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Machine Learning

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Rappel Calcul matriciel

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Produit Matriciel

$$\begin{array}{ccc} A & \cdot & B = C \\ \downarrow & & \downarrow \\ d1 \times d2 & & d2 \times d3 \end{array} \quad \downarrow \\ d1 \times d3$$

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}_{2 \times 2}$$

$$B = \begin{bmatrix} e & f & g \\ h & i & j \end{bmatrix}_{2 \times 3}$$

$$A \cdot B = \begin{array}{c} ae + bh \\ \begin{bmatrix} - & - & - \\ - & - & - \end{bmatrix} \end{array}_{2 \times 3}$$

Transposé

$$A = \begin{bmatrix} a & b \\ c & d \\ e & f \end{bmatrix}_{3 \times 2}$$

$$A^T = \begin{bmatrix} a & c & e \\ b & d & f \end{bmatrix}_{2 \times 3}$$

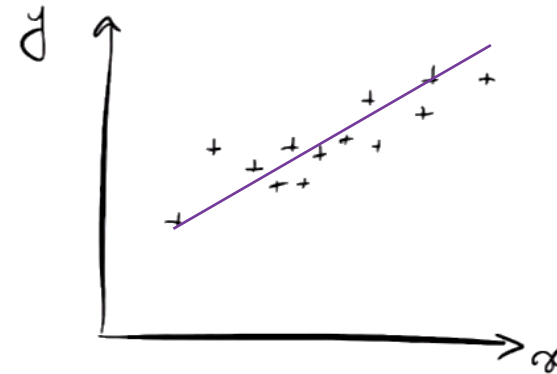
Transformation matricielle

Modèle (Régression linéaire simple)

$$f(x) = a x + b$$
$$F = X \cdot \Theta$$

Diagram illustrating the matrix representation of the linear regression model:

- F (m × 1): Vector of predicted values $f(x^{(1)}), f(x^{(2)}), \dots, f(x^{(m)})$.
- X (m × 2): Design matrix with features $x^{(1)}, \dots, x^{(m)}$ and a bias term of 1 for each row.
- Θ (2 × 1): Parameter vector $\begin{bmatrix} a \\ b \end{bmatrix}$.



x	y
$x^{(1)}$	$y^{(1)}$
\vdots	\vdots
$x^{(m)}$	$y^{(m)}$

m

n = 1

Fonction de coût

$$J(a, b) = \frac{1}{2m} \sum_{i=1}^m \left(\underbrace{a}_{1 \times 1} \underbrace{x^{(i)}}_{m \times 1} + \underbrace{b}_{m \times 1} - \underbrace{y^{(i)}}_{m \times 1} \right)^2$$

$$\frac{1}{2m} \times \begin{bmatrix} - \\ - \\ - \\ - \\ - \end{bmatrix}^2$$

$$J(\theta) = \frac{1}{2m} \sum (X\theta - Y)^2$$

Le Gradient

$$\left[\begin{array}{l} \frac{\partial J(a, b)}{\partial a} \\ \frac{\partial J(a, b)}{\partial b} \end{array} \right] = \frac{1}{m} \sum x(a x + b - y)$$

$$\frac{\partial J(\theta)}{\partial \theta} = \frac{1}{m} X^T (X\theta - Y)$$

$$X = \begin{bmatrix} x^{(1)} & 1 \\ \vdots & \vdots \\ x^{(m)} & 1 \end{bmatrix}$$

$$X^T = \begin{bmatrix} x^{(1)} & x^{(2)} & \dots & x^{(m)} \\ 1 & 1 & \dots & 1 \end{bmatrix}$$

Descente de gradient

$$\left\{ \begin{array}{l} a = a - \alpha \frac{\partial J}{\partial a} \\ b = b - \alpha \frac{\partial J}{\partial b} \end{array} \right.$$



$$\left\{ \begin{array}{l} \theta = \theta - \alpha \frac{\partial J}{\partial \theta} \end{array} \right.$$

$$X = \begin{bmatrix} x^{(1)} & 1 \\ \vdots & \vdots \\ x^{(m)} & 1 \end{bmatrix} \quad \theta = \begin{bmatrix} a \\ b \end{bmatrix} \quad Y = \begin{bmatrix} y^{(1)} \\ \vdots \\ y^{(m)} \end{bmatrix}$$

$m \times (n+1) \quad (n+1) \times 1 \quad m \times 1$

$$F = X \cdot \theta \quad J(\theta) = \frac{1}{2m} \sum (X \cdot \theta - Y)^2$$

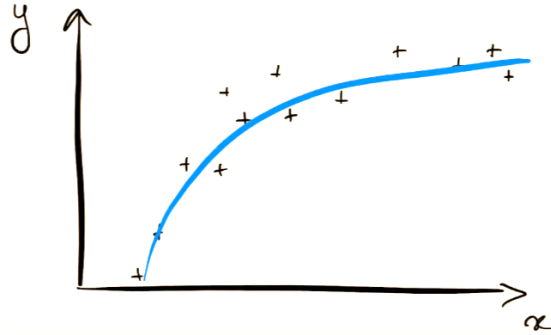
$m \times 1 \quad 1 \times 1$

$$\frac{\partial J(\theta)}{\partial \theta} = \frac{1}{m} X^T \cdot (X \cdot \theta - Y)$$

$(n+1) \times 1$

Transformation matricielle

Modèle (Régression polynomiale) Modèle (Régression multiple)



$$f(x) = ax^2 + bx + c$$

$$F = X \cdot \Theta$$

$$\begin{bmatrix} x^{(1)} & x^{(1)} & 1 \\ \vdots & \vdots & \vdots \\ x^{(m)} & x^{(m)} & 1 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$m \times 3$ 3×1

$$f(x) = ax_1 + bx_2 + c$$

$$F = X \cdot \Theta$$

$$\begin{bmatrix} x_1^{(1)} & x_2^{(1)} & 1 \\ \vdots & \vdots & \vdots \\ x_1^{(m)} & x_2^{(m)} & 1 \end{bmatrix} \cdot \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$m \times 3$ 3×1

Coefficient de détermination

$$R^2 = 1 - \frac{\sum (y - f(x))^2}{\sum (y - \bar{y})^2}$$

- Permet d'évaluer les performances d'un modèle de ML
- Plus il est proche de 1 plus le modèle est performant