

Q74

① Conceptual

⇒ 2. Ahead of time (learnability, ...)

3. Details / specific points

⇒ ④ Practical aspects

Conceptual issues

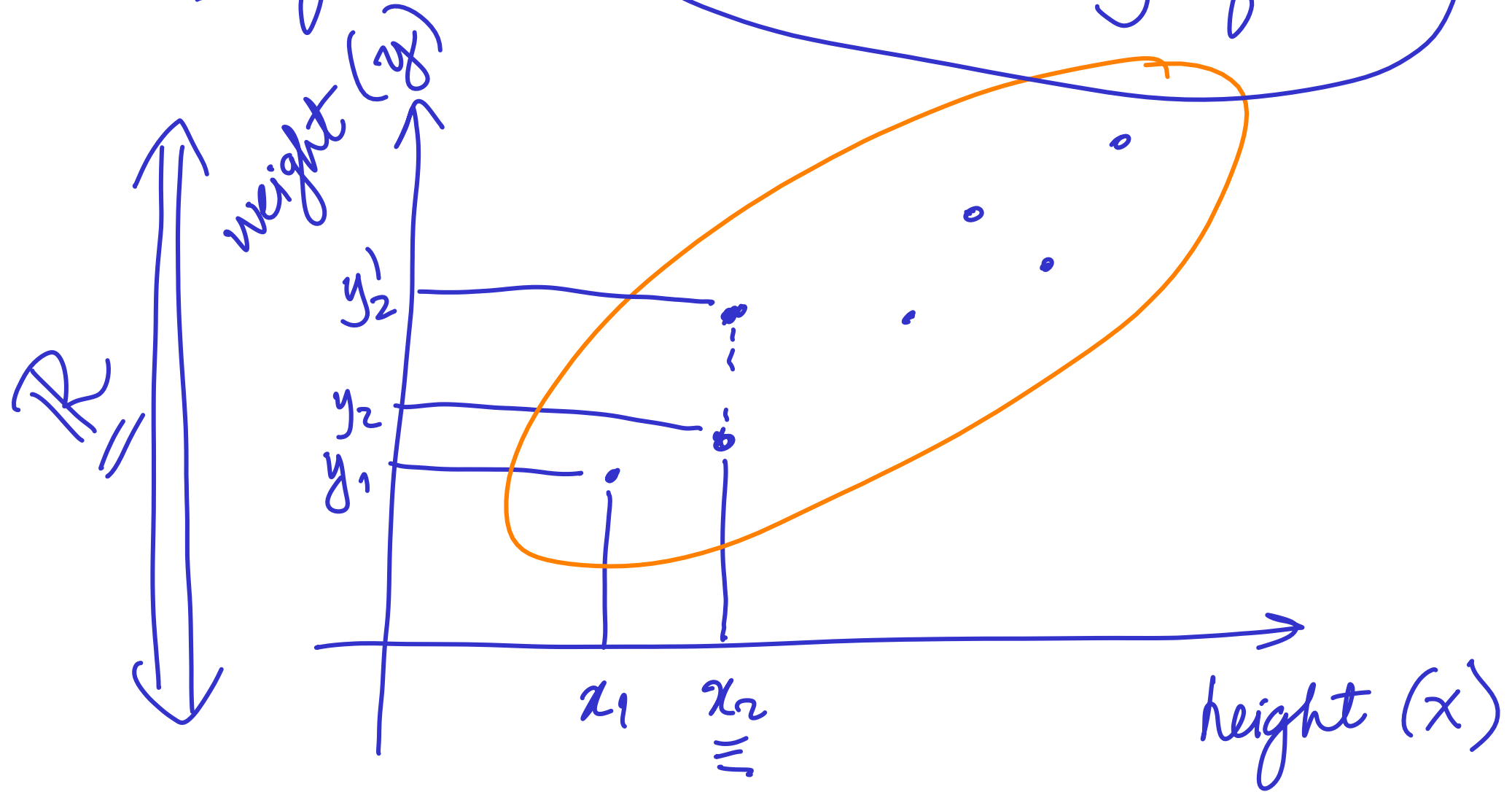
- ① Perception, Linear regression, Logistic regression

Regression Problem
X

Classification Problem.

