

Introduction to Parallel Programming for Multicore/Manycore Clusters

Part A1: FVM Code (PCG)

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Date	ID	Title
Apr-06 (W)	CS-01a	Introduction-a
Apr-13 (W)	CS-01b	Introduction-b (Introduction-a and -b are same)
Apr-20 (W)	CS-02	FVM (1/2)
Apr-27 (W)	CS-03	FVM (2/2)
May-04 (W)	(no class)	(National Holiday)
May-11 (W)	CS-04	OpenMP (1/3), Login to Odyssey
May-18 (W)	CS-05	OpenMP (2/3)
May-25 (W)	CS-06	OpenMP (3/3)
Jun-01 (W)	CS-07	FVM/ICCG(1/2)
Jun-08 (W)	CS-08	FVM/ICCG(2/2)
Jun-15 (W)	CS-09	Reordering (1/3)
Jun-22 (W)	CS-10	Reordering (2/3)
Jun-29 (W)	CS-11	Reordering (3/3)
Jul-06 (W)	CS-12	Parallel Code by OpenMP (1/3)
Jul-13 (W)	CS-13	Parallel Code by OpenMP (2/3)
Jul-20 (W)	CS-14	Parallel Code by OpenMP (3/3). Q/A

Part-A
PCG Solver
Point Jacobi
Preconditioning
(Simple, Easy)

Part-B
ICCG Solver
Incomplete
Cholesky
Preconditioning
(More
Complicated,
Difficult)

Target of Part-A

- Material: PCG solver for sparse matrices derived from FVM applications (Finite Volume Method).
 - Point Jacobi/Diagonal Scaling Preconditioner
- Parallelization on a single node of Wisteria/BDEC-01 (Odyssey) using OpenMP
 - Data Placement
 - Methods for Matrix Storage
- Keywords
 - Finite Volume Method (FVM)
 - Sparse Matrices
 - PCG Method: **Preconditioned Conjugate Gradient**

Files on PC

Files on WEB:

<http://nkl.cc.u-tokyo.ac.jp/files/fvm.tar>

```
>$ cd  
>$ tar xvf fvm.tar  
>$ cd fvm
```

Please confirm that following directories are created:

src-c, src-f, run

Call the fvm directory <\$P-FVM>

PC

Odyssey

Paraview for Visualization

<http://www.paraview.org/>

Free Software

Windows, iOS, Unix/Linux

<http://nkl.cc.u-tokyo.ac.jp/22s/ParaView.pdf>

- Background
 - Finite Volume Method
 - Preconditioned Iterative Solvers
- PCG Solver for Poisson's Equations
 - How to run
 - Data Structure
 - Program
 - Initialization
 - Coefficient Matrices
 - PCG

Target Application

- 3D Poisson Equation/Poisson's Equation

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

- Finite Volume Method (FVM)
 - Arbitrary Shape Meshes, Cell-Centered
 - “Direct” Finite Difference Method
- Boundary Conditions (B.C.) etc.
 - Dirichlet B.C., Volume Flux
- Preconditioned Iterative Solvers (PCG)
 - Conjugate Gradient + Preconditioner

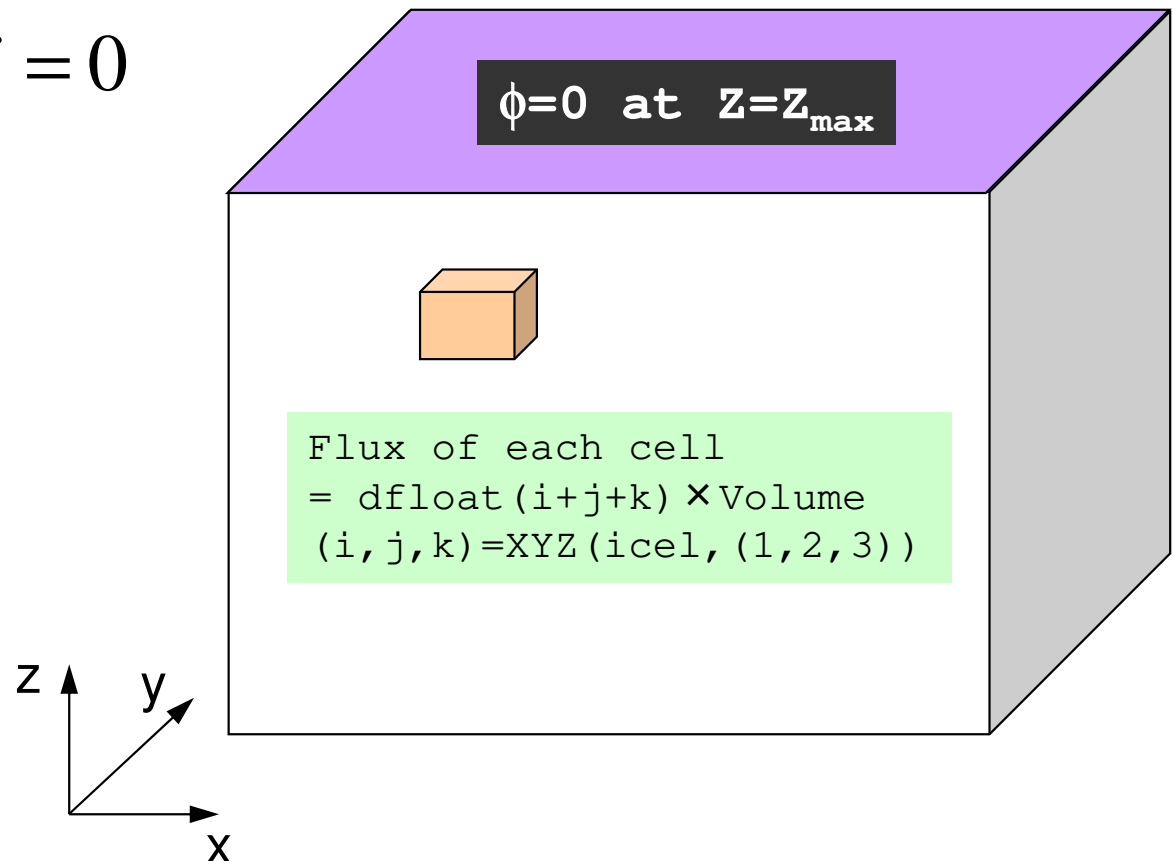
Target Problem: Variables are defined at cell-center's

