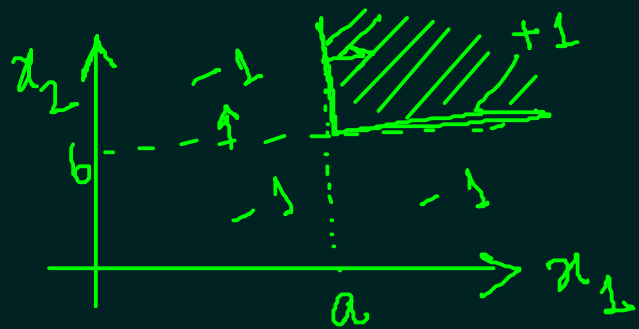


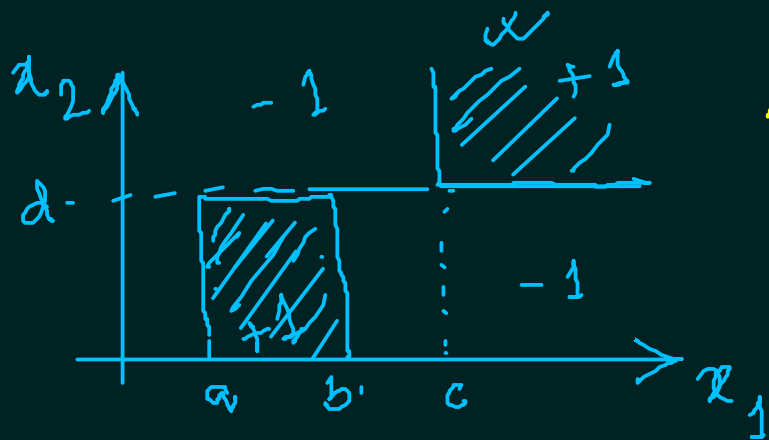
Classification Problems :

$$f: X \rightarrow Y = \{-1, +1\}$$

$$\text{Prob}(Y=1 | X) \text{ or } P(Y=0 | X)$$



Concept Learning
 $(x_1 > a) \wedge (x_2 > b)$



Decision Tree Learning

$$[(x_1 > c) \wedge (x_2 > d)] \vee [a \leq x_1 \leq b \wedge x_2 \leq d]$$

Ein \rightsquigarrow Exit



linear Separator
 (Hyper planes)

$$0.4 + 0.5x_1 - 0.6x_2 = 0$$

Linear Models
 (Primitive)

