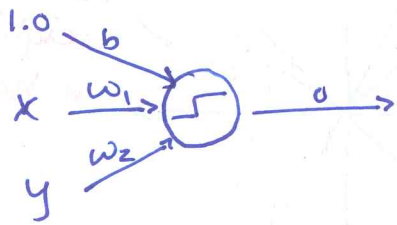
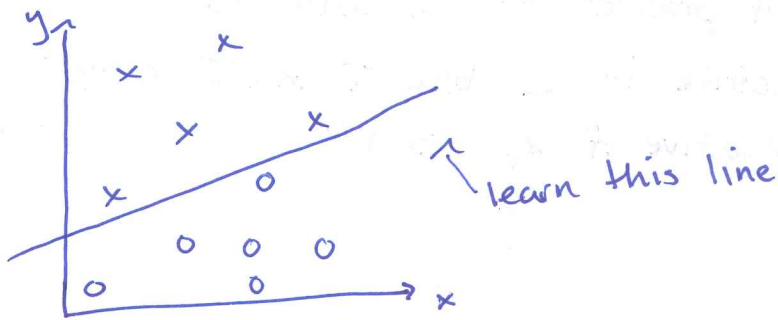


① Perceptron as classifier:



$$o = \begin{cases} 0 & \text{if } b + w_1 \cdot x + w_2 \cdot y \leq 0 \\ 1 & \text{if } b + w_1 \cdot x + w_2 \cdot y > 0 \end{cases}$$



$$p = \begin{bmatrix} 1.0 \\ x \\ y \end{bmatrix} \cdot [b \ w_1 \ w_2]$$

Whoops - this does outer product. Switch the two vectors

② Learning weights

input $i = (x_i, y_i)$ predicted ~~output~~ $p_i = \begin{bmatrix} 1.0 \\ x_i \\ y_i \end{bmatrix} \cdot [b_k \ w_{1k} \ w_{2k}]$

real output $i = o_i$

update weights:

$$b_{k+1} = b_k + (o_i - p_i) \cdot (1.0)$$

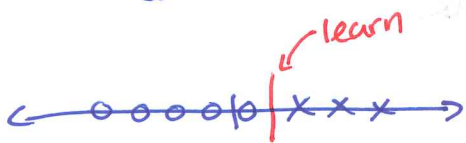
$$w_{1k+1} = w_{1k} + (o_i - p_i) \cdot x_i$$

$$w_{2k+1} = w_{2k} + (o_i - p_i) \cdot y_i$$

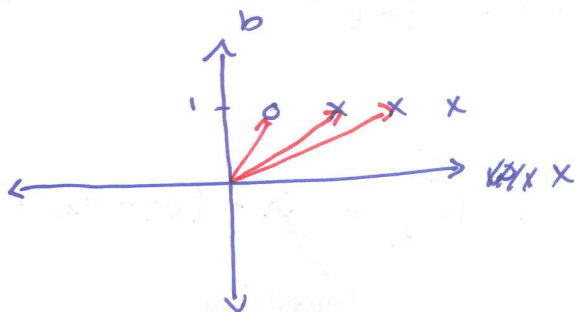
$$\bar{w}_{k+1} = \bar{w}_k + \text{error} \cdot \bar{x} \quad \leftarrow \text{move weights in direction of input if input is "true", away if input is "false"}$$

