


Machine Learning

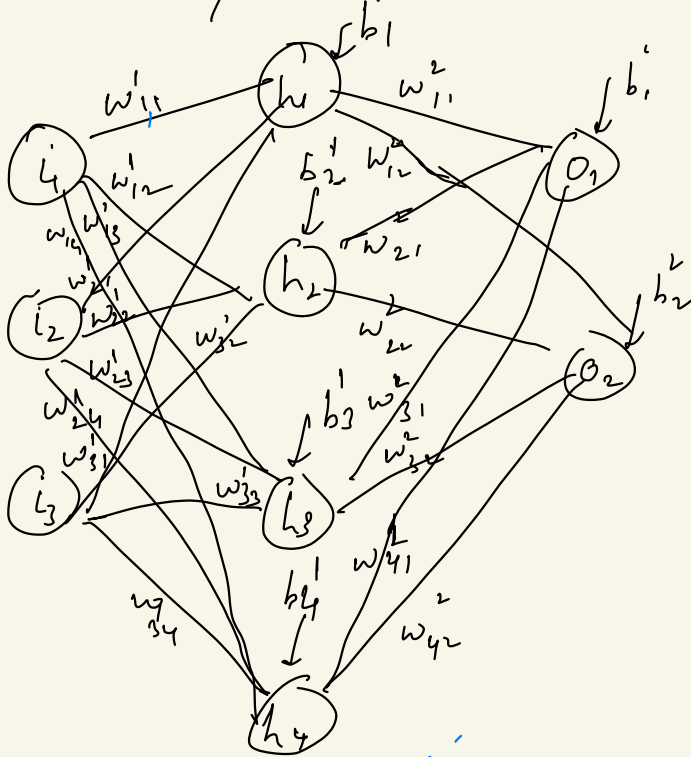
Lect . 6

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Multi Layer Perceptron

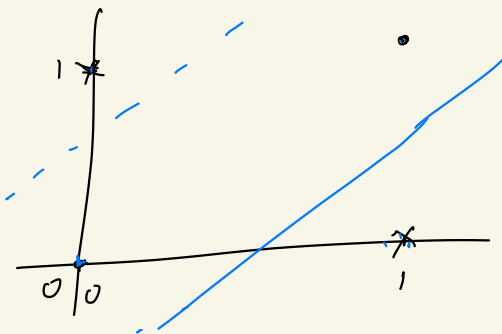
XOR problem



$$h_n = a \left(\sum_{k=1}^3 i_k w_{kn} + b_n \right)$$

$$a(x) = \text{sigmoid}(x) = \frac{1}{1 + e^{-x}}$$

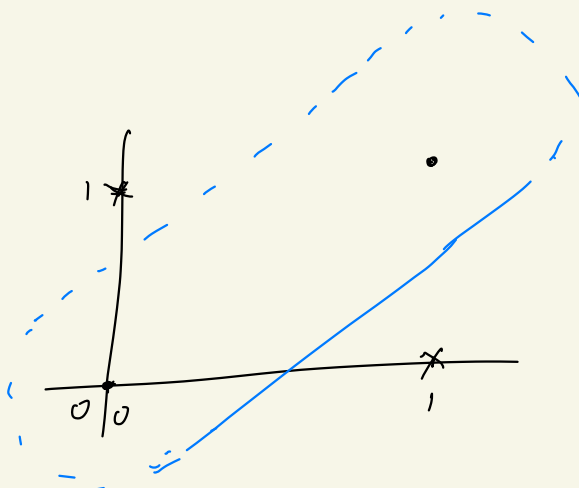
Universal approximation theorem



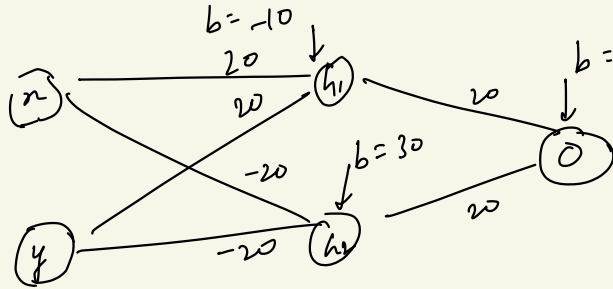
x_1	x_2	O/P y
0	0	0
0	1	1
1	0	1
1	1	0

How to separate this with a linear boundary

↓
Perceptron can't solve this
but MLP can!