Poisson Equation/ Poisson's Equation

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

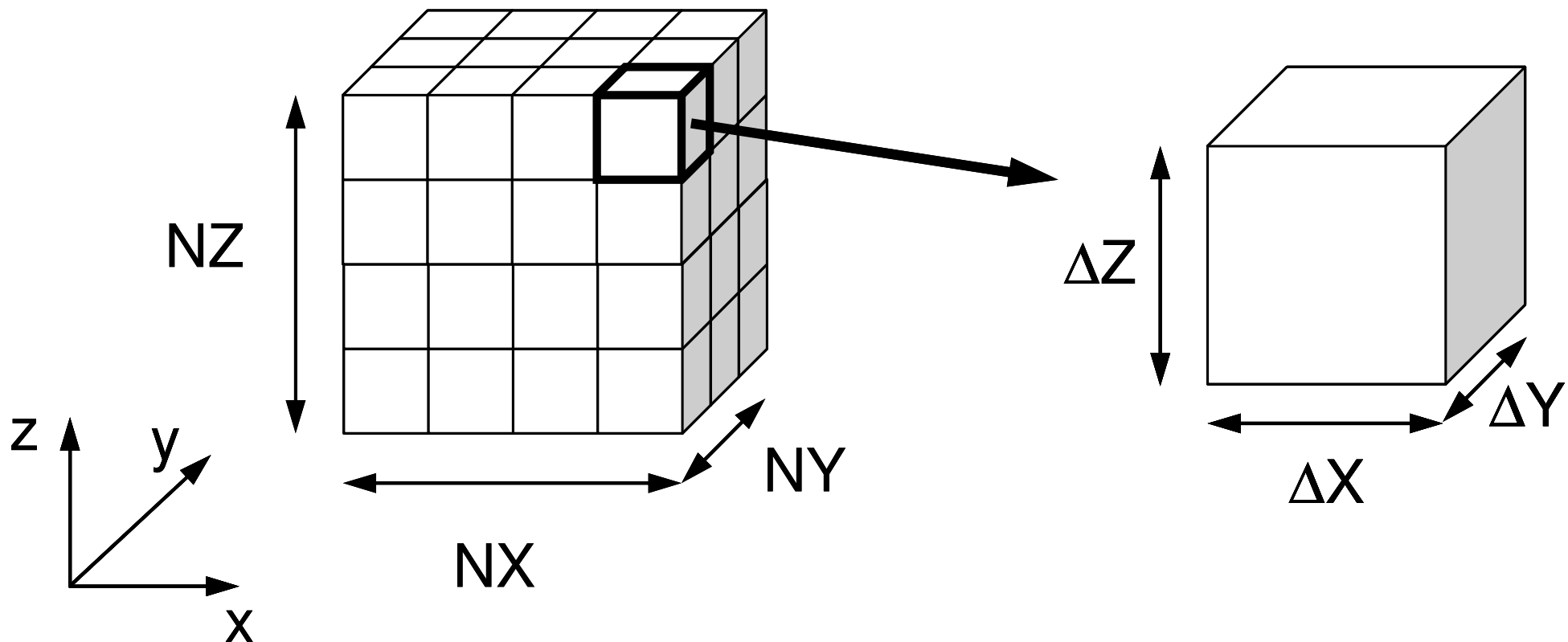
Boundary Conditions (B.C.) etc.

- Volume Flux
- $\phi=0 @ Z=Z_{\max}$



3D Structured Mesh

Internal data structure is “unstructured”



Volume Flux f

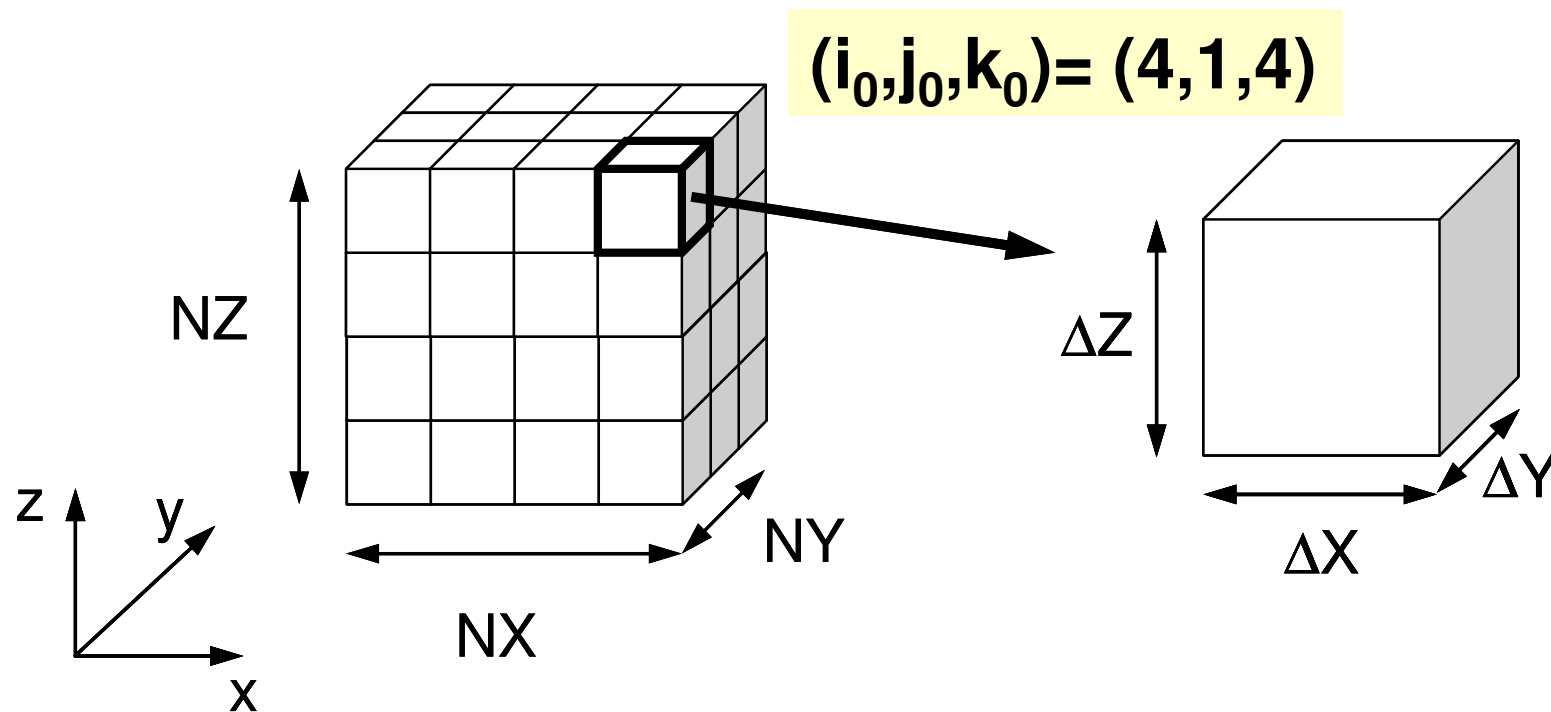
$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