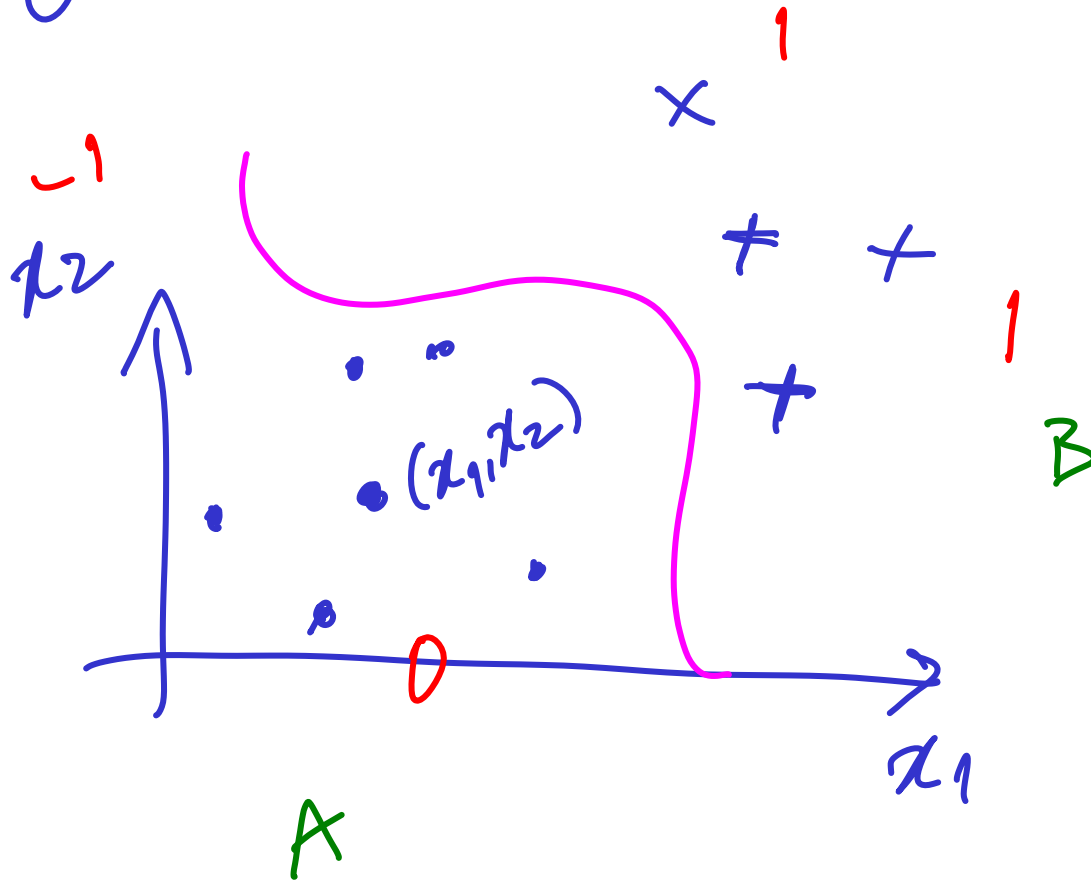
Regression

Find f s.t.
 $y \approx f(x)$



Classification.

y is class label.



Cost Function

$$J(w) = \frac{1}{N} \sum_{n=1}^N \underbrace{(y^{(n)} - \hat{y}^{(n)})^2}_{\text{(Reg)}} \quad (\text{Reg})$$

binary classif.

$$J(w) = \frac{1}{N} \sum_{n=1}^N y^{(n)} \ln(\hat{y}^{(n)}) + (1 - y^{(n)}) \ln(1 - \hat{y}^{(n)})$$

② Function optimization

$w^* \leftarrow$ optimal weight.

Gradient descent \rightarrow a tool

* for optimizing functions.

Learning Model

Target (?)

distribution $P(x, y)$
(unknown)

Observations
(examples)