$$\begin{aligned} w_1x_1 + w_2x_2 + w_0 &= 0 \\ \updownarrow \end{aligned}$$

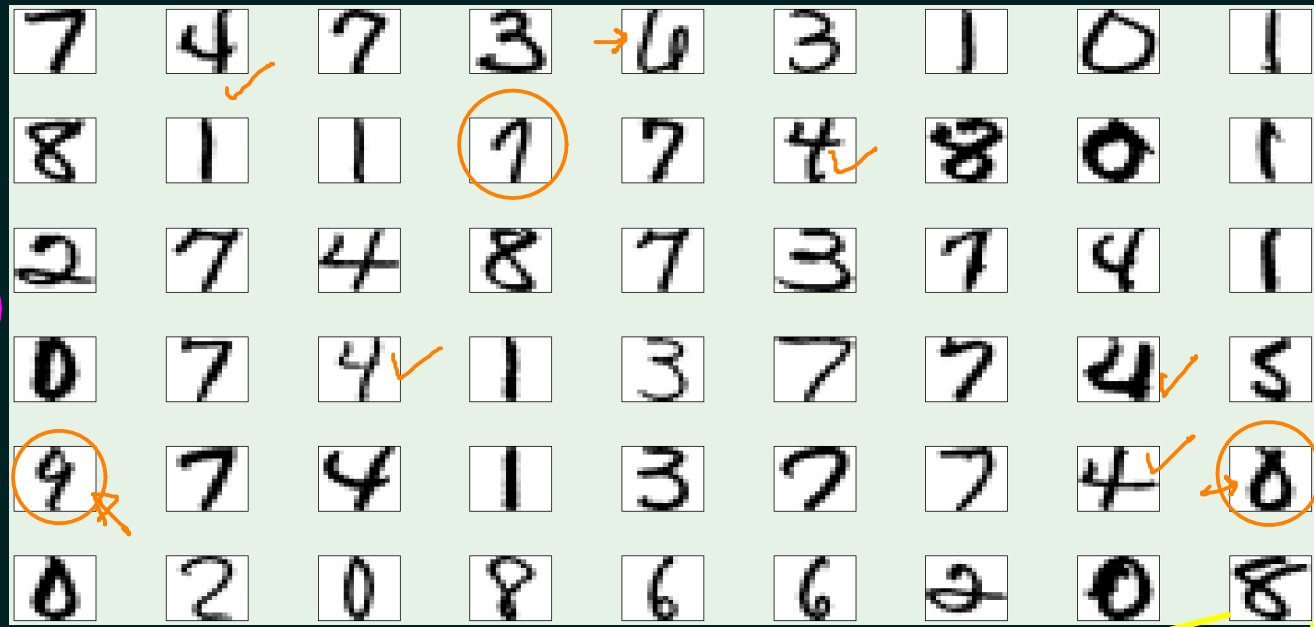
$$w_1, w_2, w_0$$

from TE

$$w_0 + w_1x_1 + \dots + w_dx_d = 0$$

$$X_{\text{new}} = \langle x_1^{\text{new}}, \dots, x_d^{\text{new}} \rangle \begin{cases} \text{for } + \text{ class } \geq 0 \\ \text{for } - \text{ class } < 0 \end{cases}$$