$$f = dfloat(i_0 + j_0 + k_0)$$

$$i_0 = XYZ(icel, 1), \quad XYZ(icel, k) (k=1, 2, 3)$$

$$j_0 = XYZ(icel, 2), \quad \text{Index for location of finite-difference}$$

$$k_0 = XYZ(icel, 3) \quad \text{mesh in X-/Y-/Z-axis.}$$

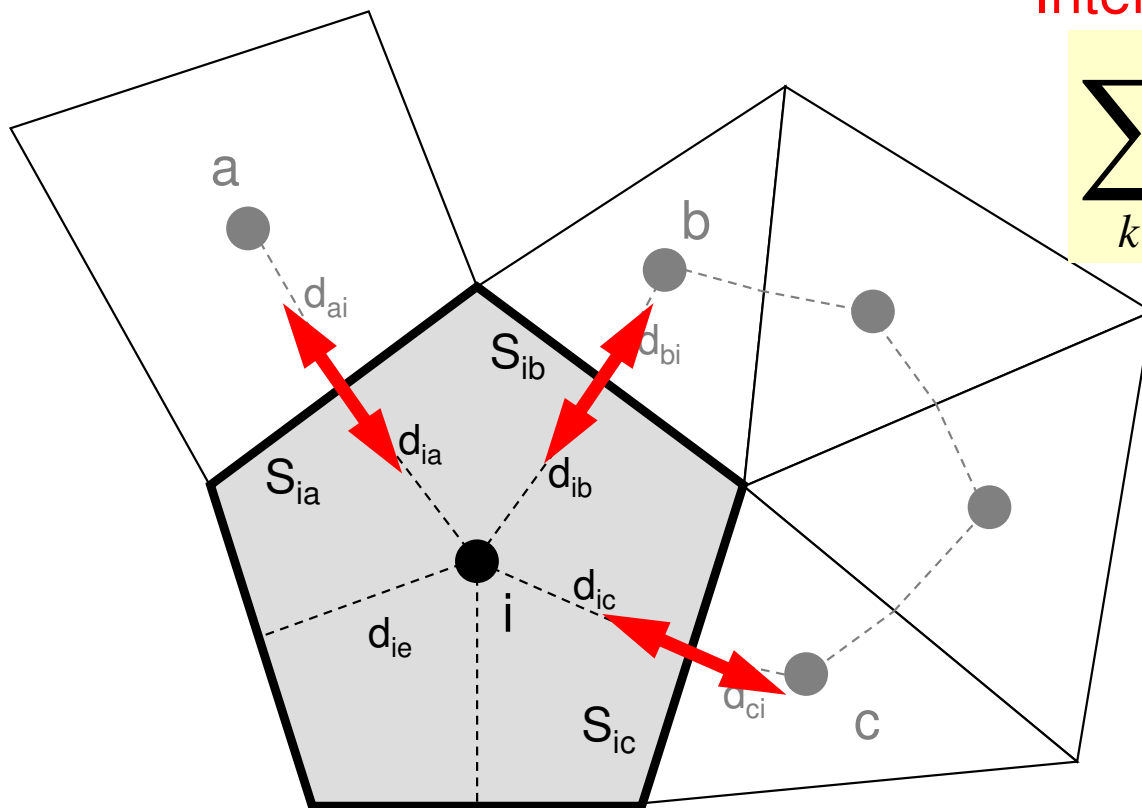


Poisson Equation by Finite Volume Method (FVM)

Conservation of Fluxes through
Surfaces

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

Diffusion:
Interaction with Neighbors



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

Volume Flux

V_i : Volume
 S : Surface Area
 d_{ij} : Distance between
 Cell-Center &
 Surface
 Q : Volume Flux

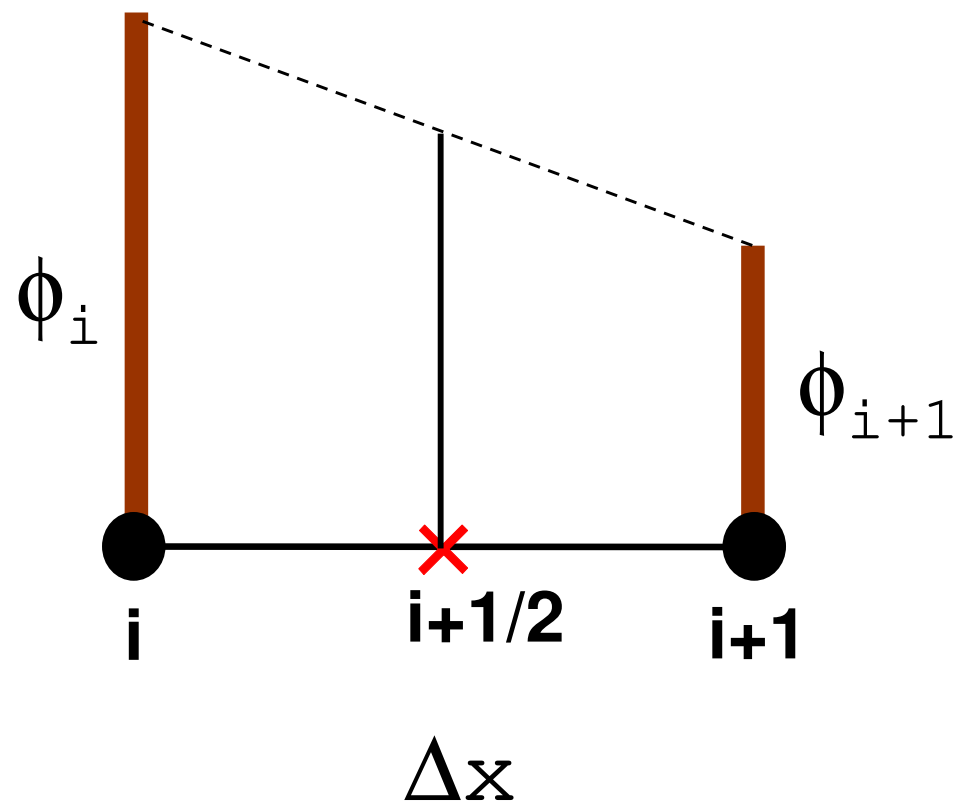
Finite Difference Method (FDM)

(有限)差分法：巨視的微分

macroscopic differentiation

$$\left(\frac{d\phi}{dx}\right)_{i+1/2} \approx \frac{\phi_{i+1} - \phi_i}{\Delta x}$$

