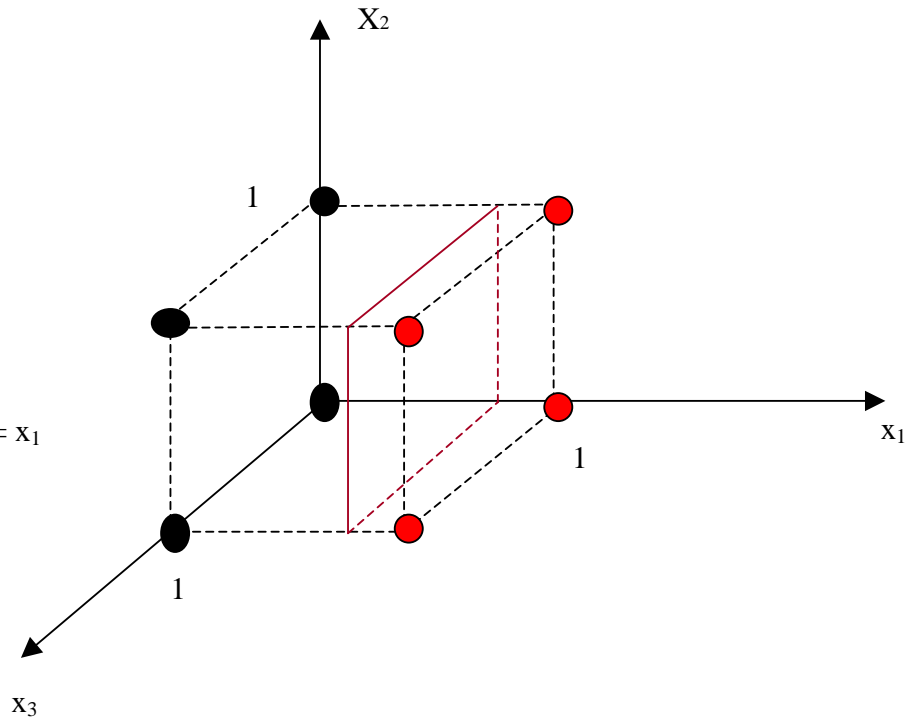
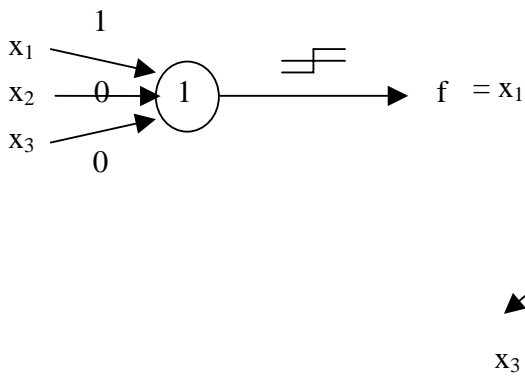


4. (Exercises 2.3) Indicate which of the following Boolean functions of 3 input variables can be realized by a single threshold element with weighted connections to the input.