x, y

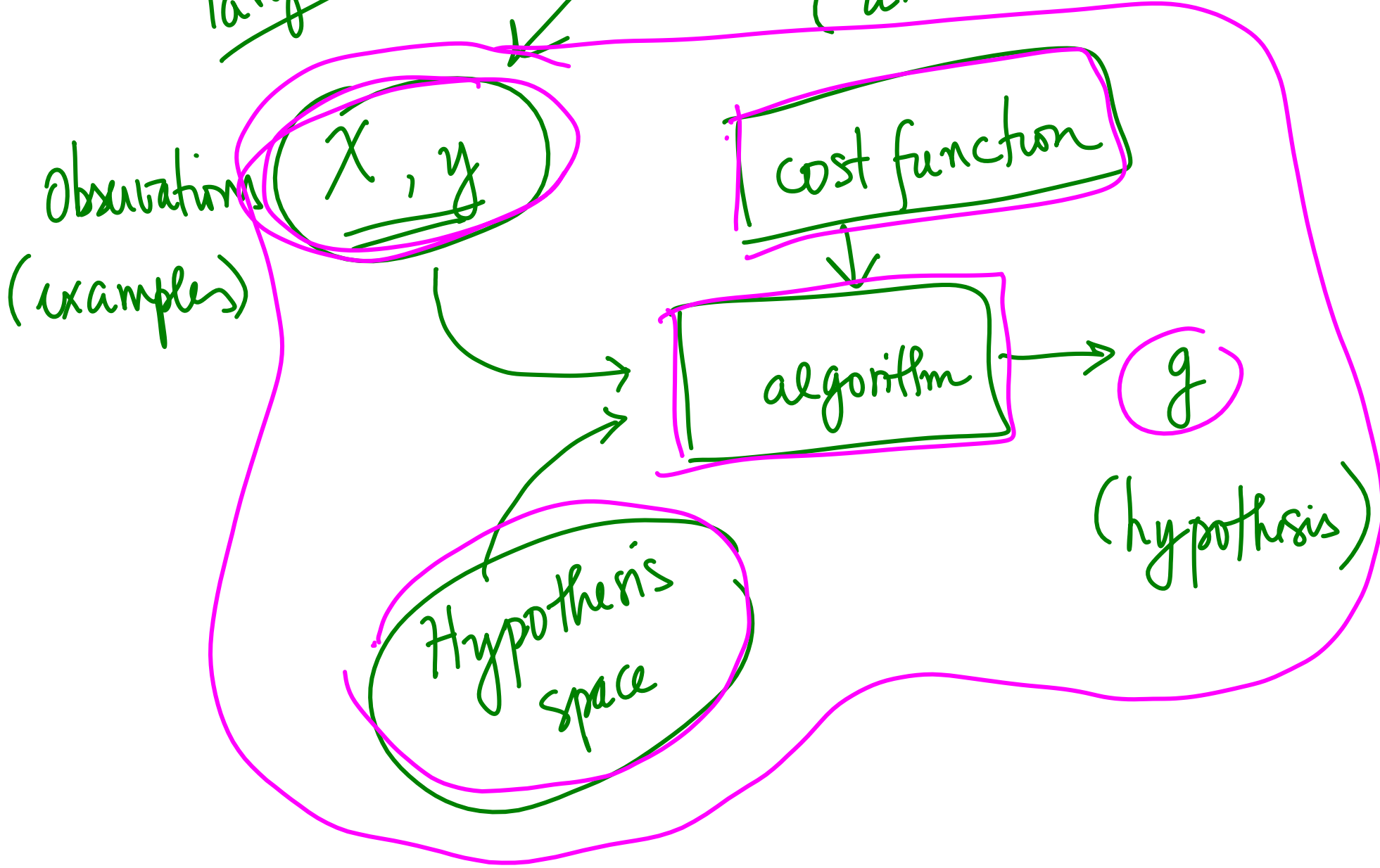
cost function