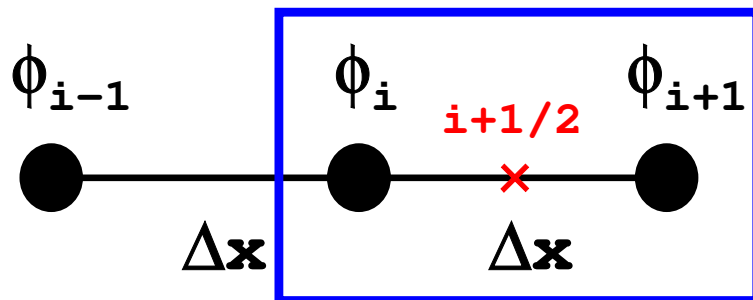
$$\left(\frac{d\phi}{dx}\right)_{i+1/2} = \lim_{\Delta x \rightarrow 0} \frac{\phi_{i+1} - \phi_i}{\Delta x}$$



2nd Order Differentiation in FDM

Taylor Series Expansion

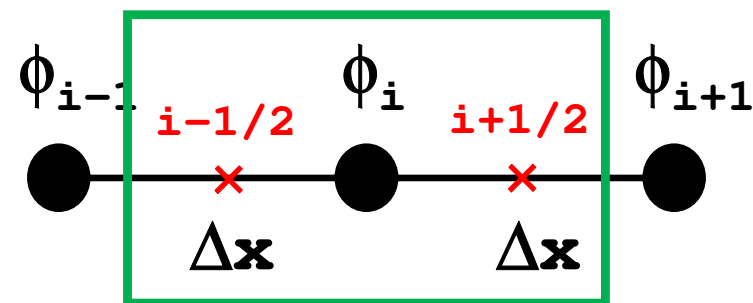
- Approximate Derivative at $\times$ (center of i and $i+1$)



$$\left(\frac{d\phi}{dx} \right)_{i+1/2} \approx \frac{\phi_{i+1} - \phi_i}{\Delta x}$$

$\Delta x \rightarrow 0$: Real Derivative

- 2nd-Order Diff. at i



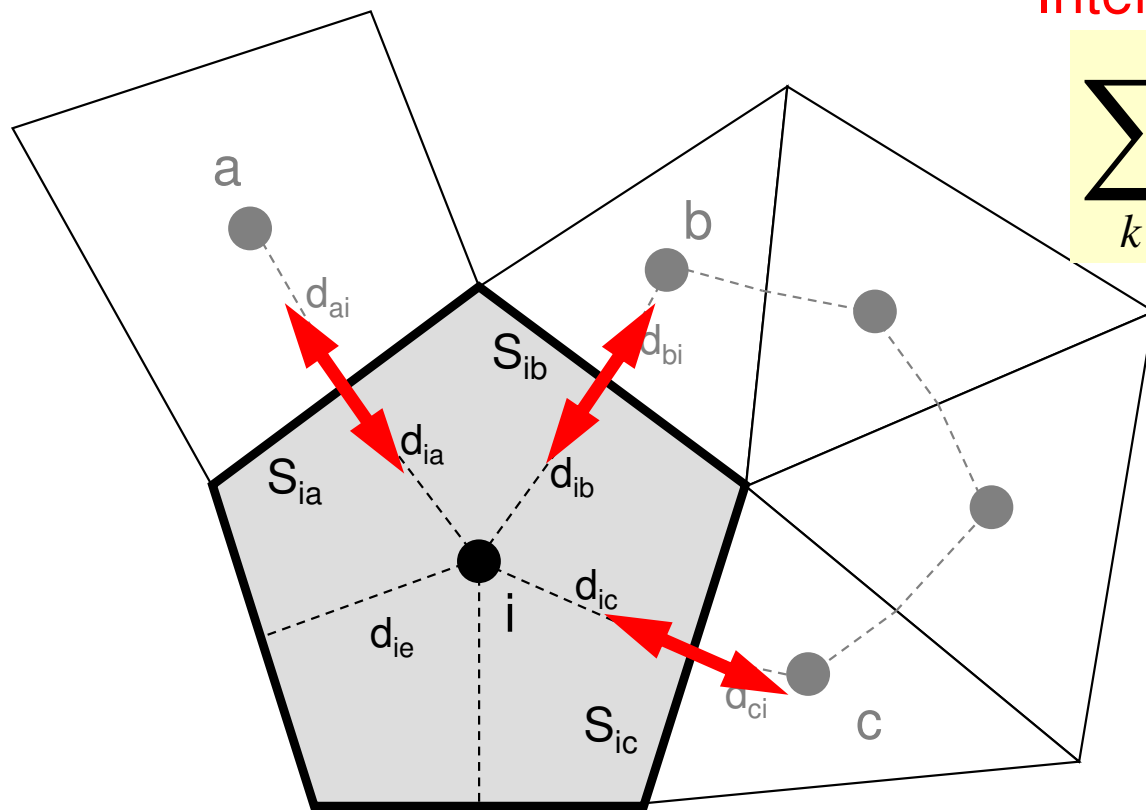
$$\left(\frac{d^2\phi}{dx^2} \right)_i \approx \frac{\left(\frac{d\phi}{dx} \right)_{i+1/2} - \left(\frac{d\phi}{dx} \right)_{i-1/2}}{\Delta x} = \frac{\frac{\phi_{i+1} - \phi_i}{\Delta x} - \frac{\phi_i - \phi_{i-1}}{\Delta x}}{\Delta x} = \frac{\phi_{i+1} - 2\phi_i + \phi_{i-1}}{\Delta x^2}$$

Poisson Equation by Finite Volume Method (FVM)

Conservation of Fluxes through
Surfaces

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

Diffusion:
Interaction with Neighbors

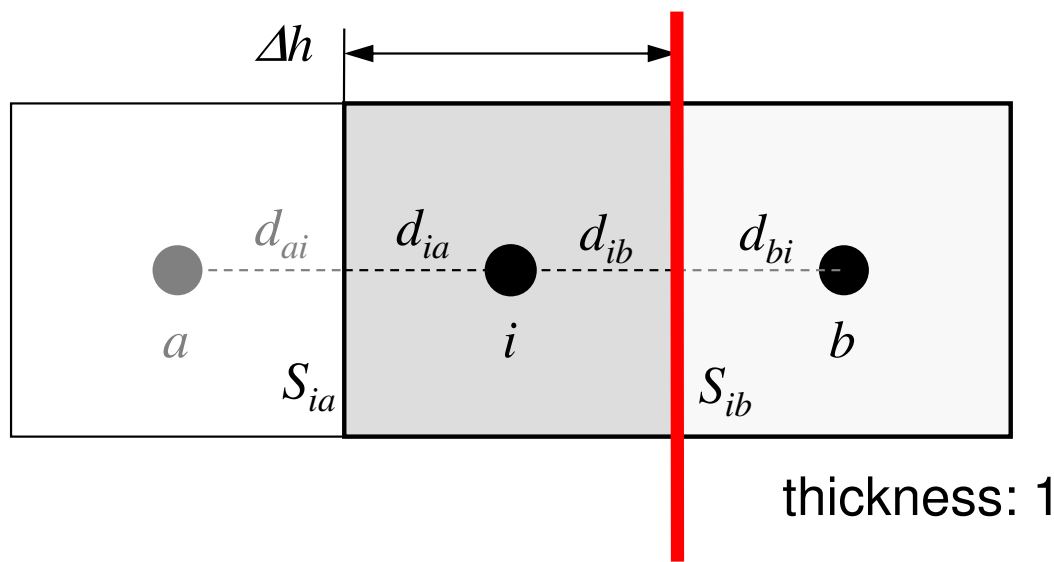


$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

Volume Flux

V_i : Volume
 S : Surface Area
 d_{ij} : Distance between
 Cell-Center &
 Surface
 Q : Volume Flux