721302 ← Pincode



$(x_1 \dots x_{256})$

$$\sum_{i=0}^{256} w_i x_i = 0$$

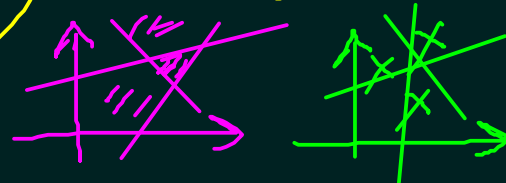
5 1

$x_0 = 1$ $\begin{cases} x_1 \rightarrow \text{intensity} \\ x_2 \rightarrow \text{Symmetry} \end{cases}$

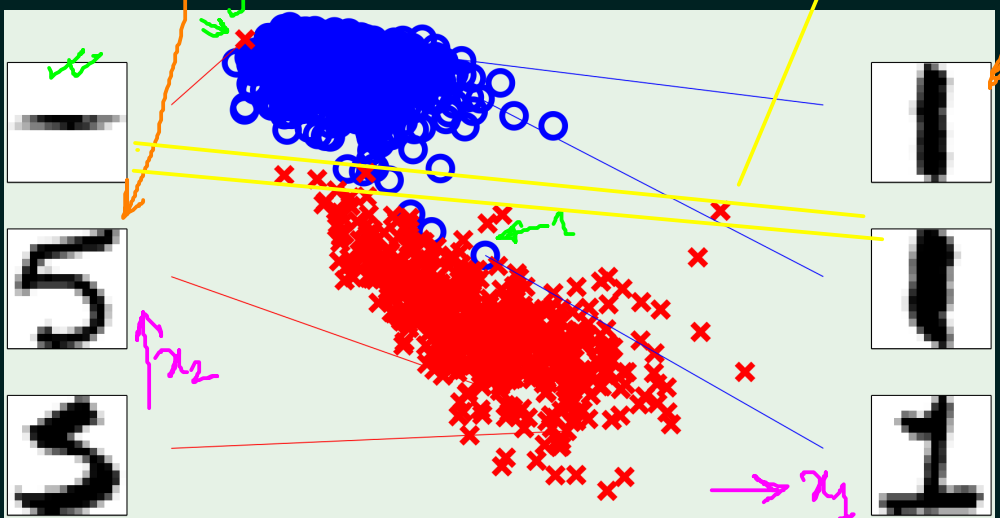
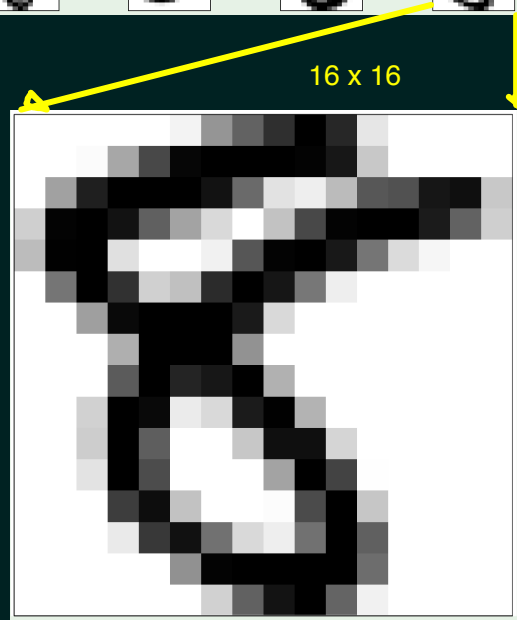