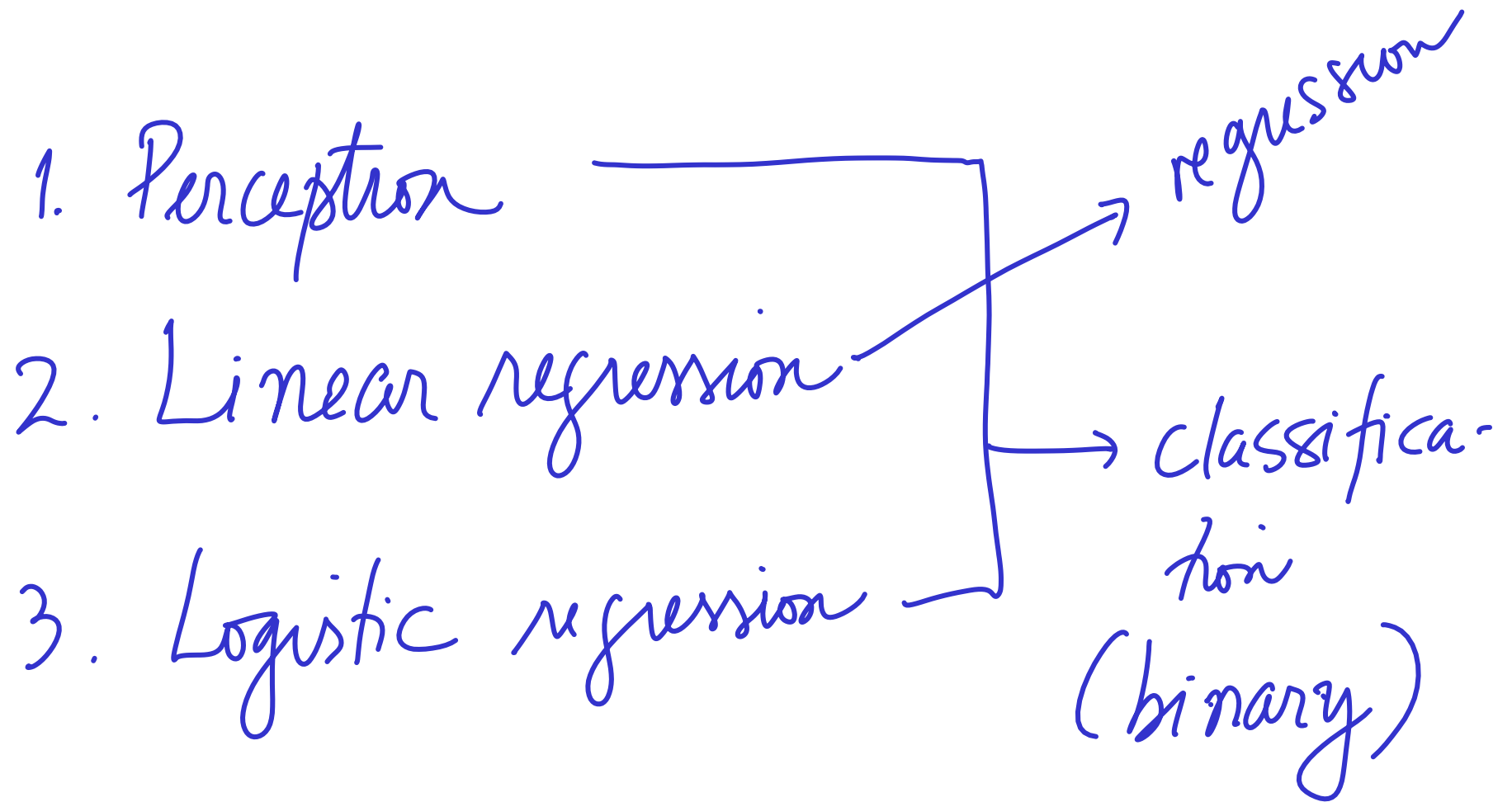
algorithm

g

(hypothesis)