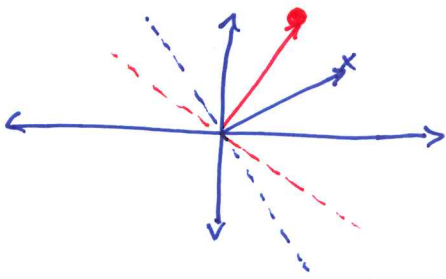
③ intuition

consider 1-D classifier

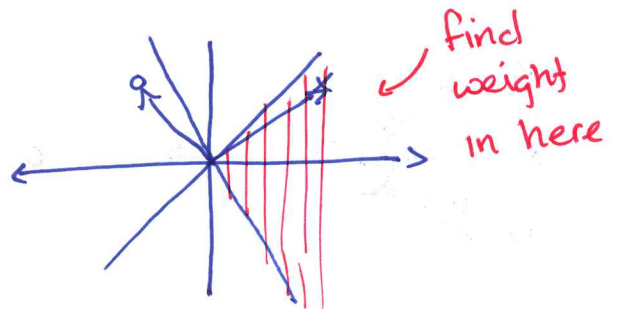
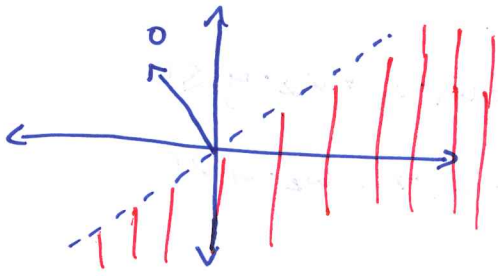


weight space

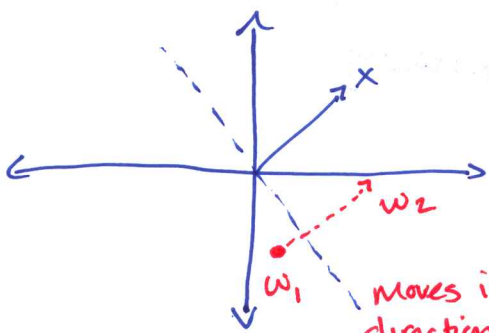




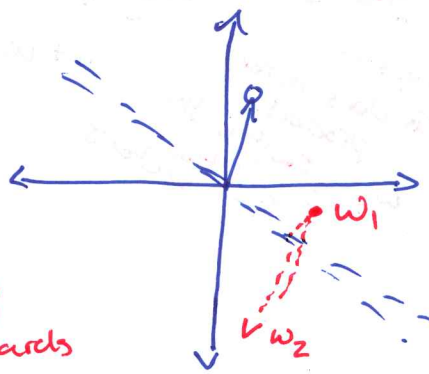
$p = \text{dot product of } \vec{x} \text{ and } \vec{w}$
 \Rightarrow positive if \angle b/w \vec{x} and $\vec{w} < 90^\circ$
 \Rightarrow negative if $\angle > 90^\circ$



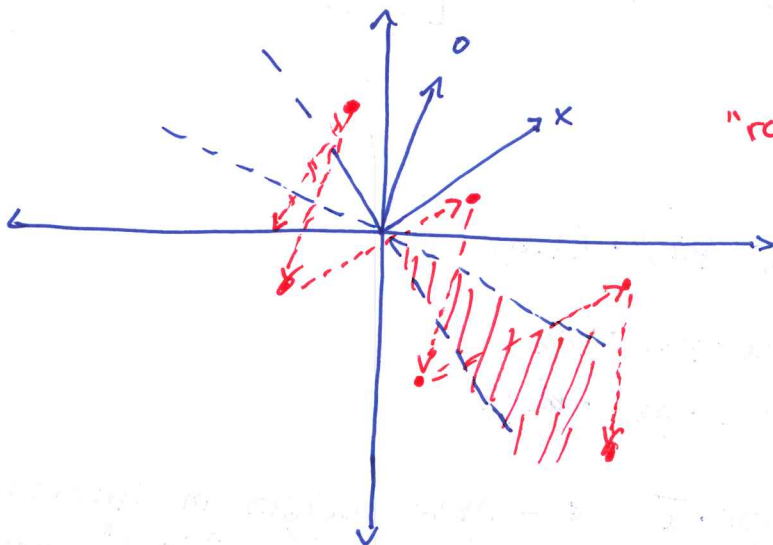
④ intuition for training:



moves in direction of input - towards "right" side



moves away from input - towards "right" side



"rolls downhill toward shaded valley"

⑤ why can't we learn XOR? not linearly separable!

⑥ more general activation fn: sigmoid:

$$f(x) = \frac{1}{1 + e^{-x}} \quad f'(x) = f(x) \cdot (1 - f(x))$$

training:

$$\vec{w}_{t+1} = \vec{w}_t + \alpha \cdot (o_i - p_i) \cdot f'(\vec{w}_t \cdot \vec{x}) \cdot \vec{x}$$

learning rate

gradient toward minimizing error

gradient descent