Comparison with 1D FDM (1/3)



$\Delta h \times \Delta h$ Square Mesh

Surface Area: $S_{ik} = \Delta h$

Volume: $V_i = \Delta h^2$

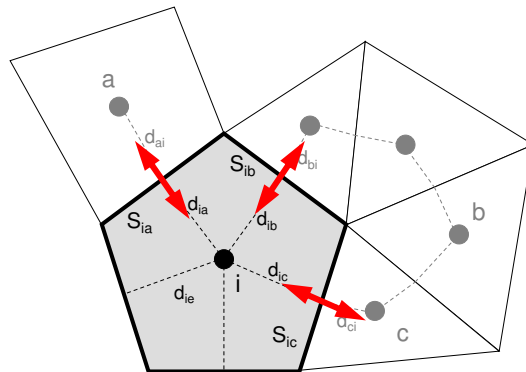
Distance (Ctr.-Suf): $d_{ij} = \Delta h/2$

Flux through this surface: $Q_{S_{ib}}$

$$Q_{S_{ib}} = -\frac{\phi_b - \phi_i}{d_{ib} + d_{bi}} \cdot S_{ib}$$

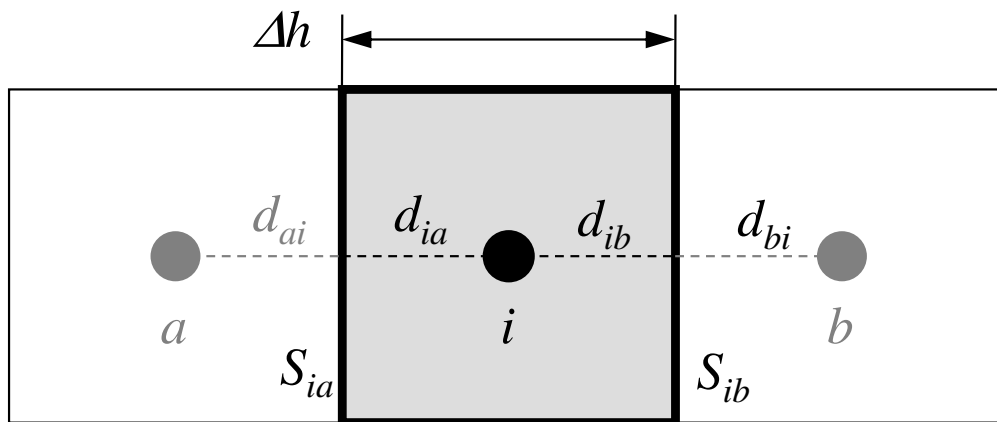
Fourier's Law

Flux through a surface
= - (gradient of potential)



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

Comparison with 1D FDM (2/3)



thickness: 1

$\Delta h \times \Delta h$ Square Mesh

Surface Area: $S_{ik} = \Delta h$

Volume: $V_i = \Delta h^2$

Distance (Ctr.-Suf): $d_{ij} = \Delta h/2$

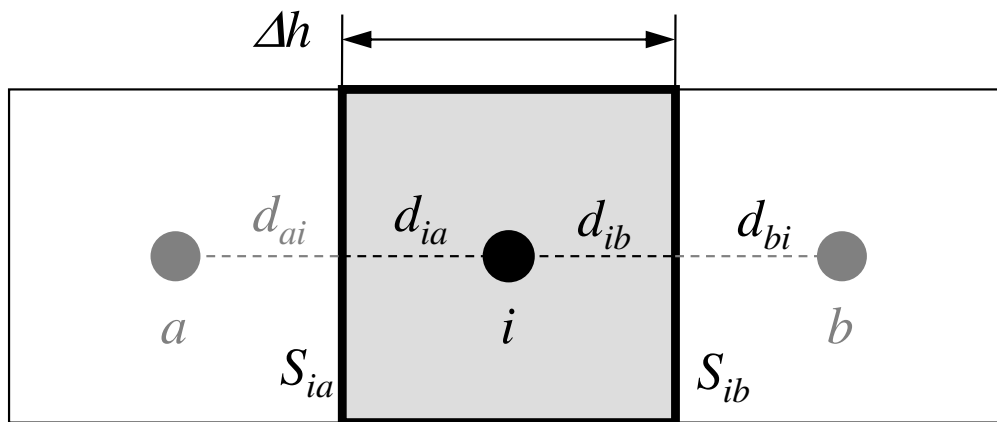
$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

Divided by V_i :

$$\frac{1}{V_i} \sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + \dot{Q}_i = 0$$

considering this part

Comparison with 1D FDM (3/3)



