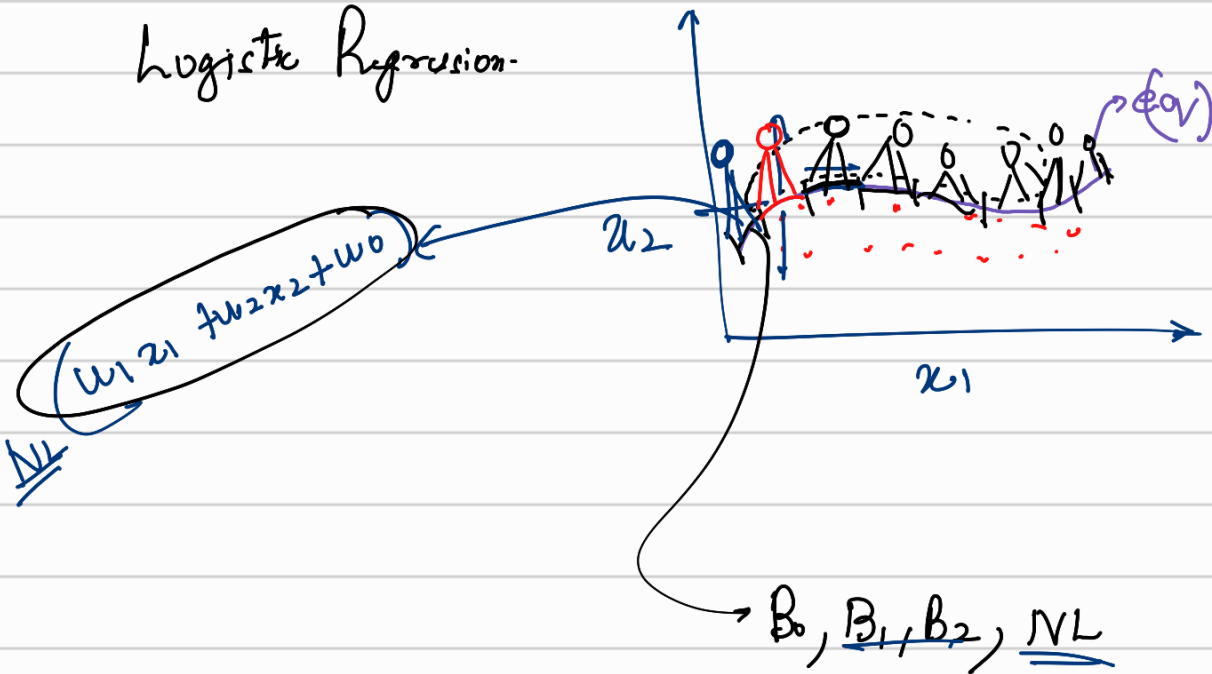
x_1	x_2	x_3

DL \rightarrow automated feature engineering

DL

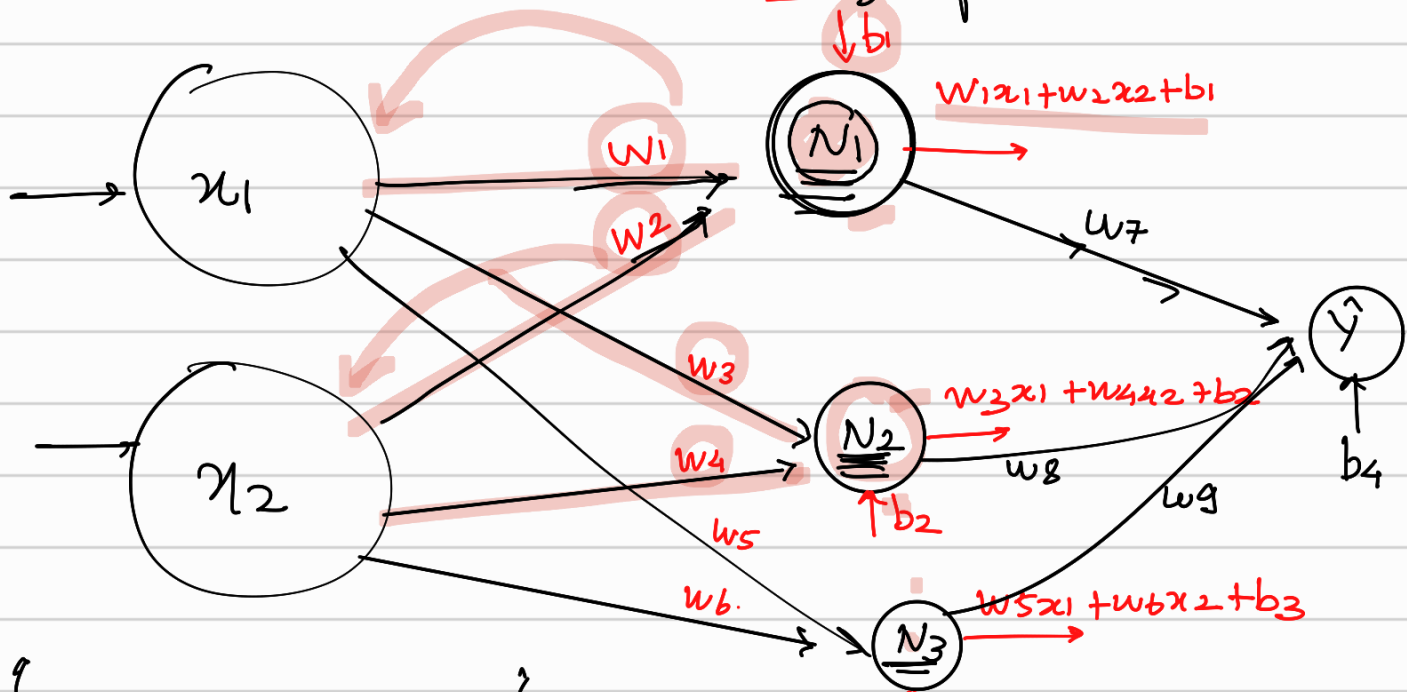
What is a Neuron

Logistic Regression



