

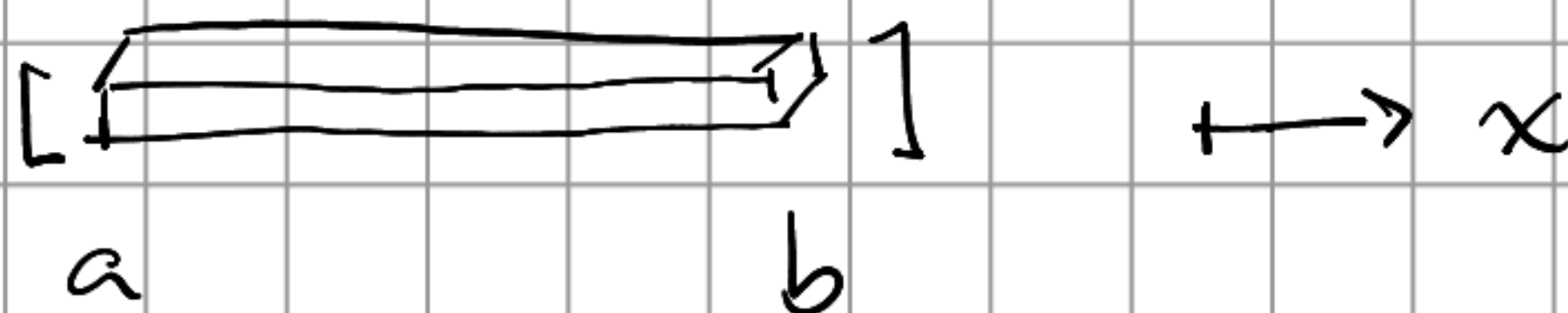
1) Métodos numérico para EDO con valores  
en la frontera

Ecuación de Calor:  $y' = f(t, y)$   
 ← difusión ← reacción ← Fuente.

$$\underbrace{-\frac{d}{dx} \left[ k(x) \frac{du}{dx} \right]}_{\text{difusión}} + \underbrace{b(x)u}_{\text{reacción}} = \underbrace{f(x)}_{\text{Fuente.}} \quad x \in (a, b)$$

$u(a) = \alpha$   
 $u(b) = \beta$

$u$ : es la temperatura en el medio  $^{\circ}\text{C}, ^{\circ}\text{F}, \text{K}$



$k(x)$ : conductividad térmica

$$\underline{\underline{k(x) > 0}}, \quad b(x) \geq 0.$$

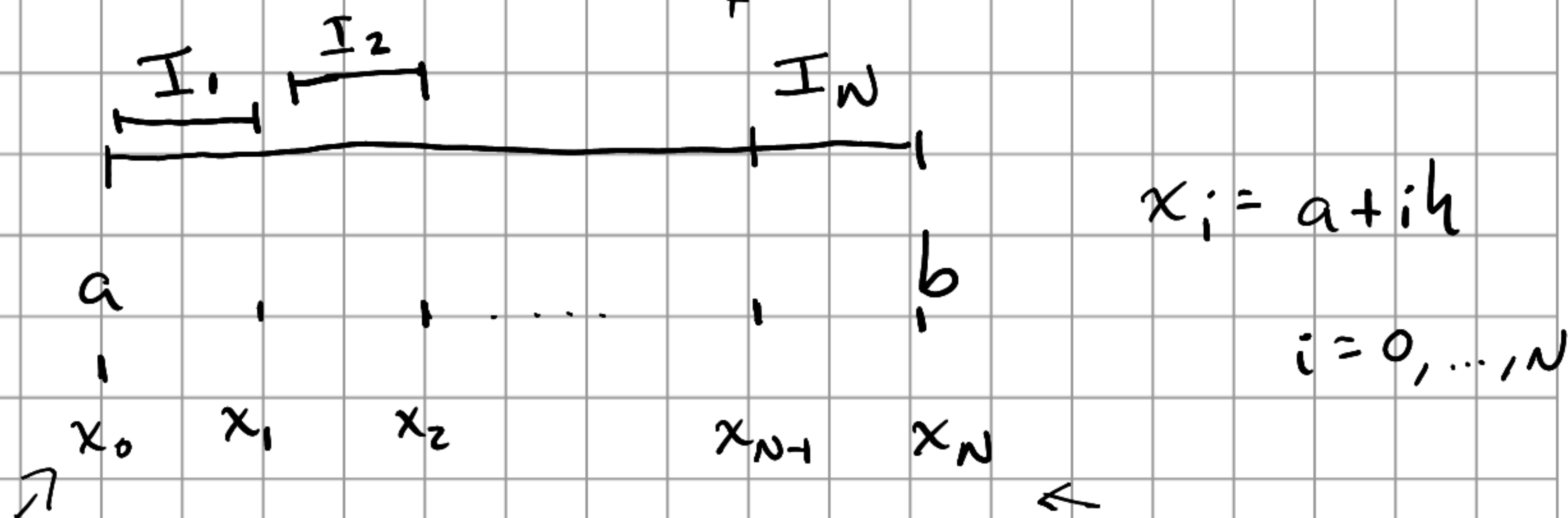
## 2) método de diferencias finitas

Suponemos que  $b(x) \equiv 0$ ,  $k(x) = k > 0$ .

