

From January 2020

1) Compute the matrix resulting from the convolution between the filter and the image in the figure, with a stride of 1 and no padding.

Image				Filter	
0	1	0	0	0	1
0	1	0	0	0	1
0	1	1	1		
0	1	0	0		

2) Build a Multi-Layer Perceptron able to classify the points: $\langle 1, 2, 0 \rangle$, $\langle 3, 0, 0 \rangle$, $\langle 1, 0, 1 \rangle$, $\langle 2, 1, 1 \rangle$, where the first two elements of each vectors are the feature values, and the last element is the class.

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3) Given the following dataset: $\{ \langle 0, -2, 1 \rangle, \langle 1, 0, 1 \rangle, \langle -1, 3, 1 \rangle, \langle 2, 3, 0 \rangle, \langle -1, 2, 0 \rangle \}$, where the last element of each vector is its class, construct a multi-layer perceptron that classifies the dataset, and draw its diagram (nodes and edges with corresponding weights).

4) Derive the update rule of the *output* neurons of a multi-layer perceptron according to the algorithm Backpropagation of Errors.

Tutorial 2

1)

0	0	1	0	0
1	0	1	0	0
0	1	0	0	0
0	1	0	0	0

\checkmark p_1	\checkmark p_2	\checkmark p_3
\checkmark p_4	p_5	p_6
p_7	p_8	p_9

$$p_1 = 0 \times 0 + 0 \times 1 + 1 \times 0 + 1 \times 1 = 1$$

0	0	1	0	0
0	0	1	0	0
0	1	1	1	0
0	1	0	0	0

$$p_2 = 0 \times 1 + 0 \times 0 + 1 \times 1 + 1 \times 0 = 1$$

0	1	0	0	0
0	1	0	0	0
0	1	1	1	0
0	1	0	0	0

$$p_3 = 0 \times 0 + 0 \times 0 + 0 \times 1 + 0 \times 0 = 0$$

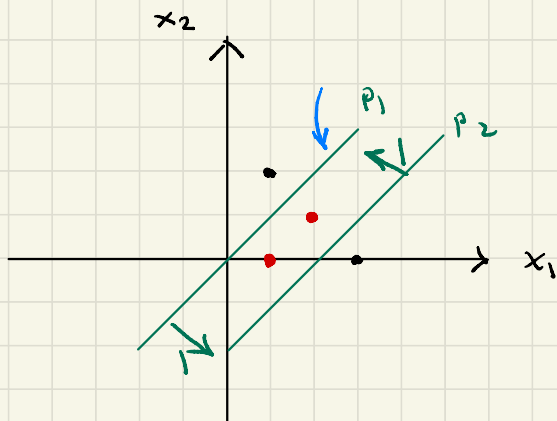
0	1	0	0	0
0	0	1	0	0
0	0	1	0	0
0	1	0	0	0

$$p_4 = 0 \times 0 + 0 \times 1 + 1 \times 0 + 1 \times 1 = 1$$

⋮

conv \rightarrow 4×4

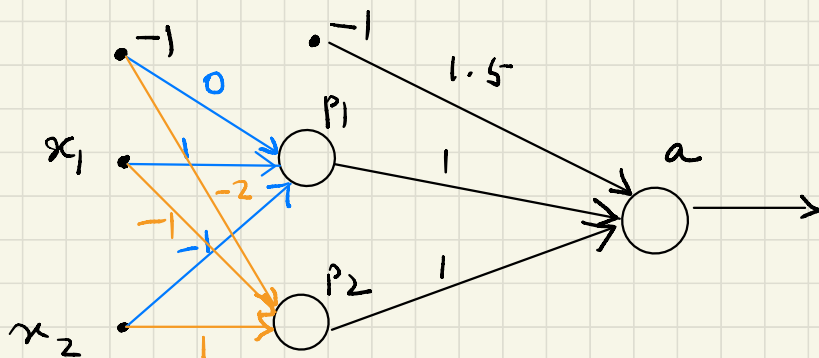
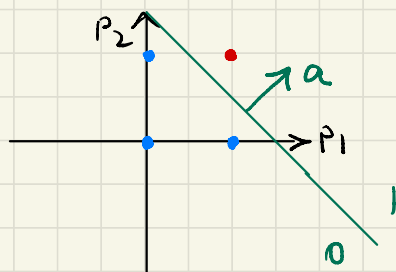
2.)



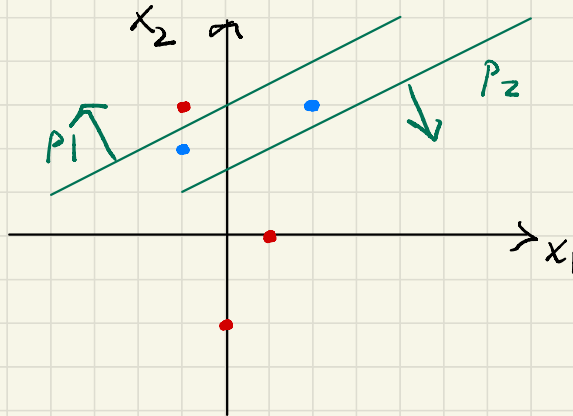
$$p_1: +1x_1 - 1x_2 + 0 = 0$$

$$p_2: -1x_1 + 1x_2 + 2 = 0$$

$$a: +1p_1 + 1p_2 - 1.5 = 0 \quad (a = p_1 \wedge p_2)$$



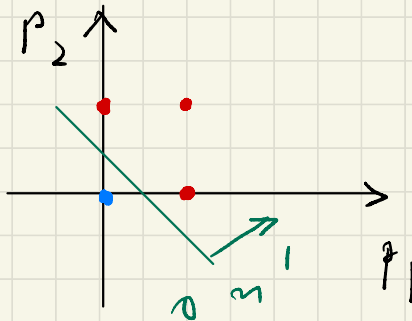
3)



$$p_1: -2x_1 + 1x_2 - 3 = 0 \quad \checkmark$$

$$p_2: +2x_1 - x_2 + 3/2 = 0 \quad \checkmark$$

$$n: -p_1 - p_2 + 0.5 = 0 \quad \left(n = p_1 \vee p_2 \right) \quad \text{or}$$



$$4) \quad E = \frac{1}{2} \left| y - t \right|^2 = \frac{1}{2} \sum_k (y_k - t_k)^2$$

$$y_k = b(a_k);$$

$$a_k = \sum_l \omega_{lk} z_l$$

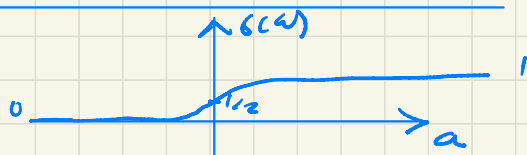
$$\frac{\partial E}{\partial \omega_{nk}} = \frac{\partial E}{\partial y_k} \cdot \frac{\partial y_k}{\partial a_k} \cdot \frac{\partial a_k}{\partial \omega_{nk}}$$

$$= \underbrace{(y_k - t_k) \cdot b'(a_k) \cdot (1 - b(a_k))}_{\delta_k} \cdot z_n$$

$$= \delta_k \cdot z_n$$

$$\omega_{nkc} \leftarrow \omega_{nk} - \eta \cdot \frac{\partial E}{\partial \omega_{nkc}}$$

$$b(a) = \frac{1}{1 + e^{-a}}$$



$$b'(a) = \frac{e^{-a}}{(1 + e^{-a})^2} = \frac{1}{1 + e^{-a}} \times \left(1 - \frac{1}{1 + e^{-a}}\right)$$

$$= b(a) (1 - b(a))$$