Hypothesis
space

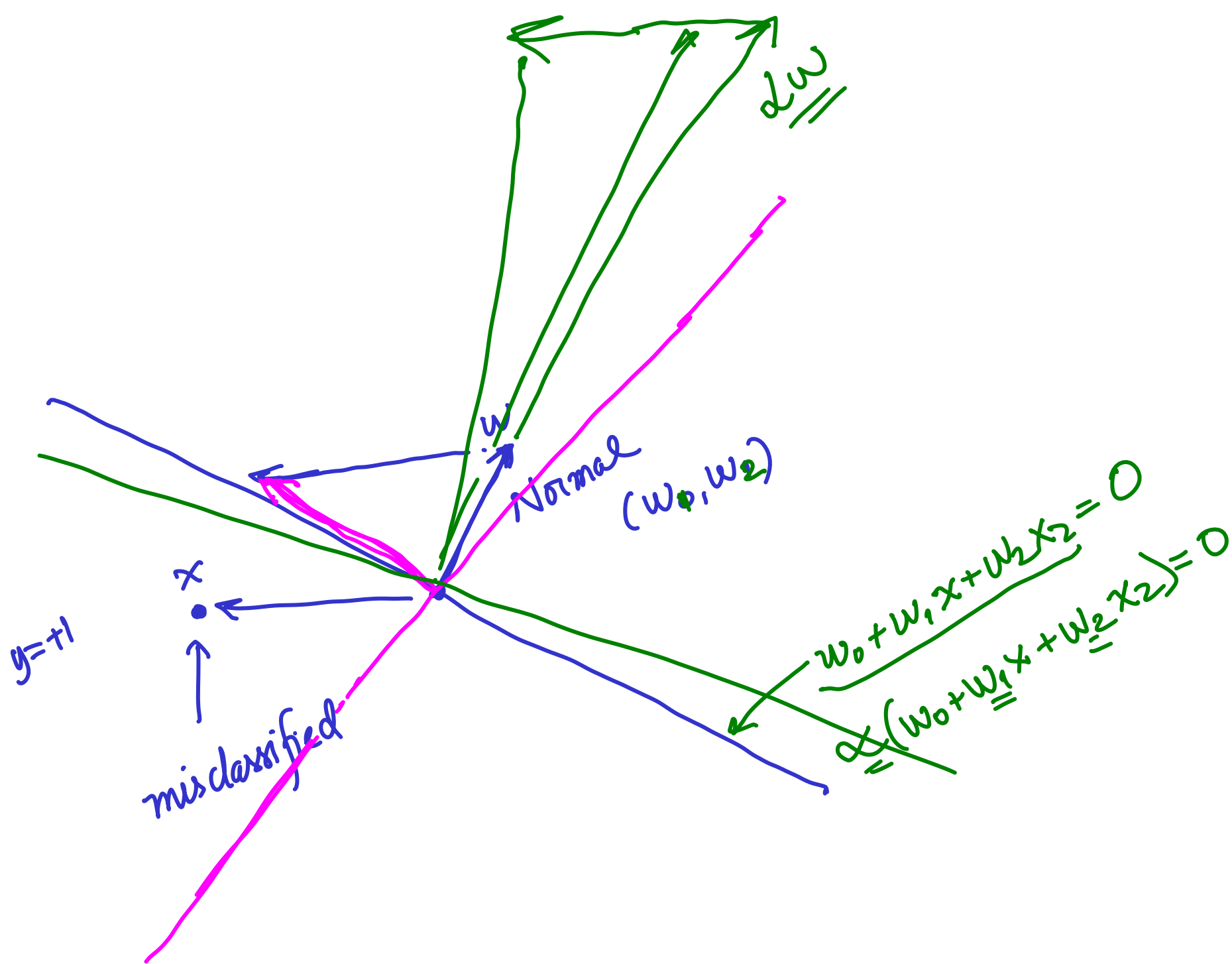


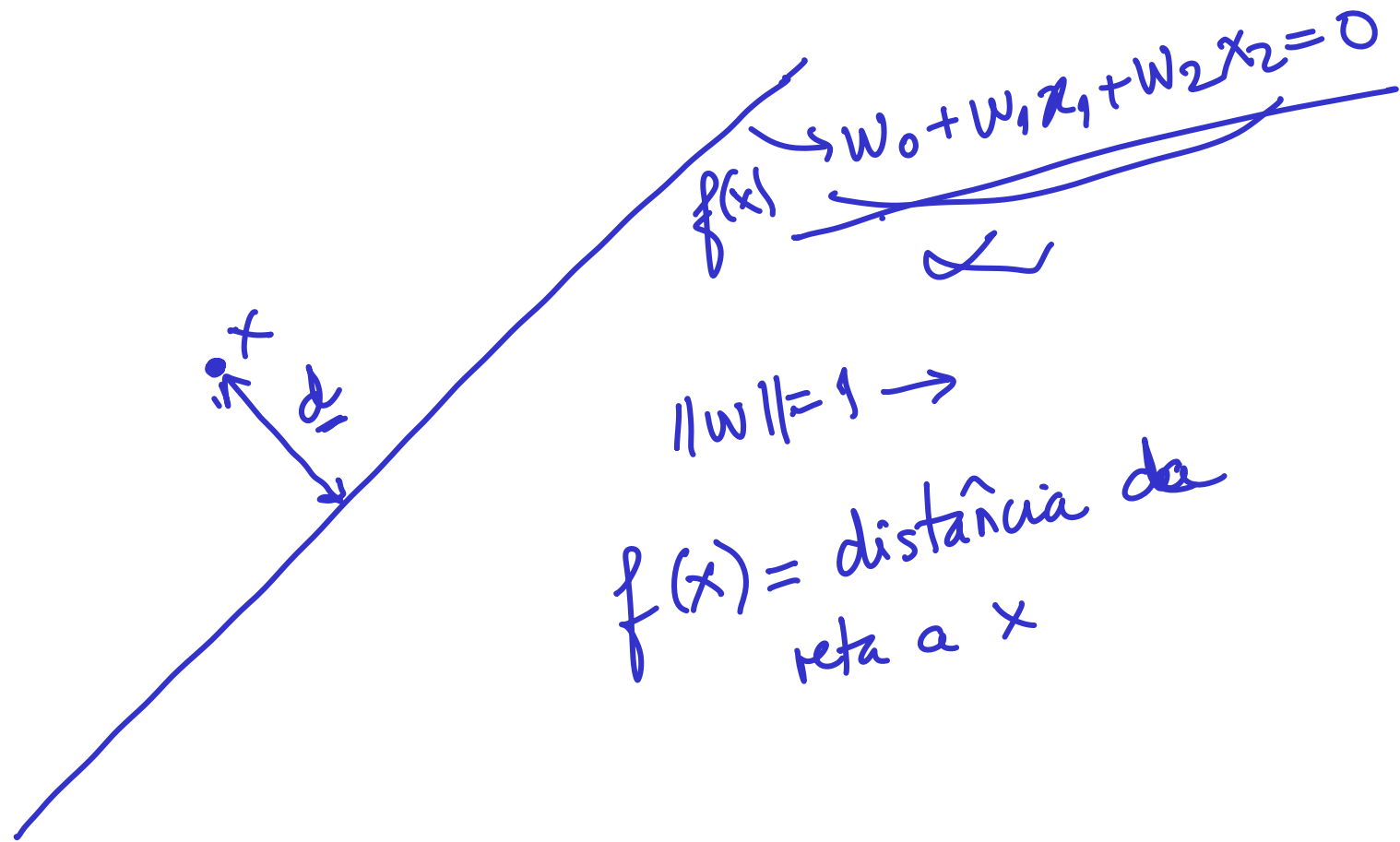


Perceptron

$$w \leftarrow w + yx$$

↑
misclassified





$$\|w\| = 1 \rightarrow$$

$f(x) = \text{distância da}$
 $\text{reta a } x$

$$W \leftarrow W + \eta y x$$

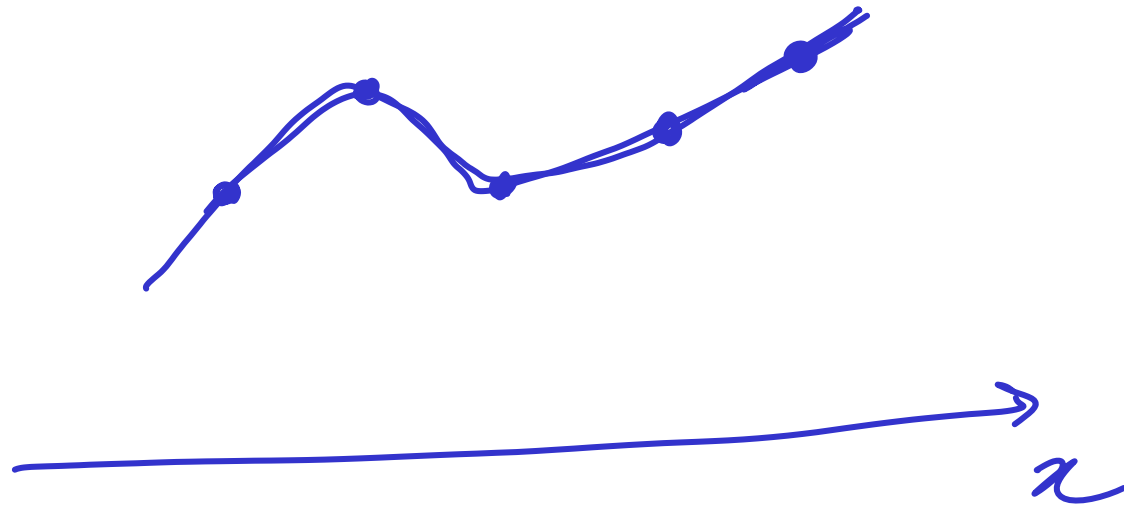
learning rate.

accelerate convergence?

η

Linear regression .

interpolation \times regression



$$h(x) = w^T x$$

$$= \underline{w_0 + w_1 x}$$

$$\S \quad h(x) = \underline{w_0 + w_1 x + w_2 x^2}$$

$$\Rightarrow \boxed{J(w) = \frac{1}{N} \sum_i \left(y^{(n)} - \underbrace{h(x^{(n)})}_{\hat{y}^{(n)}} \right)^2}$$

Logistic Regression

(x, y)



$\{0, +1\}$

Cross-entropy loss

$$J(w) = \frac{1}{N} \sum y \ln(\hat{y}) + (1-y) \ln(1-\hat{y})$$

Target



$$f(x) = P(y=+1/x)$$



$$h(x) = \theta(\underline{w^T x})$$



$$\begin{aligned} \Theta(w^T x) &\rightarrow 1 \\ w^T x &\rightarrow +\infty \end{aligned}$$

