



Pen and Paper 4

② We need a dataset, where for one point

$$\arg \min_{\substack{X_i \in D \\ X_i \neq X}} \|X - X_i\|_2^2 \neq \arg \min_{\substack{X_i \in D \\ X_i \neq X}} k(X, X_i)$$

$$D \subseteq [0, 1]^2$$

$$\|X - X_i\|_2^2 = (x_1 - y_1)^2 + (x_2 - y_2)^2$$

$$k(X, Y) = \min(x_1, y_1) + \min(x_2, y_2)$$

$$X = (1, 0)$$

$$\|X - X_1\|_2^2 = 1$$

$$X_1 = (0, 0)$$

$$\|X - X_2\|_2^2 = \frac{1}{2}$$

$\Rightarrow X_2$ is the n.n.

$$X_2 = \left(\frac{1}{2}, \frac{1}{2}\right)$$

$$k(X, X_1) = \min(1, 0) + \min(0, 0) = 0$$

$$k(X, X_2) = \min\left(1, \frac{1}{2}\right) + \min\left(0, \frac{1}{2}\right) = \frac{1}{2}$$

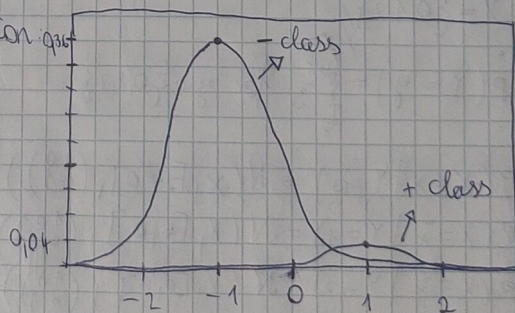
$\Rightarrow X_1$ is the n.n.

③ a) data distribution

$$P(N(\mu, \sigma^2) = \mu) \approx 0.4$$

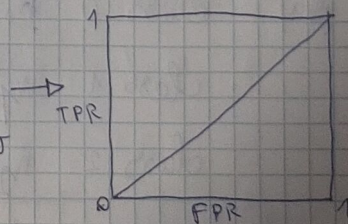
$$0.4 \cdot 0.9 = 0.36$$

$$0.4 \cdot 0.1 = 0.04$$



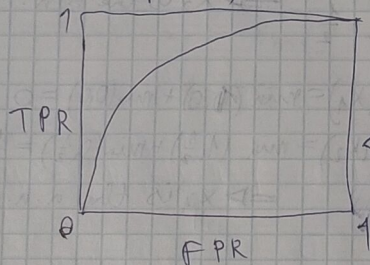
$$TPR = \frac{TP}{TP + FN} = \frac{0.1 \cdot 0.9}{0.1 \cdot 0.9 + 0.1(1-0.9)} = 0.9$$

$$FPR = \frac{FP}{FP + TN} = \frac{0.9 \cdot 0.1}{0.9 \cdot 0.1 + 0.9(1-0.9)} = 0.9$$



$$\begin{aligned}
 b) \text{ TPR} &= \frac{TP}{TP+FN} = \frac{P(\text{pred}=+ \text{ and } \text{real}=+)}{P(\text{real}=+)} = \frac{P(Y=+)}{P(Y=+)} \\
 &= \frac{P(Y=+) \cdot P(\hat{Y}=+ | Y=+)}{P(Y=+)} = P(X > \tau | Y=+) = 1 - P(X < \tau | Y=+) \\
 &= 1 - \Phi(\tau - 1)
 \end{aligned}$$

$$\begin{aligned}
 \text{FPR} &= \frac{FP}{FP+TN} = \frac{P(Y=-) \cdot P(\hat{Y}=+ | Y=-)}{P(Y=-)} = P(\hat{Y}=+ | Y=-) \\
 &= P(X < \tau | Y=-) = 1 - P(X > \tau | Y=-) = 1 - \Phi(\tau + 1)
 \end{aligned}$$



$$\Phi(\tau + 1) > \Phi(\tau - 1)$$

$$\leftarrow 1 - \Phi(\tau + 1) < 1 - \Phi(\tau - 1)$$

$$④ \quad Y = \{0, 1, 2\}$$

$$P(Y=y) = \frac{1}{3} \text{ for } y \in \{0, 1, 2\}$$

$$P(X=x | Y=0) = \mathcal{N}(x, [-3, 0], \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix})$$

$$P(X=x | Y=1) = \mathcal{N}(x, [0, 0], \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix})$$

$$P(X=x | Y=2) = \mathcal{N}(x, [0, 3], \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix})$$

$$a) \text{ class 0: } w_0 = \left[-\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right], b_0 = 0 \Rightarrow y = -x$$