$$w_0 + w_1 x_1 + w_2 x_2 = 0$$

Separator



Symmetry



Perceptions

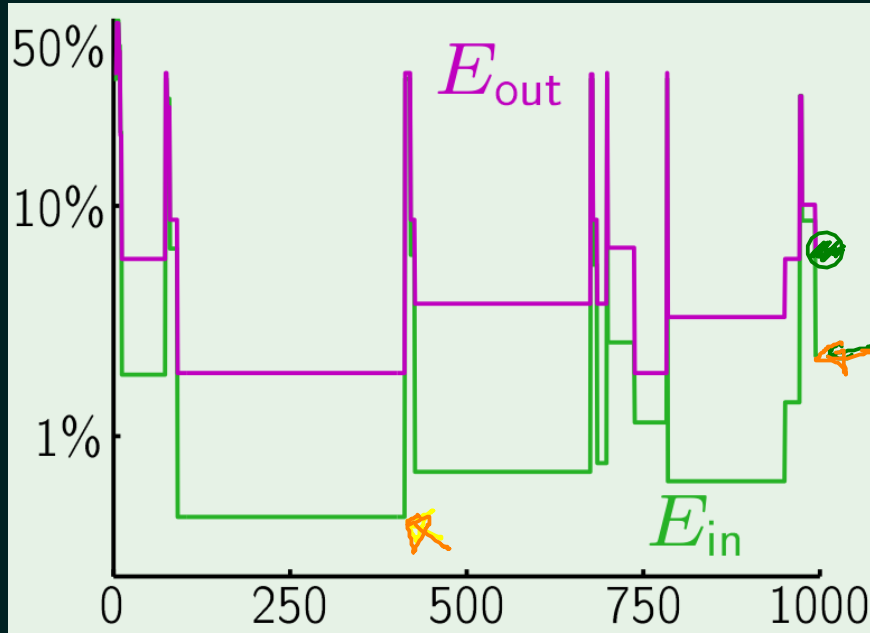
$$s = \sum_{i=0}^d w_i d_i = 0$$

$\begin{matrix} TE_1 \\ TE_2 \\ \vdots \\ TE_N \end{matrix}$ } batch itr

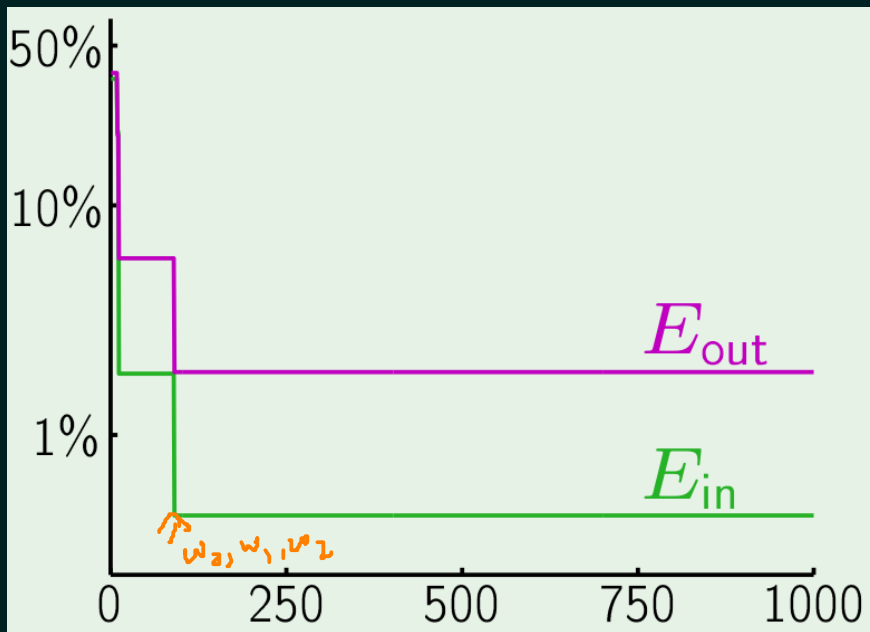
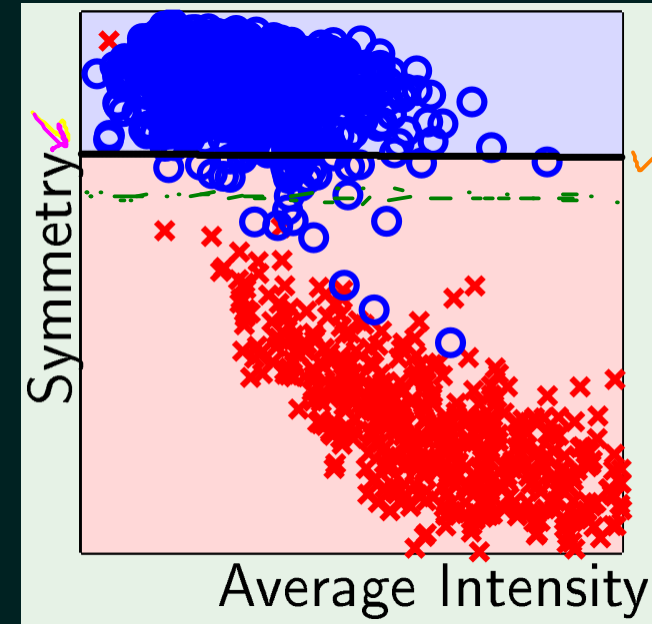
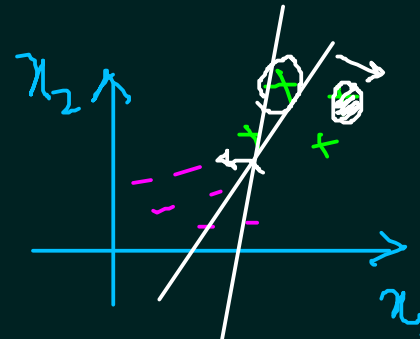
$E_{in} \approx E_{out}$
(assumed)

TE₁ → +
(-)