$$h(x) > 0.5$$

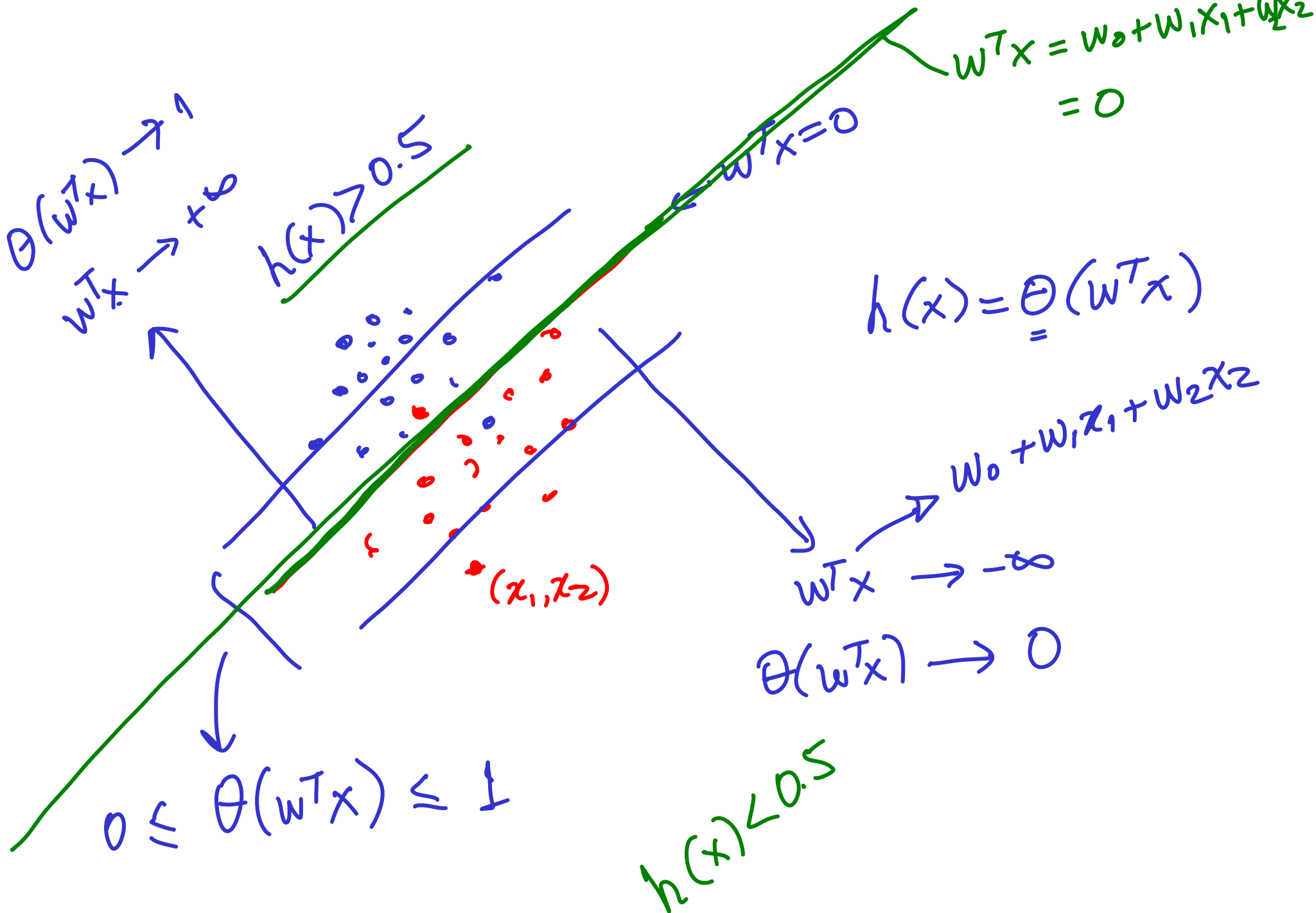
$$h(x) = \Theta(w^T x)$$

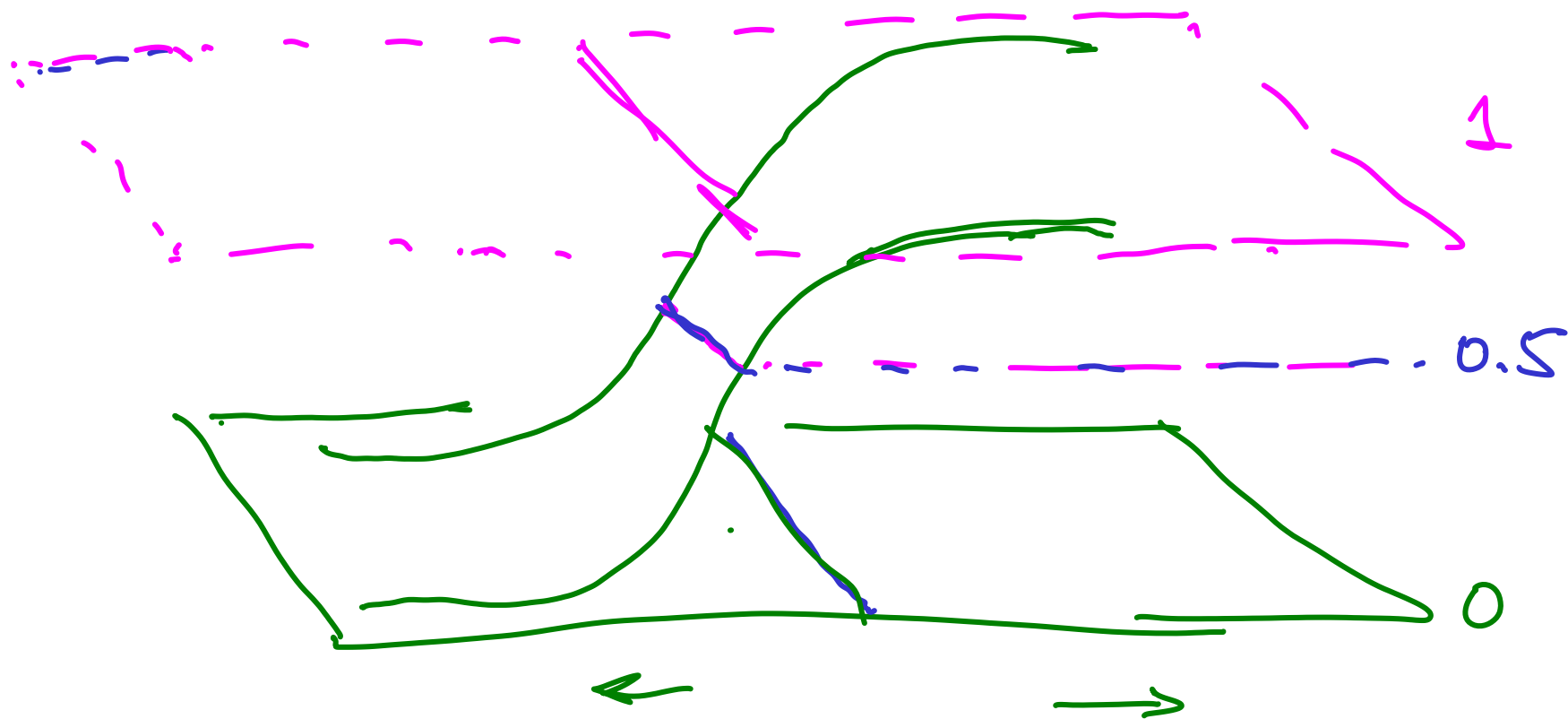
$$w^T x = w_0 + w_1 x_1 + w_2 x_2$$

$$\begin{aligned} w^T x &\rightarrow -\infty \\ \Theta(w^T x) &\rightarrow 0 \end{aligned}$$

$$h(x) < 0.5$$

$$0 \leq \Theta(w^T x) \leq 1$$

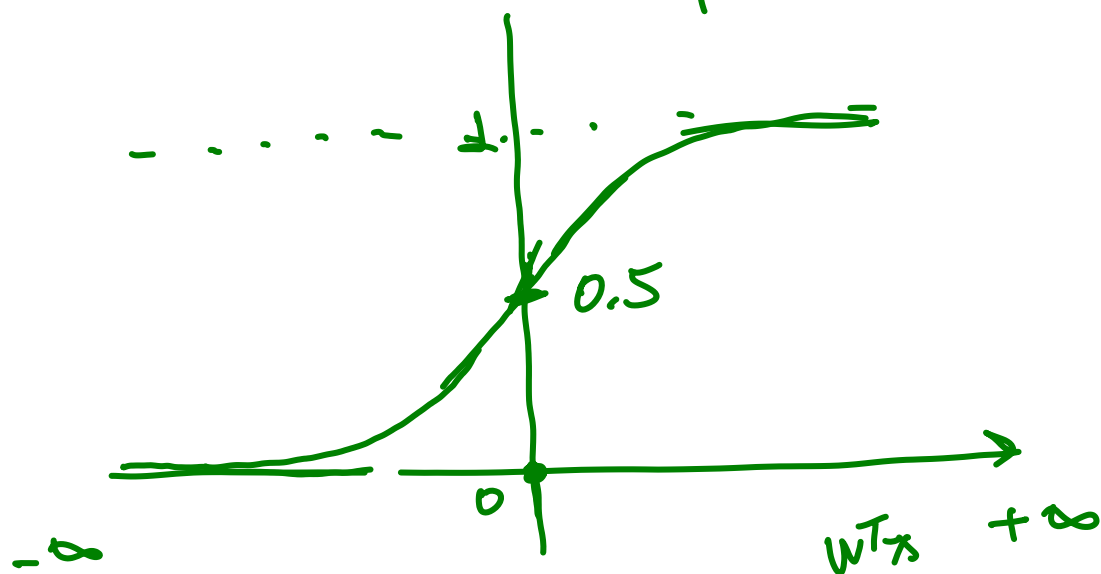




negativos $\rightarrow y=0$

positivos $\rightarrow y=1$