x_1	x_2	y

(1 HL) ↓ features



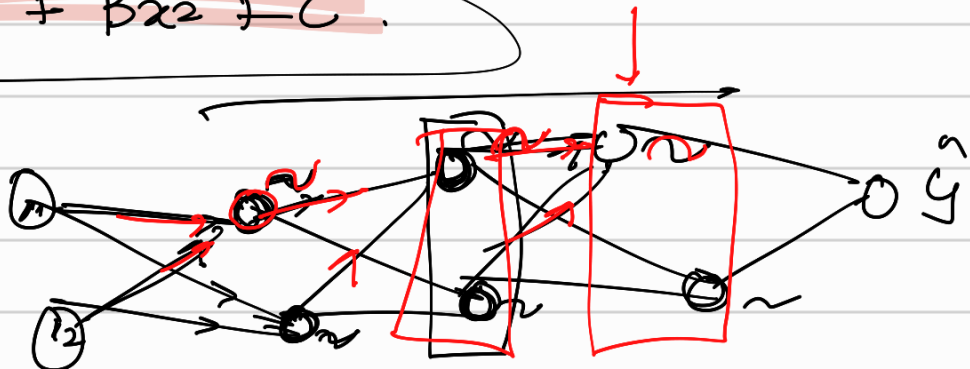
$$N_1 = (w_1 x_1 + w_2 x_2 + b_1), N_2 = (w_3 x_1 + w_4 x_2 + b_2), N_3 = (w_5 x_1 + w_6 x_2 + b_3)$$