LA → batch
→ Itr ✓

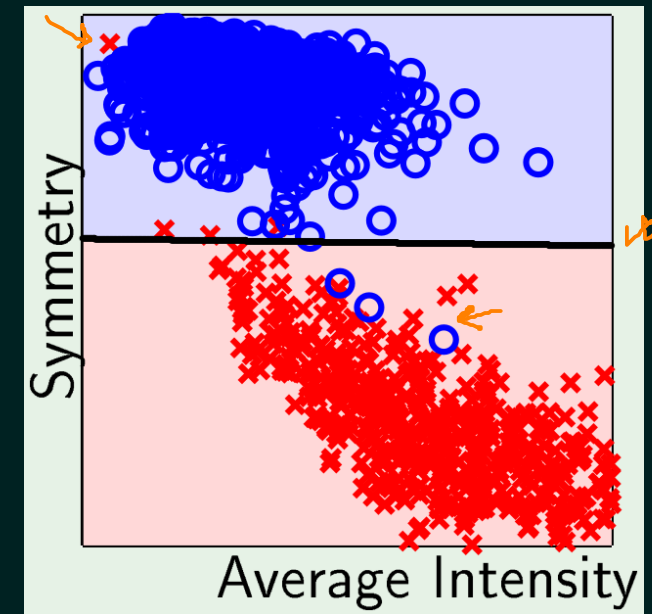


Perceptron Learning Algo

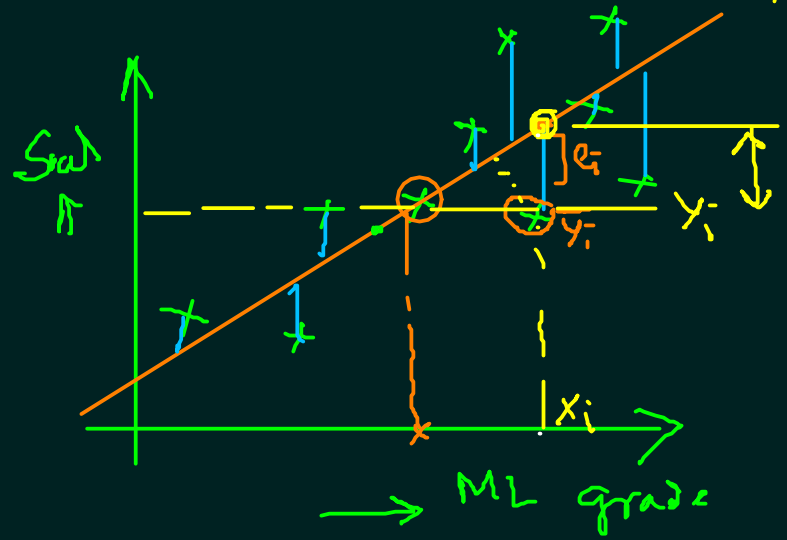


'Pocket' Learning Algo

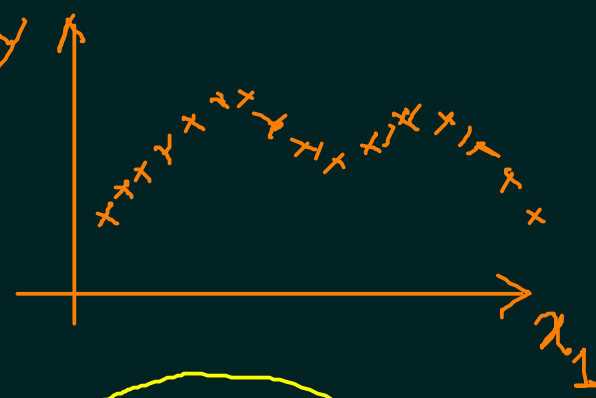
5 and 6
[2.5% error]



