

We have taken

$$u_t = (k u_x)_x + \psi(x) \Rightarrow \frac{\partial u}{\partial t} = \frac{\partial}{\partial x} \left(k \frac{\partial u}{\partial x} \right) + \psi(x)$$

\Downarrow

$$u''(x) = f(x) \quad \frac{d^2 u}{dx^2} = f(x)$$

\Downarrow

$$\frac{u(x_j+h) - 2u(x_j) + u(x_j-h))}{h^2} \approx f(x_j) \quad j=1, 2, \dots, m, \quad h=1/(m+1)$$

\Downarrow

$$\frac{u_{j+1} - 2u_j + u_{j-1}}{h^2} = F_j \quad F_j = f(x_j)$$

\Downarrow

$$\begin{cases} A U = F \\ u_0 = \alpha \\ u_{m+1} = \beta \end{cases}$$

$$A = \begin{bmatrix} -1 & 1 & 0 & \dots & 0 \\ 1 & -2 & 1 & \dots & 0 \\ 0 & 1 & -2 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & -1 \end{bmatrix}, \quad U = \begin{bmatrix} u_1 \\ \vdots \\ u_m \end{bmatrix}$$

$$F = \begin{bmatrix} f_1 - \alpha/h^2 \\ \vdots \\ f_m - \beta/h^2 \end{bmatrix}$$

\Downarrow

$$U = "A^{-1}" F$$

We need methods - efficient and effective

Thomas Algorithm...

→ discuss Wikipedia page on Assignment #2.

Another method: Jacobi Iteration:

Suppose we have a system of the form

$$Ax = b, \quad A \in \mathbb{R}^{m \times m}, \quad x \in \mathbb{R}^m, \quad b \in \mathbb{R}^m$$

Now we can rewrite

$$A = (L + D + U)$$

$$L = \begin{bmatrix} 0 & & & 0 \\ x & & & \\ & \ddots & & \\ x & & x & 0 \end{bmatrix} \quad D = \begin{bmatrix} x & & 0 \\ & \ddots & \\ 0 & & x \end{bmatrix} \quad U = \begin{bmatrix} 0 & x & x & x \\ & \ddots & & \\ & & \ddots & x \\ 0 & & & 0 \end{bmatrix}$$

$$\Rightarrow Ax = (L + D + U)x$$

$$= (L + U)x + Dx = b$$

$$\hookrightarrow Dx = b - (L + U)x$$

$$\hookrightarrow x = D^{-1}(b - (L + U)x)$$

Suggests an iteration.

$$x^{(k+1)} = D^{-1}(b - (L + U)x^{(k)})$$

So, we need to start with some $x^{(0)}$ as a guess and compute $x^{(1)}$ and on...

So,

$$\begin{aligned}x^{(k+1)} &= D^{-1}(b - \underbrace{(L+D+U)}_{(A-D)}x^{(k)}) \\&= D^{-1}(b - (A-D)x^{(k)}) \\&= D^{-1}(b - Ax^{(k)} + Dx^{(k)}) \\&= D^{-1}(b - Ax^{(k)}) + x^{(k)} \\&= x^{(k)} + D^{-1}r^{(k)} \\&\quad \text{[residual vector.]}\end{aligned}$$

Def. Suppose we have a linear system of equation with

$$Ax = b$$

Then for any vector $x \in \mathbb{R}^n$, the residual vector r is defined by

$$r = b - Ax$$

This gives a way to test a vector to see if x satisfies the linear system

$$r=0 \Rightarrow b - Ax = 0 \Rightarrow Ax = b.$$

So, we have two linear system solvers.

1. Thomas Alg.
2. Jacobi Iteration

We already have seen how to make the Thomas algorithm more efficient. ④

- 1 store as vectors
- 2 use smart ways to eliminate

The same is true for the Jacobi iteration. We just need to be careful with matrix null.

Code for matrix-vector product.

$$A \cdot x \rightarrow y$$

```
for i in range(m):  
    y[i] = 0.0  
    for j in range(m):  
        y[i] = y[i] + A[i][j] * x[j]  
    end  
end  
y[i] = sum.
```

← sum = 0.0

← y[i] = sum.

Version to avoid access issue for arrays