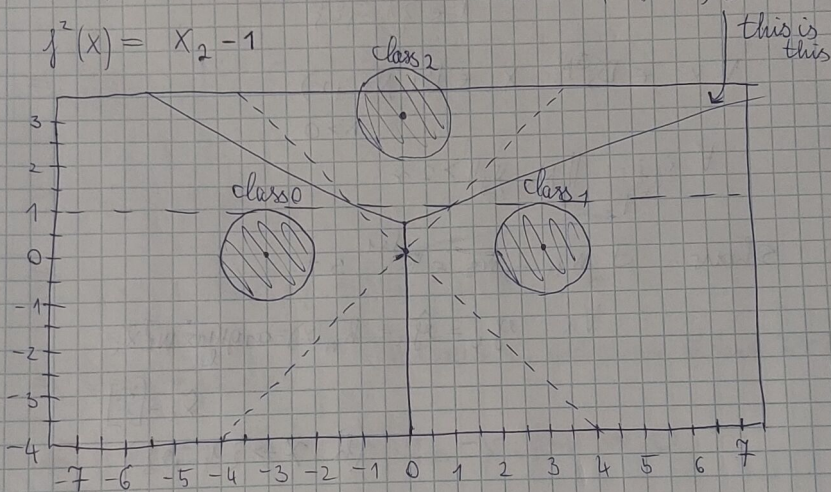
$$\text{class 1: } w_1 = \left[\frac{1}{\sqrt{2}}, -\frac{1}{\sqrt{2}}\right], b_1 = 0 \Rightarrow y = x$$

$$\text{class 2: } w_2 = [0, 1], b_2 = -1 \Rightarrow y = 1$$

$$f^0(x) = -\frac{1}{\sqrt{2}}x_1 - \frac{1}{\sqrt{2}}x_2 \Rightarrow x_2 - 1 = \frac{1}{\sqrt{2}}x_1 - \frac{1}{\sqrt{2}}x_2$$

$$f^1(x) = \frac{1}{\sqrt{2}}x_1 - \frac{1}{\sqrt{2}}x_2 \quad \sqrt{2}x_2 - \sqrt{2} = x_1 - x_2$$

$$f^2(x) = x_2 - 1 \quad 0 = x_1 - (1 + \sqrt{2})x_2 + \sqrt{2}$$



decision boundaries: ---

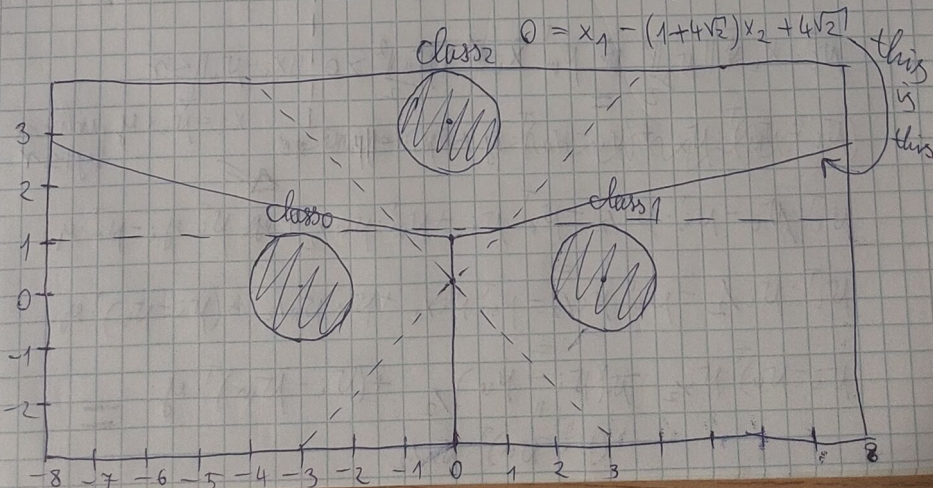
one-vs-all classifiers: ———

b) almost everything is the same, just we need:

