

# Elliptic Equation



Typical equation: Poisson Equation

$$\Delta\phi = \rho$$

$$\frac{d^2\phi}{dx^2} = \rho \quad \frac{\partial^2\phi}{\partial x^2} + \frac{\partial^2\phi}{\partial y^2} = \rho \quad \frac{\partial^2\phi}{\partial x^2} + \frac{\partial^2\phi}{\partial y^2} + \frac{\partial^2\phi}{\partial z^2} = \rho$$

Very important equation in science and engineering

- Maxwell's Equation for Scalar Potential
- Incompressible Navier-Stokes Equation
- Steady State of Diffusion (Thermal Conduction) Equation

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# Poisson Equation



1-dimensional case:

$$\frac{\phi_{j+1} - 2\phi_j + \phi_{j-1}}{\Delta x^2} = \rho_j$$

2-dimensional case:

$$\frac{\phi_{i+1,j} - 2\phi_{i,j} + \phi_{i-1,j}}{\Delta x^2} + \frac{\phi_{i,j+1} - 2\phi_{i,j} + \phi_{i,j-1}}{\Delta y^2} = \rho_{i,j}$$

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# Poisson Equation



N x N linear coupling equations

$$\begin{aligned} \phi_1 &= 0 \\ \phi_1 - 2\phi_2 + \phi_3 &= \rho_2 \Delta x^2 \\ \phi_2 - 2\phi_3 + \phi_4 &= \rho_3 \Delta x^2 \\ \phi_3 - 2\phi_4 + \phi_5 &= \rho_4 \Delta x^2 \\ &\dots \\ \phi_{j-1} - 2\phi_j + \phi_{j+1} &= \rho_j \Delta x^2 \\ &\dots \\ \phi_{N-2} - 2\phi_{N-1} + \phi_N &= \rho_{N-1} \Delta x^2 \\ \phi_N &= 0 \end{aligned}$$

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# Poisson Equation



$$\begin{pmatrix} 1 & & & & \\ 1 & -2 & 1 & & \\ & \cdot & \cdot & \cdot & \\ & & 1 & -2 & 1 \\ & & & \cdot & \cdot & \cdot \\ & & & & 1 & -2 & 1 \\ & & & & & 1 \end{pmatrix} \begin{pmatrix} \phi_1 \\ \phi_2 \\ \vdots \\ \phi_j \\ \vdots \\ \phi_{N-1} \\ \phi_N \end{pmatrix} = \Delta x^2 \begin{pmatrix} \rho_1 \\ \rho_2 \\ \vdots \\ \rho_j \\ \vdots \\ \rho_{N-1} \\ \rho_N \end{pmatrix}$$

Gauss Elimination Method is not available, because the matrix size is too large.

Sparse Matrix:

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# 3-D Poisson Equation



$$\begin{pmatrix} \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \\ \text{---} \end{pmatrix} \begin{pmatrix} \phi_{1,1,1} \\ \phi_{2,1,1} \\ \phi_{i,j,k} \\ \phi_{N-1,N,N} \\ \phi_{N,N,N} \end{pmatrix} = \Delta x^2 \begin{pmatrix} \rho_{1,1,1} \\ \rho_{2,1,1} \\ \rho_{i,j,k} \\ \rho_{N-1,N,N} \\ \rho_{N,N,N} \end{pmatrix}$$

For example,  $N_x \times N_y \times N_z = 100 \times 100 \times 100$   
 Double precision 8 byte  $\times (1000000)^2 = 8 \times 10^{12} = 8 \text{ TB}$   
 c.f. TSUBAME : 20TB

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# Matrix Solver



## Relaxation Method:

- Point Jacobi
- Gauss-Seidel
- SOR
- ICCG (Incomplete Conjugate Gradient)
- ILUCR (Incomplete LU Conjugate Residual)
- BiCGStab (Bi-conjugate Conjugate Gradient Stabilize)

- Advantage : Memory and CPU time
- Disadvantage : No guarantee to be solved limited types of matrix

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# Point Jacobi Method



Expecting iterative convergence:

$$\frac{\phi_{j+1}^n - 2\phi_j^{n+1} + \phi_{j-1}^n}{\Delta x^2} = \rho_j$$

The **n+1**-th value is calculated by the **n**-th value.

$$\phi_j^{n+1} = \frac{1}{2} (\phi_{j+1}^n + \phi_{j-1}^n - \Delta x^2 \rho_j)$$

If  $\phi_j^{n+1} \approx \phi_j^n$ ,  $\phi_j^n$  is the solution of Poisson equation.

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# Point Jacobi Method



Starting from the initial value  $\phi_j^0$ ,

$$\phi_j^1 = \frac{1}{2} (\phi_{j+1}^0 + \phi_{j-1}^0 - \Delta x^2 \rho_j)$$

Iterative calculations give  $\phi_j^1, \phi_j^2, \phi_j^3, \dots, \phi_j^n, \phi_j^{n+1}$

If  $|\phi_j^{n+1} - \phi_j^n| < \varepsilon$  is satisfied, the iteration has been converged.

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# Point Jacobi Method



von Neumann's stability analysis  
for the iterative process:

Assuming the perturbation  $\phi_j^n = \delta \phi^n e^{ik \cdot j \Delta x}$

$$\delta \phi^{n+1} / \delta \phi^n = \frac{1}{2} (e^{ik \Delta x} + e^{-ik \Delta x}) = \cos k \Delta x$$

The iteration process is stable, but slow.  
Actual stability depends on the source term.

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# Point Jacobi Method



Introduction of the relaxation factor :  $\omega$

$$\phi_j^{n+1} = (1 - \omega) \phi_j^n + \omega \frac{1}{2} (\phi_{j+1}^n + \phi_{j-1}^n - \Delta x^2 \rho_j)$$

Stability analysis  $\delta \phi^{n+1} / \delta \phi^n = (1 - \omega) + \omega \cos k \Delta x < 1$   
( $0 \leq \omega \leq 1$ )

Small  $\delta \phi^{n+1} / \delta \phi^n$  means rapid decrease of the error.

$\omega = 1$  is found to be the fastest convergence of Jacobi  
Iteration method.

[Source Code](#)

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# SOR Method



Any faster convergence technique ?

When we calculate  $\phi_j^{n+1}$ , if  $\phi_{j-1}^{n+1}$  has been calculated,  
it is better convergence to use  $\phi_{j-1}^{n+1}$ .

$$\phi_j^{n+1} = \frac{1}{2} (\phi_{j+1}^n + \phi_{j-1}^{n+1} - \Delta x^2 \rho_j)$$

This iteration method is called "Gauss-Seidel" method.

[Source Code](#)

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# SOR Method



Introduction of the relaxation factor :  $\omega$

$$\phi_j^{n+1} = (1 - \omega) \phi_j^n + \omega \frac{1}{2} (\phi_{j+1}^n + \phi_{j-1}^{n+1} - \Delta x^2 \rho_j)$$

The stability analysis shows that the iteration  
process is stable for  $0 < \omega < 2$ .

Acceleration of the iteration :  $1 < \omega < 2$ .

**SOR (Successive Over-relaxation) Method**

The fastest convergence is achieved for  $\omega \approx 1.82$ .

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