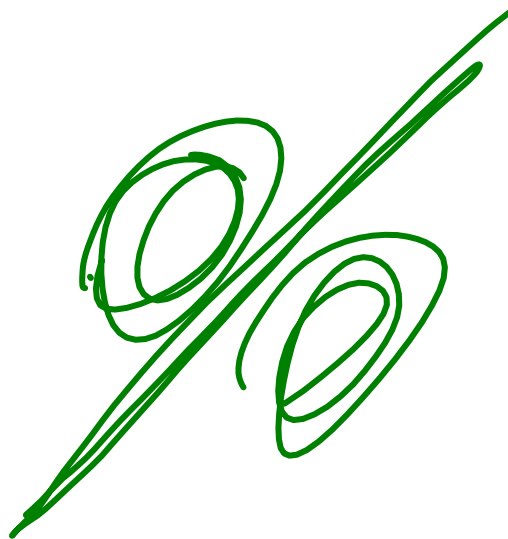
$$\theta(z) = \frac{1}{1 + e^{-z}} = \underline{\underline{0.5}} \quad 0 \leq \theta(z) \leq 1.$$



$$\hat{y} = \begin{cases} 1, & \text{if } \underline{\theta(w^T x)} \geq 0.5 \\ 0, & \text{c.c} \end{cases}$$

$$\downarrow$$

$$\underline{\underline{\theta(w^T x)}}$$



$$\underline{h(x)} = \theta(\underset{\uparrow}{w^T x}) \simeq P(y=+1/x)$$