Credit Approval $\begin{cases} \rightarrow \text{yes} \\ \rightarrow \text{No} \end{cases}$ } classify $\frac{y}{TW}$
 Credit limit \rightarrow o/p = real value (REGRESSION)



linear Regression y



TE: (N)

$$x_i = [(x_{i1} \dots x_{id}), y_i]$$

$$y_i = \sum_{i=0}^d w_i x_i$$

$$\Rightarrow W^T \cdot x = y$$

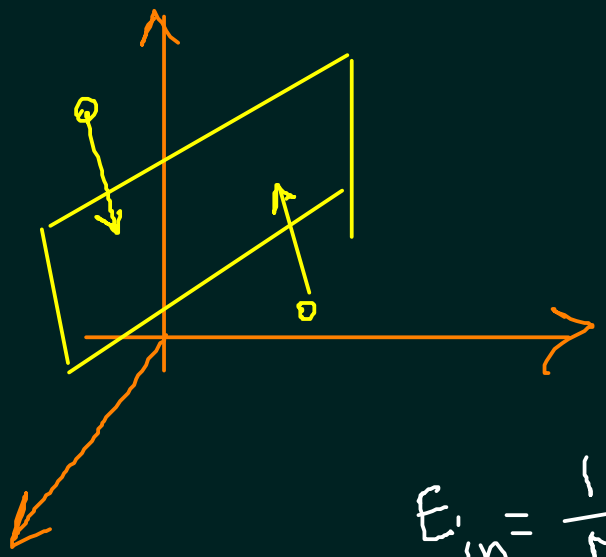
$$W = \begin{bmatrix} w_0 \\ w_1 \\ \vdots \\ w_d \end{bmatrix} \quad x = \begin{bmatrix} x_0 \\ x_1 \\ \vdots \\ x_d \end{bmatrix}$$

$(d+1) \times 1$ $(d+1) \times 1$

$$e_i = (y_i - \sum_{i=0}^d w_i x_i)$$

minimize for all TEs

$$E_{min} = \frac{1}{N} \sum e_i^2 \rightarrow \text{minimize (Sum-Sq. error)}$$



$$E_{in}(w) = \frac{1}{N} \sum_{n=1}^N (w^T x_n - y_n)^2 = \frac{1}{N} \|Xw - y\|^2 \quad w = \begin{bmatrix} w_0 \\ \vdots \\ w_d \end{bmatrix}$$

$$(Xw - y)^T (Xw - y) \quad x_i = \begin{bmatrix} x_{i0} \\ x_{i1} \\ \vdots \\ x_{id} \end{bmatrix} \quad X = \begin{bmatrix} -x_1^T \\ -x_2^T \\ \vdots \\ -x_N^T \end{bmatrix} \quad y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_N \end{bmatrix}$$