Ecuación de calor: 
$$\left\{ \begin{array}{l} -k \frac{d^2 u}{dx^2} = f(x) \quad x \in (a, b) \\ u(a) = \alpha \\ u(b) = \beta \end{array} \right.$$

(\*)  $k \frac{d^2 u}{dx^2} = -f(x)$

Creemos un malla por  $[a, b]$ . Sea  $N$  el # de sub-intervalos (uniformes) que divide a  $[a, b]$



$h = |I_i| \quad i = 1, \dots, N$ , Observar que tenemos  $N+1$  puntos en la malla.

Como  $x_0 = a \Rightarrow u(x_0) = \alpha$  y similarmente  
 $u(x_N) = \beta$ .

Observar que para  $x_i \in (a, b)$

$$\underline{(*)} \quad \underline{k u''(x_i) = -f(x_i)} \quad (\text{Dif. finitas})$$

3) Usando el teorema de Taylor con residuo para  $u(x_{i+1})$  y  $u(x_{i-1})$ , respectivamente, alrededor del punto  $u(x_i)$  obtenemos:

$$u(x_{i+1}) = u(x_i + h) = u(x_i) + hu'(x_i) + \frac{h^2}{2} u''(x_i) + \frac{h^3}{6} u'''(x_i) + ch^4$$

$$u(x_{i-1}) = u(x_i - h) = u(x_i) - hu'(x_i) + \frac{h^2}{2} u''(x_i) - \frac{h^3}{6} u'''(x_i) + ch^4$$

$$\Rightarrow u(x_{i+1}) + u(x_{i-1}) = 2u(x_i) + \underline{h^2 u''(x_i)} + ch^4$$

$$u''(x_i) = \frac{u(x_{i+1}) - 2u(x_i) + u(x_{i-1}))}{h^2} + ch^2 \swarrow$$

Formula de diferencia central de orden 2 para  $\underline{u''(x_i)}$ .

$$\Rightarrow \underline{(*)} \quad \underline{\frac{u(x_{i+1}) - 2u(x_i) + u(x_{i-1}))}{h^2} = -\frac{1}{k} f(x_i) + ch^2} \quad \checkmark$$

$$\hat{u}_i \sim u(x_i) \quad \leftarrow$$

$$\Rightarrow \left\{ \begin{array}{l} \hat{u}_{i+1} - 2\hat{u}_i + \hat{u}_{i-1} = -\frac{h^2}{k} f(x_i) \end{array} \right. \quad i=1, \dots, N-1$$

$$\hat{u}_0 = \alpha (= u(x_0) = u(a))$$

$$\hat{u}_N = \beta (= u(x_N) = u(b))$$

$$\begin{array}{l} \uparrow \\ x_i \in (a, b) \\ \hat{u}_1, \dots, \hat{u}_{N-1} \end{array}$$

Este es un sistema lineal

$$A \underline{u} = f$$

$$i=1$$

$$\hat{u}_2 - 2\hat{u}_1 + \hat{u}_0 = -\frac{h^2}{k} f(x_1)$$

$$\hat{u}_2 - 2\hat{u}_1 = -\frac{h^2}{k} f(x_1) - \alpha$$

$$i=N-1$$

$$\hat{u}_N - 2\hat{u}_{N-1} + \hat{u}_{N-2} = -\frac{h^2}{k} f(\hat{u}_{N-1})$$

$$-2\hat{u}_{N-1} + \hat{u}_{N-2} = -\frac{h^2}{k} f(\hat{u}_{N-1}) - \beta$$

$$i=2, \dots, N-2$$

$$\hat{u}_{i+1} - 2\hat{u}_i + \hat{u}_{i-1} = -\frac{h^2}{k} f(x_i)$$

↑

4) En forme matriciel

$$\hat{\underline{u}} = [\hat{u}_1, \hat{u}_2, \dots, \hat{u}_{N-1}]^t \in \mathbb{R}^{N-1}$$

↘

↙

$$\begin{bmatrix} -2 & 1 & & \\ 1 & -2 & 1 & \\ & 1 & -2 & 1 \\ & & \ddots & \ddots \\ & & & 1 & -2 & 1 \\ & & & & 1 & -2 \end{bmatrix} \begin{bmatrix} \hat{u}_1 \\ \hat{u}_2 \\ \hat{u}_3 \\ \vdots \\ \hat{u}_{N-2} \\ \hat{u}_{N-1} \end{bmatrix} = \begin{bmatrix} -\frac{h^2}{k} f(x_1) - \alpha \\ -\frac{h^2}{k} f(x_2) \\ \vdots \\ -\frac{h^2}{k} f(x_{N-2}) \\ -\frac{h^2}{k} f(x_{N-1}) - \beta \end{bmatrix}$$

$\tilde{\underline{A}}$

$$\tilde{\underline{A}} \in \mathbb{R}^{N-1} \times \mathbb{R}^{N-1}$$

$$\tilde{\underline{f}} \in \mathbb{R}^{N-1}$$

$$\hat{\underline{u}} = \tilde{\underline{A}}^{-1} \cdot \tilde{\underline{f}}$$