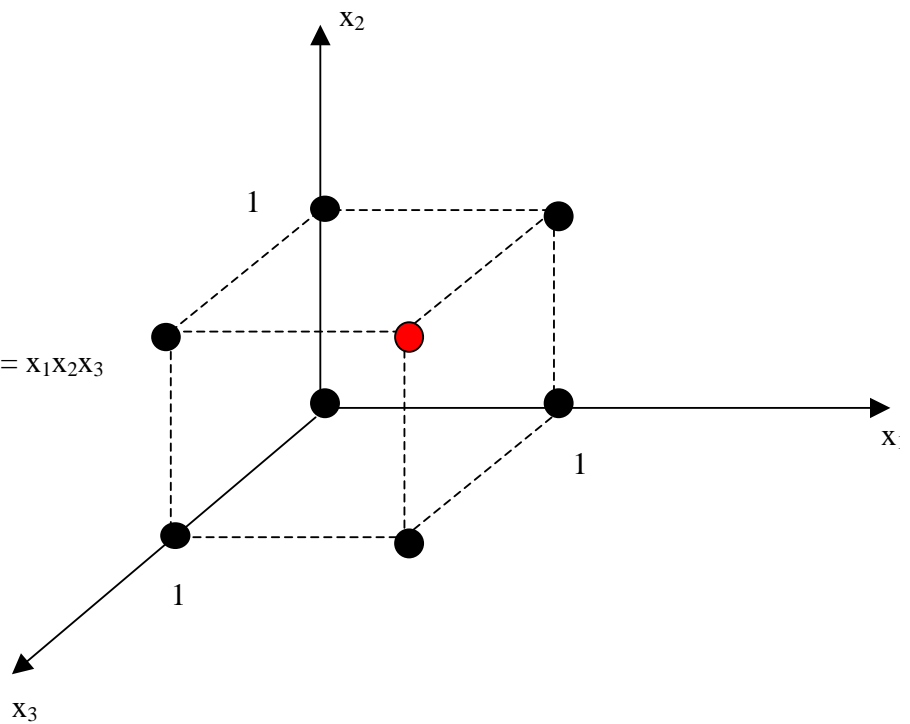
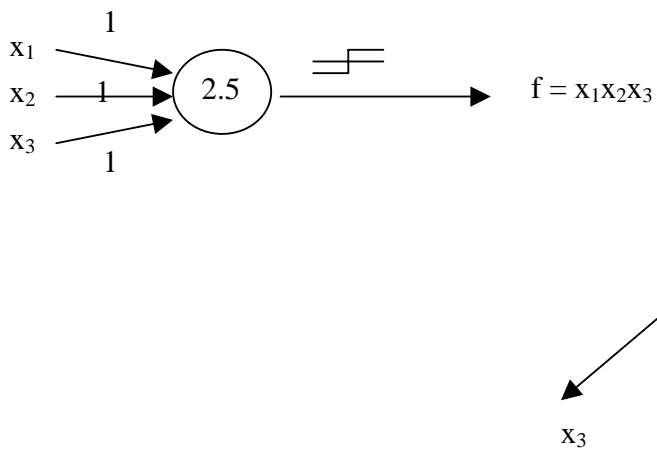
- (1).  $x_1$
- (2).  $x_1x_2x_3$
- (3).  $x_1+x_2$
- (4).  $(x_1x_2x_3) + (\bar{x}_1\bar{x}_2\bar{x}_3)$
- (5). 1

Solution:

(1) yes.



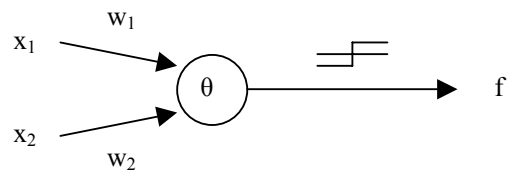
(2). Yes



- (3).  $x_1 + x_2$  (**Yes.**)  
 (4).  $(x_1 x_2 x_3) + (\bar{x}_1 \bar{x}_2 \bar{x}_3)$  (**No.**)  
 (5). 1 (**Yes.**)

## 5. Learning of A Single TLU

- TLU:



- Relation function:

$$f = 1 \quad \text{if} \quad (w_1 x_1 + w_2 x_2 - \theta) \geq 0$$

$$= 0 \quad \text{otherwise}$$

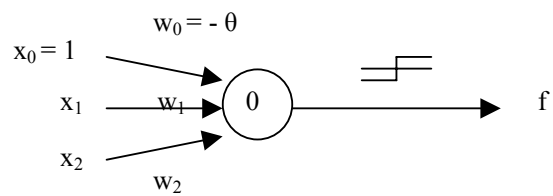
- Training set:

	$x_1$	$x_2$	$d$
$e_1$	5	1	0
$e_2$	2	1	0
$e_3$	1	1	1
$e_4$	3	3	1
$e_5$	4	2	0
$e_6$	2	3	1

- To learn:  $w_1, w_2, \theta$

### Solution 1:

(1)



$$f = 1 \quad \text{if} \quad (w_0 x_0 + w_1 x_1 + w_2 x_2) \geq 0$$

$$= 0 \quad \text{otherwise}$$

(2)

$$W^{new} = W^{old} + c (d - f) X$$

(3) Take  $c = 1$  and the initial weight values to be 0.