$\Delta h \times \Delta h$ Square Mesh

Surface Area: $S_{ik} = \Delta h$

Volume: $V_i = \Delta h^2$

Distance (Ctr.-Suf): $d_{ij} = \Delta h/2$

thickness: 1

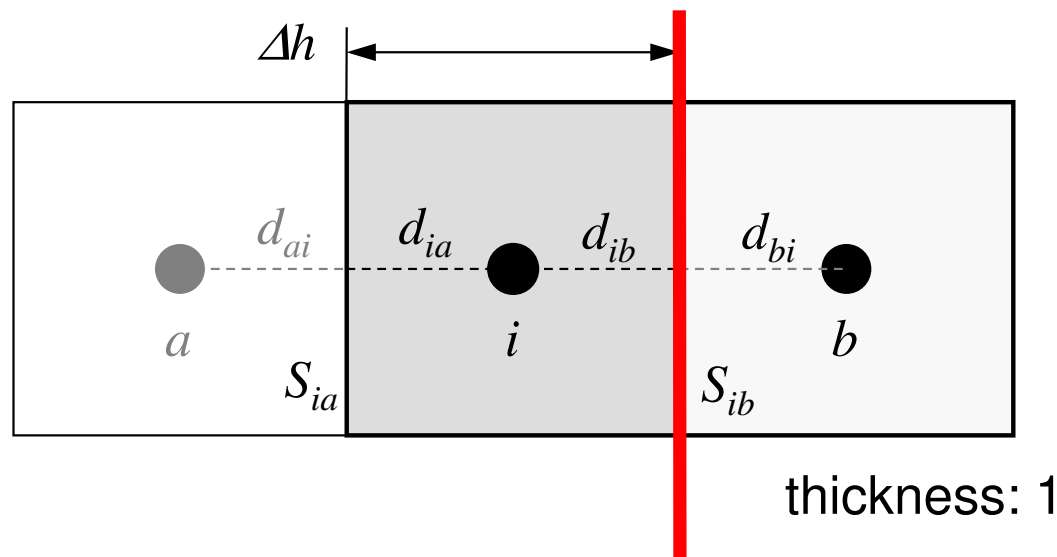
$$\begin{aligned}
 \frac{1}{V_i} \sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) &= \frac{1}{(\Delta h)^2} \sum_{k=a,b} \frac{\Delta h}{\frac{\Delta h}{2} + \frac{\Delta h}{2}} (\phi_k - \phi_i) \\
 &= \frac{1}{(\Delta h)^2} \sum_{k=a,b} \frac{\Delta h}{\frac{\Delta h}{2} + \frac{\Delta h}{2}} (\phi_k - \phi_i) = \frac{1}{(\Delta h)^2} \sum_{k=a,b} \frac{\Delta h}{\Delta h} (\phi_k - \phi_i) = \frac{1}{(\Delta h)^2} \sum_{k=a,b} (\phi_k - \phi_i) \\
 &= \frac{1}{(\Delta h)^2} (\phi_a - \phi_i) + \frac{1}{(\Delta h)^2} (\phi_b - \phi_i) = \boxed{\frac{\phi_a - 2\phi_i + \phi_b}{(\Delta h)^2}}
 \end{aligned}$$

for i-th cell
linear equations

Heat Equation (1/3)

$$\frac{\partial}{\partial x} \left(\lambda \frac{\partial T}{\partial x} \right) + \frac{\partial}{\partial y} \left(\lambda \frac{\partial T}{\partial y} \right) + \frac{\partial}{\partial z} \left(\lambda \frac{\partial T}{\partial z} \right) + Q = 0$$

λ : Thermal Conductivity



$\Delta h \times \Delta h$ Square Mesh

Surface Area: $S_{ik} = \Delta h$

Volume: $V_i = \Delta h^2$

Distance (Ctr.-Suf): $d_{ij} = \Delta h/2$

Heat Flux through this surface: $Q_{S_{ib}}$

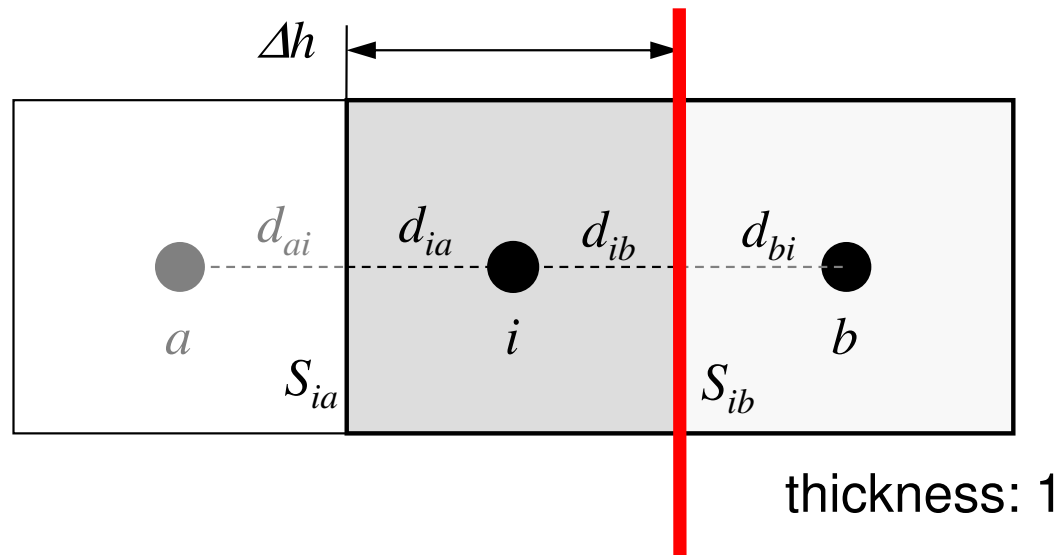
$$Q_{S_{ib}} = -\lambda \frac{T_b - T_i}{\Delta h} \cdot S_{ib}$$

$$\lambda_i = \lambda_b = \lambda$$

Heat Equation (2/3)

$$\frac{\partial}{\partial x} \left(\lambda \frac{\partial T}{\partial x} \right) + \frac{\partial}{\partial y} \left(\lambda \frac{\partial T}{\partial y} \right) + \frac{\partial}{\partial z} \left(\lambda \frac{\partial T}{\partial z} \right) + Q = 0$$

λ : Thermal Conductivity



$\Delta h \times \Delta h$ Square Mesh

Surface Area: $S_{ik} = \Delta h$

Volume: $V_i = \Delta h^2$

Distance (Ctr.-Suf): $d_{ij} = \Delta h/2$

Heat Flux through this surface: $Q_{S_{ib}}$

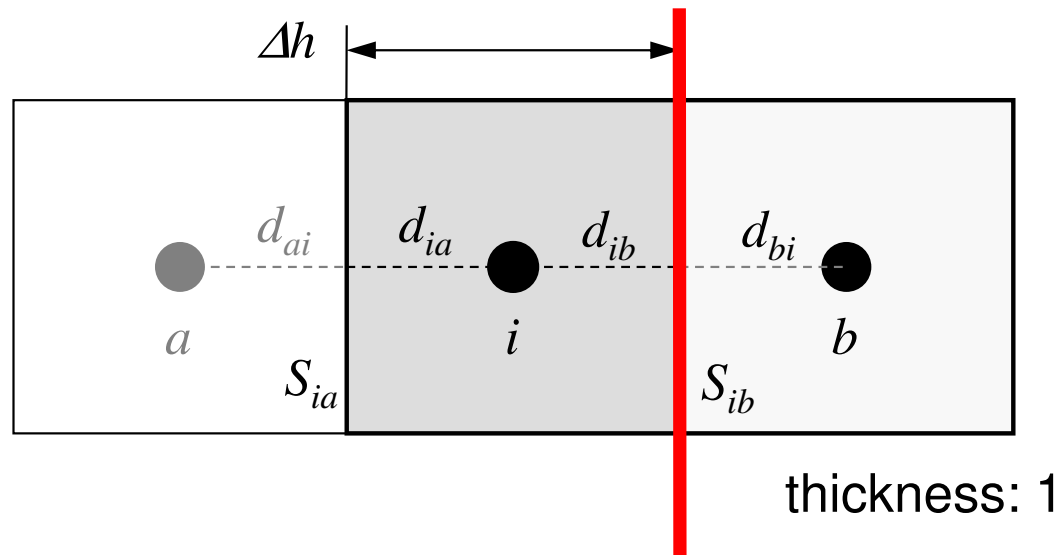
$$Q_{S_{ib}} = -\lambda \frac{T_b - T_i}{\frac{\Delta h}{2} + \frac{\Delta h}{2}} \cdot S_{ib} = -\frac{T_b - T_i}{\left[\left(\frac{\Delta h}{2} \right) / \lambda \right] + \left[\left(\frac{\Delta h}{2} \right) / \lambda \right]} \cdot S_{ib}$$