⇒ Non linear separation

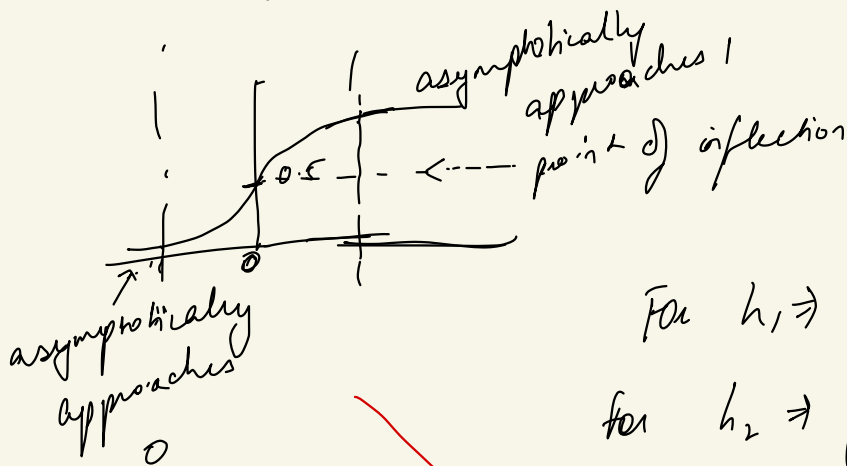


x_1	x_2	O/p \hat{y}
0	0	0.000045
0	1	0.99995
1	0	0.9995
1	1	0.000045

$$h_1 = \text{sigmoid}(20x + 20y - 10) = 0.5$$

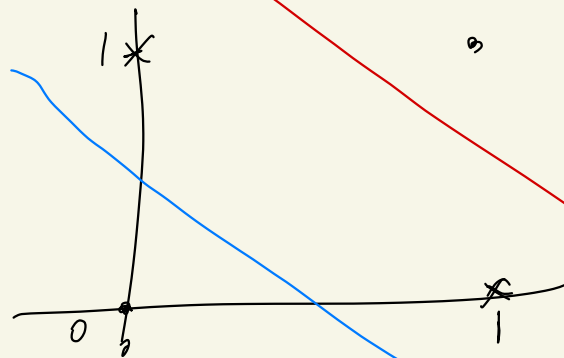
$$h_2 = \text{sigmoid}(-20x - 20y + 30) = 0.5$$

$$o = 20h_1 + 20h_2 - 30$$



For $h_1 \rightarrow y \approx 0.5 - x$

for $h_2 \rightarrow y \approx 1.5 - x$



the line if only h_1 is present
the line if only neuron h_2 is present
has 2 class division

with only these given set of weights you can learn such complex XOR functions

But how to learn these weights?

↓

Soln: Backpropagation

$$\sigma(20 \times 0 + 20 \times 0 - 10) \approx 0$$

$$\sigma(-20 \times 0 - 20 \times 0 + 30) \approx 1$$

$$\sigma(20 \times 0 + 20 \times 1 - 30) \approx 0$$

$$\sigma(20 \times 1 + 20 \times 1 - 10) \approx 1$$

$$\sigma(-20 \times 1 - 20 \times 1 + 30) \approx 0$$

$$\sigma(20 \times 1 + 20 \times 0 - 30) \approx 0$$

$$\sigma(20 \times 0 + 20 \times 1 - 10) \approx 1$$

$$\sigma(-20 \times 0 - 20 \times 1 + 30) \approx 1$$

$$\sigma(20 \times 1 + 20 \times 1 - 30) \approx 1$$

$$\sigma(20 \times 1 + 20 \times 0 - 10) \approx 1$$

$$\sigma(-20 \times 1 - 20 \times 0 + 30) \approx 1$$

$$\sigma(20 \times 1 + 20 \times 1 - 30) \approx 1$$