Iteration		$W^{old} = (w_0, w_1, w_2)$	$X = (x_0, x_1, x_2)$	$d$	$\Sigma$	$f$	$d = f ?$	$W^{new} = (w_0, w_1, w_2)$
1	$e_1$	(0, 0, 0)	(1, 5, 1)	0	0	1	no	(-1, -5, -1)
2	$e_2$	(-1, -5, -1)	(1, 2, 1)	0	-12	0	yes	(-1, -5, -1)
3	$e_3$	(-1, -5, -1)	(1, 1, 1)	1	-7	0	no	(0, -4, 0)
4	$e_4$	(0, -4, 0)	(1, 3, 3)	1	-12	0	no	(1, -1, 3)
5	$e_5$	(1, -1, 3)	(1, 4, 2)	0	3	1	no	(0, -5, 1)
6	$e_6$	(0, -5, 1)	(1, 2, 3)	1	-7	0	no	(1, -3, 4)
7	$e_1$	(1, -3, 4)	(1, 5, 1)	0	-10	0	yes	(1, -3, 4)
8	$e_2$	(1, -3, 4)	(1, 2, 1)	0	-1	0	yes	(1, -3, 4)
9	$e_3$	(1, -3, 4)	(1, 1, 1)	1	2	1	yes	(1, -3, 4)
10	$e_4$	(1, -3, 4)	(1, 3, 3)	1	4	1	yes	(1, -3, 4)
11	$e_5$	(1, -3, 4)	(1, 4, 2)	0	-3	0	yes	(1, -3, 4)
12	$e_6$	(1, -3, 4)	(1, 2, 3)	1	7	1	yes	(1, -3, 4)

(4). Stable. So we have :

$$w_0 = 1;$$

$$w_1 = -3;$$

$$w_2 = 4;$$

Since  $w_0 = -\theta$ , We got  $\theta = -1$ . The boundary line is:  $-3x_1 + 4x_2 + 1 = 0$ .

**Solution 2:**

(1)(2) are same with above.

(3) Take  $c=0.5$  and the initial weight values:  $w_0 = 0, w_1 = 1, w_2 = 1$

Iteration		$W^{old} = (w_0, w_1, w_2)$	$X = (x_0, x_1, x_2)$	$d$	$\Sigma$	$f$	$d = f ?$	$W^{new} = (w_0, w_1, w_2)$
1	$E_1$	$(0, 1, 1)$	$(1, 5, 1)$	0	6	1	no	$(-0.5, -1.5, 0.5)$
2	$E_2$	$(-0.5, -1.5, 0.5)$	$(1, 2, 1)$	0	-3	0	yes	$(-0.5, -1.5, 0.5)$
3	$E_3$	$(-0.5, -1.5, 0.5)$	$(1, 1, 1)$	1	-1.5	0	no	$(0, -1, 1)$
4	$E_4$	$(0, -1, 1)$	$(1, 3, 3)$	1	0	1	yes	$(0, -1, 1)$
5	$E_5$	$(0, -1, 1)$	$(1, 4, 2)$	0	-2	0	yes	$(0, -1, 1)$
6	$E_6$	$(0, -1, 1)$	$(1, 2, 3)$	1	1	1	yes	$(0, -1, 1)$
7	$E_1$	$(0, -1, 1)$	$(1, 5, 1)$	0	-4	0	yes	$(0, -1, 1)$
8	$e_2$	$(0, -1, 1)$	$(1, 2, 1)$	0	-1	0	yes	$(0, -1, 1)$
9	$e_3$	$(0, -1, 1)$	$(1, 1, 1)$	1	0	1	yes	$(0, -1, 1)$

Stable. So we have another set of weights:

$$w_0 = 0,$$

$$w_1 = -1,$$

$$w_2 = 1$$

then  $\theta = 0$  and the boundary is:  $-x_1 + x_2 = 0$

- If a training set of examples is linearly separable, then applying the perceptron weight updating rule can always converge to some solution (i.e., a set of weights) in a *finite number* of steps for any initial choice of weights.
- The exact number of steps needed depends on:
  - ✓ Initial weight values
  - ✓ Learning rate in weight updating rule
  - ✓ Order of presentation of training examples