$$\lambda_i = \lambda_b = \lambda$$

Heat Equation (3/3)

$$\frac{\partial}{\partial x} \left(\lambda \frac{\partial T}{\partial x} \right) + \frac{\partial}{\partial y} \left(\lambda \frac{\partial T}{\partial y} \right) + \frac{\partial}{\partial z} \left(\lambda \frac{\partial T}{\partial z} \right) + Q = 0$$

λ : Thermal Conductivity



$\Delta h \times \Delta h$ Square Mesh

Surface Area: $S_{ik} = \Delta h$

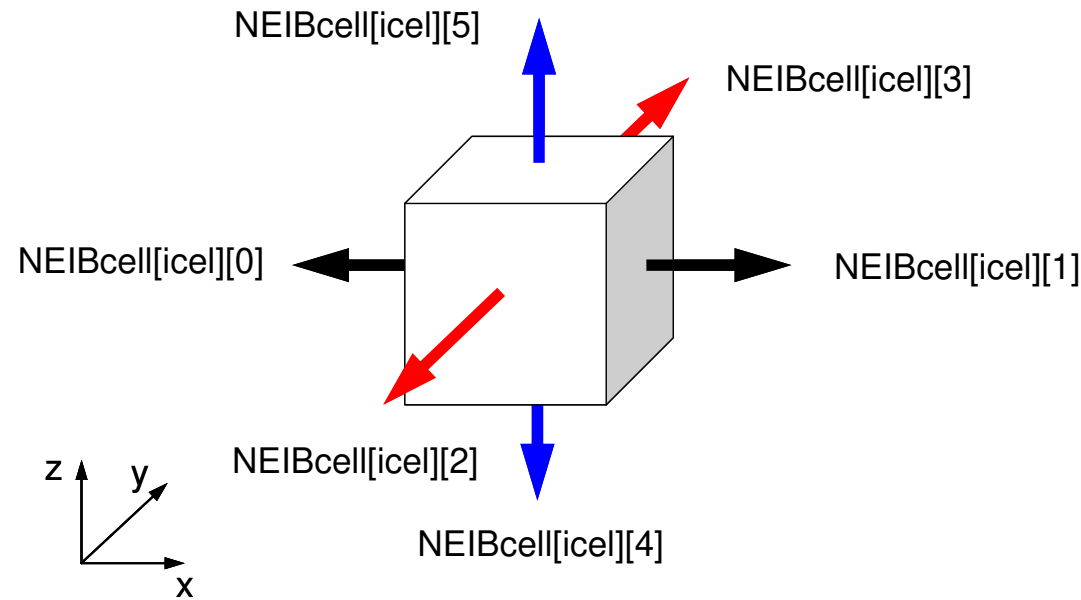
Volume: $V_i = \Delta h^2$

Distance (Ctr.-Suf): $d_{ij} = \Delta h/2$

Heat Flux through this surface: $Q_{S_{ib}}$

$$Q_{S_{ib}} = - \frac{T_b - T_i}{\left[\left(\frac{\Delta h}{2} \right) / \lambda_i \right] + \left[\left(\frac{\Delta h}{2} \right) / \lambda_b \right]} \cdot S_{ib} \quad \lambda_i \neq \lambda_b$$

in 3D



$$\begin{aligned}
 & \frac{\phi_{neib[icel][0]} - \phi_{icel}}{\Delta x} \Delta y \Delta z + \frac{\phi_{neib[icel][1]} - \phi_{icel}}{\Delta x} \Delta y \Delta z + \\
 & \frac{\phi_{neib[icel][2]} - \phi_{icel}}{\Delta y} \Delta z \Delta x + \frac{\phi_{neib[icel][3]} - \phi_{icel}}{\Delta y} \Delta z \Delta x + \\
 & \frac{\phi_{neib[icel][4]} - \phi_{icel}}{\Delta z} \Delta x \Delta y + \frac{\phi_{neib[icel][5]} - \phi_{icel}}{\Delta z} \Delta x \Delta y + f_{icel} \Delta x \Delta y \Delta z = 0
 \end{aligned}$$

Linear Equations

$$\begin{aligned} & \frac{\phi_{neib[icel][0]} - \phi_{icel}}{\Delta x} \Delta y \Delta z + \frac{\phi_{neib[icel][1]} - \phi_{icel}}{\Delta x} \Delta y \Delta z + \\ & \frac{\phi_{neib[icel][2]} - \phi_{icel}}{\Delta y} \Delta z \Delta x + \frac{\phi_{neib[icel][3]} - \phi_{icel}}{\Delta y} \Delta z \Delta x + \\ & \frac{\phi_{neib[icel][4]} - \phi_{icel}}{\Delta z} \Delta x \Delta y + \frac{\phi_{neib[icel][5]} - \phi_{icel}}{\Delta z} \Delta x \Delta y + f_{icel} \Delta x \Delta y \Delta z = 0 \end{aligned}$$

$$\sum_k \frac{S_{icel-k}}{d_{icel-k}} (\phi_k - \phi_{icel}) = -f_{icel} V_i$$

$$-\left[\sum_k \frac{S_{icel-k}}{d_{icel-k}} \right] \phi_{icel} + \left[\sum_k \frac{S_{icel-k}}{d_{icel-k}} \phi_k \right] = -f_{icel} V_i \quad (icel = 1, N)$$