$(d+1) \times 1$ $(d+1) \times N$ $N \times (d+1)$ $N \times 1$

$$\nabla E_{in}(w) = 0$$

$$\Rightarrow \nabla E_{in}(w) = \frac{2}{N} X^T (Xw - y) = 0 \Rightarrow$$

$$(AB)^T = B^T A^T$$

$$\Rightarrow (X^T X)w = X^T y$$

$$\Rightarrow w = (X^T X)^{-1} X^T y$$

$(d+1) \times 1$ $(d+1) \times N$ $N \times (d+1)$ $(d+1) \times N$

$$w = X' y$$

$$X' = (X^T X)^{-1} X^T$$

$(d+1) \times 1$ pseudo inverse

Linear Regression Algo

$$X' X = (X^T X)^{-1} (X^T X) = I$$

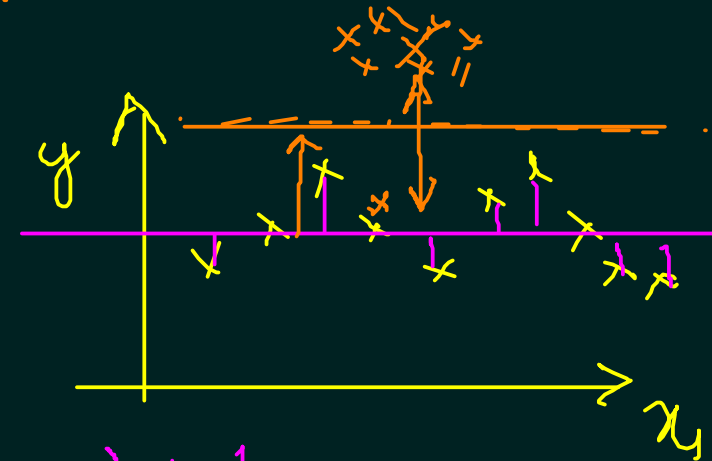
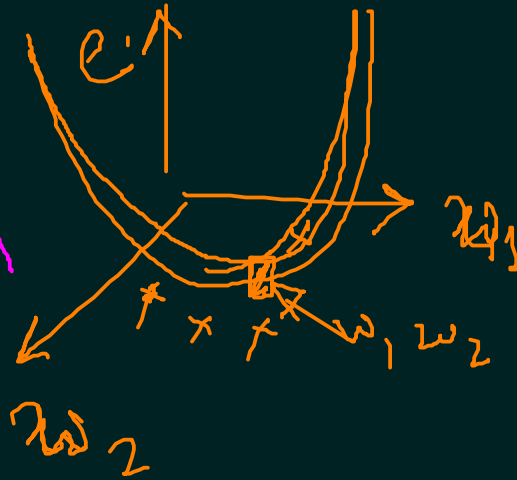
Linear Regression:

$$h(x) = y \rightarrow \text{Real value}$$

Can I use it for classification?

$\{-1, +1\}$ Classification

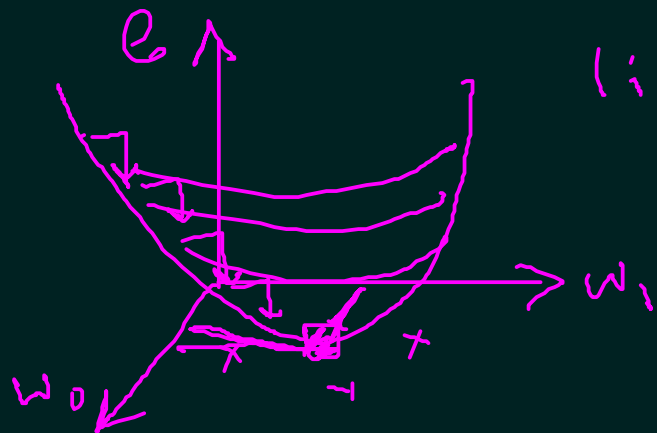
(Yes)



Linear Regression Boundary

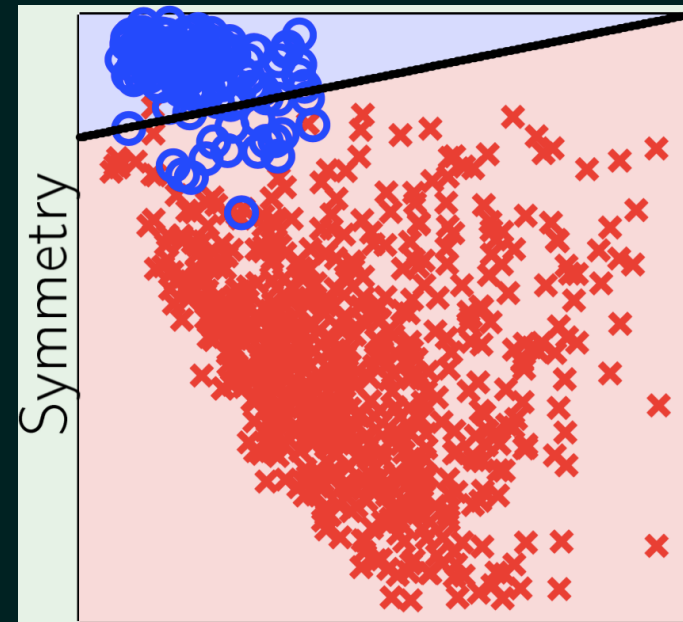
$$w_1 x_1 + w_2 x_2 + w_0 = 0$$

linear w.r.t. w



gradient descent

batch stochastic



Average Intensity