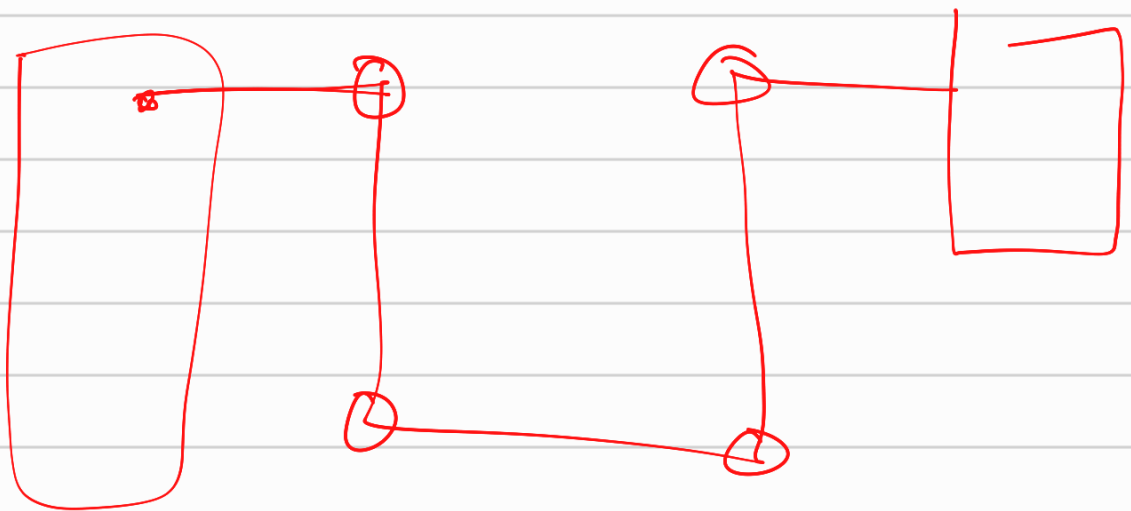
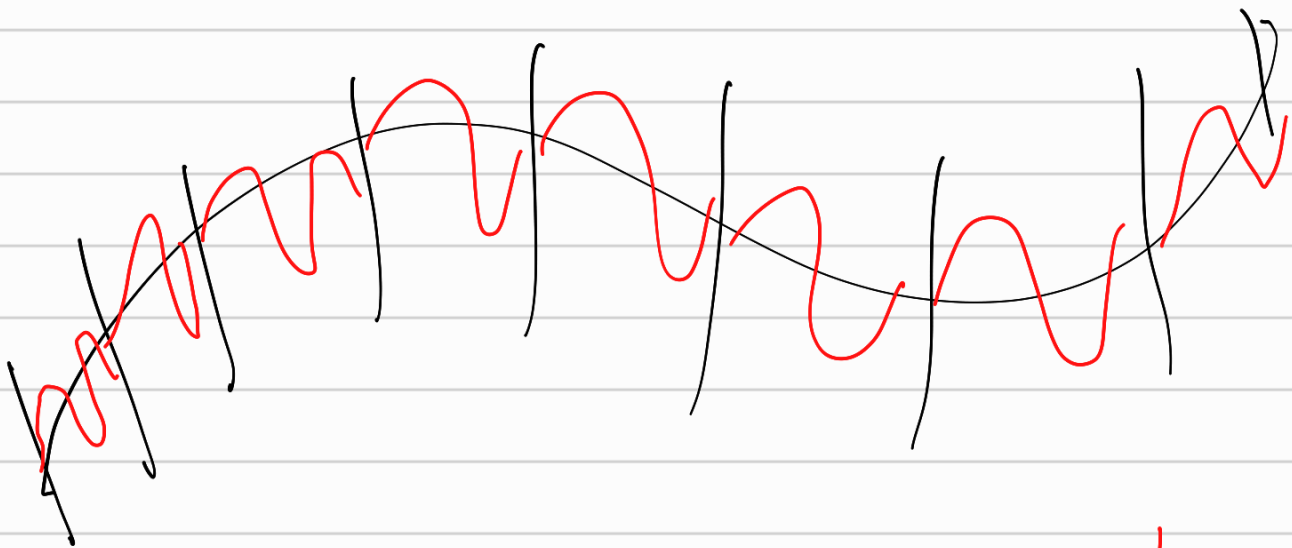
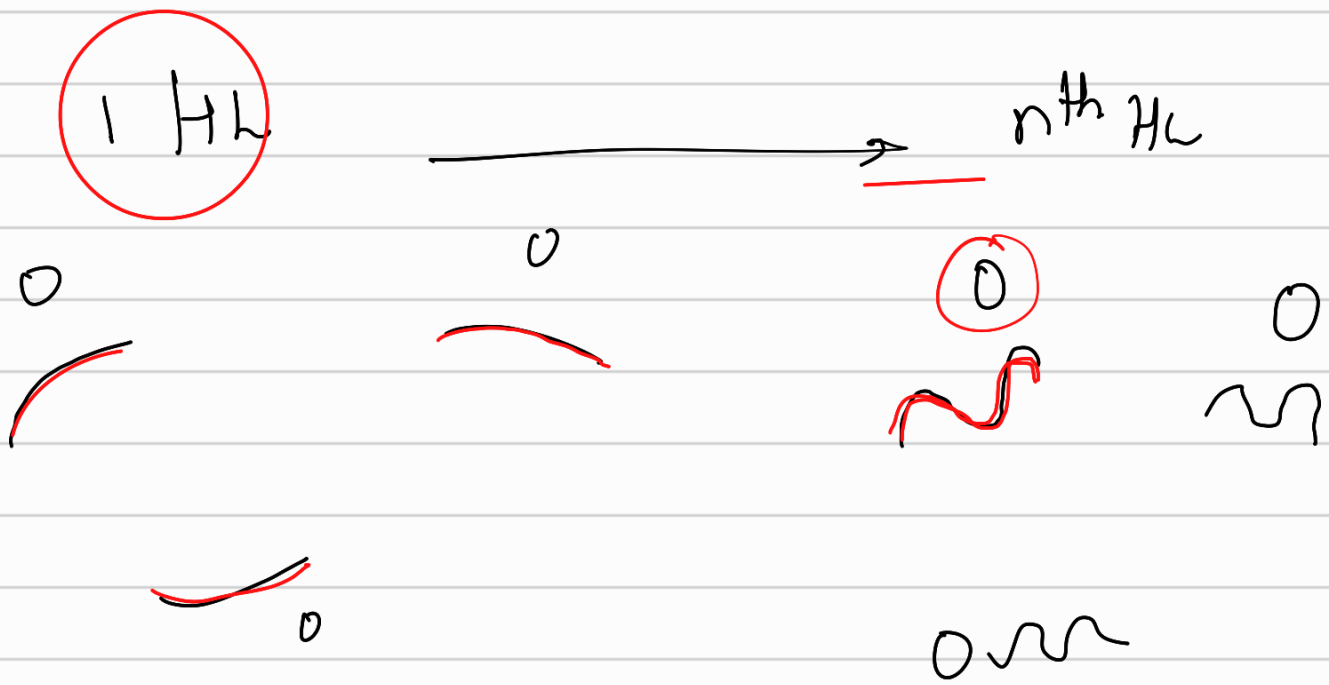
$$\hat{y} = w_7 N_1 + w_8 N_2 + w_9 N_3 + b_4$$

$$\Rightarrow \hat{y} = (w_7) (w_1 x_1 + w_2 x_2 + b_1) + w_8 (w_3 x_1 + w_4 x_2 + b_2) + w_9 (w_5 x_1 + w_6 x_2 + b_3) + b_4$$

$$\Rightarrow \hat{y} = x_1 (w_7 w_1 + w_8 w_3 + w_9 w_5) + x_2 (w_7 w_2 + w_8 w_4 + w_9 w_6) + w_7 b_1 + w_8 b_2 + w_9 b_3 + b_4$$

$$\Rightarrow \hat{y} = A x_1 + B x_2 + C$$





Gradient Descent



: minimise the loss by updating the w, b .
for all neurons.

Random