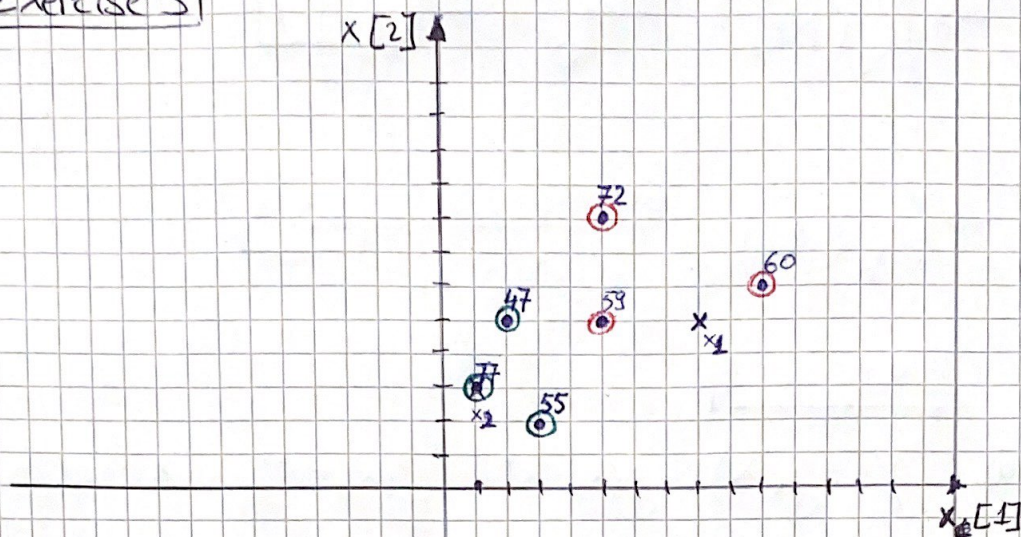


### Exercise 3



- Case 1:  $k=1$ . Let  $f_{k_1}$  be the KNN regressor with  $k=1$ .

→ The first point of the training set coincides with the test point  $x_2$ . Thus, since  $k=1$ , there's no need to look further.  $f_{k_1}(x_2) = 77$

→ By looking at the plot by naked eye, the fourth and sixth points of the training set seem nearest to  $x_1$ .  $\|x_1 - \begin{pmatrix} 5 \\ 5 \end{pmatrix}\|$  compared to  $\|x_1 - \begin{pmatrix} 10 \\ 6 \end{pmatrix}\|$ .

$$\sqrt{(8-5)^2 + (5-5)^2} = 3 \text{ compared to } \sqrt{(8-10)^2 + (5-6)^2}.$$

Clearly,  $\sqrt{5} < 3$ .  $\begin{pmatrix} 10 \\ 6 \end{pmatrix}$  is the nearest point to  $x_1$ . Hence,  $f_{k_1}(x_1) = 60$ .

- Case 2:  $k=3$ . Let  $f_{k_2}$  be the KNN regressor with  $k=3$ .

The three nearest points to  $x_1$  have been marked in red. The three nearest points to  $x_2$  have been marked in green.



$$\rightarrow f_{k_2}(x_2) = \frac{77+47+55}{3} = \boxed{59.7}$$

$$\rightarrow f_{k_2}(x_2) = \frac{60+72+59}{3} = \boxed{63.7}$$

### Exercise 1

a)  $Y := g(X)$   $g(x) = x^4$   $f(x) = x^3$   $X \sim \mathcal{U}[-1, 1]$

$$EPE(f) = \mathbb{E}[L_2(Y, f(X))]$$

$$EPE(f) = \iint_{R_x R_y} (y - f(x))^2 p(x, y) dx dy =$$

$$\iint_{R_x R_y} \underbrace{(y - f(x))^2}_{g(y)} p_X(x) \underbrace{\delta(y - x^4)}_{y_0} dx dy =$$

STEP EXPLANATION

$$\int_{R_y} A \cdot B dy = \int_{R_y} g(y) \delta(y - y_0) dy = g(y_0)$$

\* Please note that

$$p_X(t) = \frac{1}{2} \quad \forall t \in [-1, 1] \text{ and}$$

$p_X(t) = 0$  otherwise because  $x \in \mathcal{U}[-1, 1]$

$$\int_{R_x} (x^4 - x^3)^2 p_X(x) dx =$$

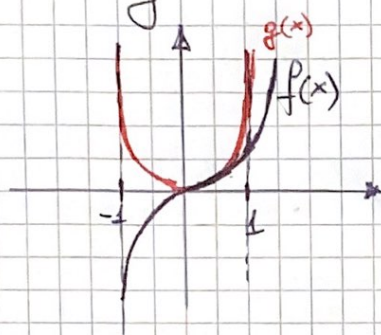
$$\frac{1}{2} \int_{R_x} (x^4 - x^3)^2 dx = \frac{1}{2} \int_{-1}^1 (x^4 - x^3)^2 dx = \frac{1}{2} \int_{-1}^1 x^8 - 2x^7 + x^6 dx =$$



$$\frac{1}{2} \left[ \frac{x^9}{9} \right]_{-1}^1 - 2 \left[ \frac{x^8}{8} \right]_{-1}^1 + \left[ \frac{x^7}{7} \right]_{-1}^1 = \frac{1}{2} \left[ \frac{2}{9} + \frac{2}{7} \right] = \frac{1}{9} + \frac{1}{7}$$

$$EPE(f) = \boxed{0.254}$$

The error is notably higher compared to the lecture's example because  $\forall x \in [-1, 0]$   $f(x)$  diverges to the third quadrant of the Cartesian plane while  $g(x)$  is decreasing monotonously in the second quadrant.



b) The regressor for  $Y = x^2$

$$\mathbb{E}[Y|X=x] = \int_{R_Y} y p(y|x) dy = \int_{R_Y} y \frac{p(y, x)}{p_X(x)} dy$$

From the definition of  $p(y, x)$  using the Dirac function, we have:

$$\mathbb{E}[Y|X=x] = \int_{R_Y} y \cdot \frac{\cancel{p_X(x)} \delta(y - x^2)}{\cancel{p_X(x)}} dy$$

$$\Rightarrow \mathbb{E}[Y|X=x] = \int_{R_Y} y \underbrace{\delta(y - x^2)}_{\substack{\text{in } \\ y_0}} dy = \boxed{x^2}$$

From Knowledge 3.1 of the Lecture notes