$$4x_2 - 4 = \frac{1}{\sqrt{2}}x_1 - \frac{1}{\sqrt{2}}x_2$$

$$4\sqrt{2}x_2 - 4\sqrt{2} = x_1 - x_2$$

$$\text{class2} \quad 0 = x_1 - (1 + 4\sqrt{2})x_2 + 4\sqrt{2}$$



$$(1) D = \{(x_i, y_i)\}_{i=0}^n \quad x_i \in \mathbb{R}^d, y_i \in \{1, \dots, \ell\}$$

ℓ centroids: $\mu_\ell \in \mathbb{R}^d$
 $\ell \in \{1, \dots, \ell\}$

$$\forall x_i \in D_x^{\exists \mu_\ell} \quad x_i \in B(\mu_\ell, r)$$

$$r > 0$$

$$\forall i, j: \quad \mu_i \neq \mu_j \implies \|\mu_i - \mu_j\| > 2r$$

show: $\exists w_1, \dots, w_\ell \in \mathbb{R}^{d+1}$ s.t.

$$\forall i: \quad y_i = \hat{y}_i = h_w(x) = \arg \max_\ell w_\ell^T \hat{x}_i$$

$$\hat{x}_i = \begin{bmatrix} x_i \\ 1 \end{bmatrix}$$

$$(h: \mathcal{X} \rightarrow \{1, \dots, \ell\})$$

$$w_i = [\mu_i, -\|\mu_i\|^2/2] \text{ will work:}$$

We need that if $y_i = \ell$, then

$$w_\ell^T \hat{x}_i > w_m^T \hat{x}_i \quad \forall m = 1, \dots, \ell, m \neq \ell$$

$$\mu_\ell^T x_i - \frac{\|\mu_\ell\|^2}{2} > \mu_m^T x_i - \frac{\|\mu_m\|^2}{2} \quad \text{if } \|x_i - \mu_\ell\| \leq r$$

$$(\mu_\ell - \mu_m)^T x_i > \frac{\|\mu_\ell\|^2 - \|\mu_m\|^2}{2} \quad \text{where } x_i = \mu_\ell + y, \text{ where } \|y\| \leq r$$

$$\mu_\ell^T \mu_\ell - \mu_\ell^T \mu_m - \mu_m^T \mu_\ell + \mu_m^T \mu_m + (\mu_\ell - \mu_m)^T y$$

$$\mu_\ell^T \mu_\ell - \mu_m^T \mu_\ell - \mu_m^T \mu_\ell + \mu_m^T \mu_m + (\mu_\ell - \mu_m)^T y$$

$$(\mu_\ell - \mu_m)^T \mu_\ell - (\mu_\ell - \mu_m)^T \mu_m + (\mu_\ell - \mu_m)^T y = (\mu_\ell - \mu_m)^T \mu_\ell/2$$

$$= (\mu_i - \mu_m)^T \cdot (\mu_i - \mu_m) / 2 + (\mu_i - \mu_m)^T \cdot y_i =$$

$$= \|\mu_i - \mu_m\|_2^2 / 2 + (\mu_i - \mu_m)^T \cdot y_i =$$

$$= \|\mu_i - \mu_m\|_2^2 / 2 - (\mu_m - \mu_i)^T \cdot y_i \geq$$

$$\geq \|\mu_i - \mu_m\|_2^2 / 2 - \|\mu_m - \mu_i\| \cdot \|y_i\| \geq$$

$$\geq \|\mu_i - \mu_m\|_2^2 / 2 - \|\mu_m - \mu_i\| \cdot r \rightarrow$$

~~$\|\mu_i - \mu_m\|_2^2 / 2 - \|\mu_m - \mu_i\| \cdot r$~~ because μ_i and μ_m are at least $2r$ apart

$$\geq \|\mu_i - \mu_m\| \cdot r - \|\mu_m - \mu_i\| \cdot r = 0 \quad \square$$