Diagonal

Off-Diagonal



$$[A]\{\phi\} = \{f\}$$



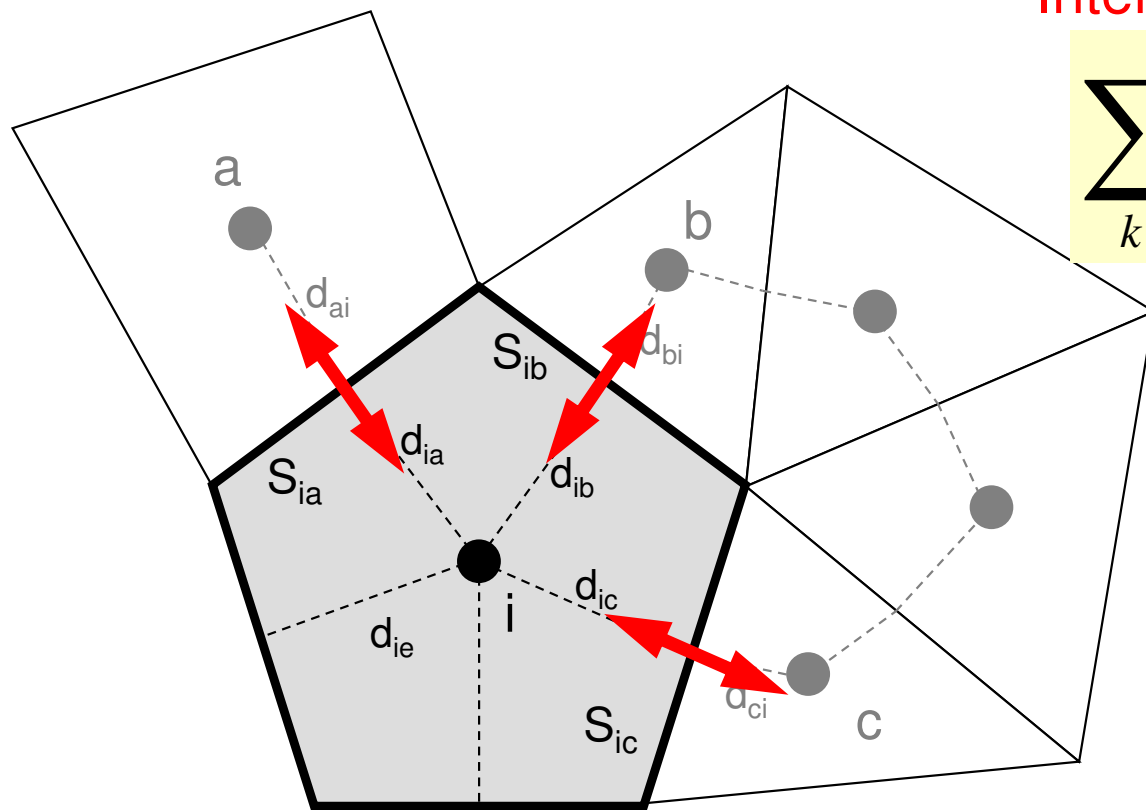
$$\begin{array}{cccccccccccc}
D & X & & & X & X & & & & & & \\
X & D & X & & X & X & X & & & & & \\
& & X & D & X & & X & X & X & & & \\
& & & X & D & & & X & X & & & \\
X & X & & & D & X & & & X & X & & \\
X & X & X & & X & D & X & & X & X & X & \\
\boxed{X} & \boxed{X} & \boxed{X} & & \boxed{X} & \boxed{D} & \boxed{X} & & \boxed{X} & \boxed{X} & \boxed{X} & \\
& & X & X & & X & D & & X & X & & \\
& & & & X & X & & D & X & & X & X \\
& & & & X & X & X & & X & D & X & \\
& & & & X & X & X & & X & D & X & \\
& & & & & & X & X & & D & X & \\
& & & & & & X & X & X & & X & D \\
& & & & & & & & X & X & & D
\end{array}
\left\{ \begin{array}{l} \Phi_1 \\ \Phi_2 \\ \Phi_3 \\ \Phi_4 \\ \Phi_5 \\ \Phi_6 \\ \Phi_7 \\ \Phi_8 \\ \Phi_9 \\ \Phi_{10} \\ \Phi_{11} \\ \Phi_{12} \\ \Phi_{13} \\ \Phi_{14} \\ \Phi_{15} \\ \Phi_{16} \end{array} \right\} = \left\{ \begin{array}{l} F_1 \\ F_2 \\ F_3 \\ F_4 \\ F_5 \\ F_6 \\ F_7 \\ F_8 \\ F_9 \\ F_{10} \\ F_{11} \\ F_{12} \\ F_{13} \\ F_{14} \\ F_{15} \\ F_{16} \end{array} \right\}$$

Coefficient Matrices for FVM are sparse

Only neighboring cells are considered

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

Diffusion:
Interaction with Neighbors



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

Volume Flux

V_i : Volume
 S : Surface Area
 d_{ij} : Distance between
 Cell-Center &
 Surface
 Q : Volume Flux

Sparse Matrix for FVM

- Sparse Matrix
 - Many “0”s
- Storing all components (e.g. $A(i,j)$) is not efficient for sparse matrices
 - $A(i,j)$ is suitable for dense matrices

$$\begin{bmatrix}
 D & X & & & X & X & & & & & & & & & & & \\
 X & D & X & & X & X & X & & & & & & & & & & \\
 & & X & D & X & & X & X & X & & & & & & & \\
 & & & X & D & & & X & X & & & & & & & \\
 X & X & & & D & X & & & X & X & & & & & & \\
 X & X & X & & X & D & X & & X & X & X & & & & & \\
 & & X & X & X & & X & D & X & & X & X & X & & & \\
 & & & X & X & & & X & D & & & X & X & & & \\
 & & & & X & X & & & D & X & & & X & X & & \\
 & & & & & X & X & & X & D & X & & & X & X & \\
 & & & & & & X & X & & D & X & & & & X & \\
 & & & & & & & X & X & X & & X & D & X & & \\
 & & & & & & & & X & X & & & X & D & X & \\
 & & & & & & & & & X & X & & & X & D & \\
 & & & & & & & & & & X & X & & & X & D
 \end{bmatrix}
 \begin{Bmatrix}
 \Phi_1 \\ \Phi_2 \\ \Phi_3 \\ \Phi_4 \\ \Phi_5 \\ \Phi_6 \\ \Phi_7 \\ \Phi_8 \\ \Phi_9 \\ \Phi_{10} \\ \Phi_{11} \\ \Phi_{12} \\ \Phi_{13} \\ \Phi_{14} \\ \Phi_{15} \\ \Phi_{16}
 \end{Bmatrix}
 =
 \begin{Bmatrix}
 F_1 \\ F_2 \\ F_3 \\ F_4 \\ F_5 \\ F_6 \\ F_7 \\ F_8 \\ F_9 \\ F_{10} \\ F_{11} \\ F_{12} \\ F_{13} \\ F_{14} \\ F_{15} \\ F_{16}
 \end{Bmatrix}$$

- Number of non-zero off-diagonal components is $O(100)$ in FVM (only 6 in this case)
 - If number of unknowns is 10^8 :
 - $A(i,j)$: $O(10^{16})$ words $\sim O(10^{17})$ bytes for DP: 100PB (400 x Odyssey)
 - Actual Non-zero Comp.: $O(10^9)$ words: 10GB (Odyssey: 32GB/node)
- Only (really) non-zero off-diag. components should be stored on memory

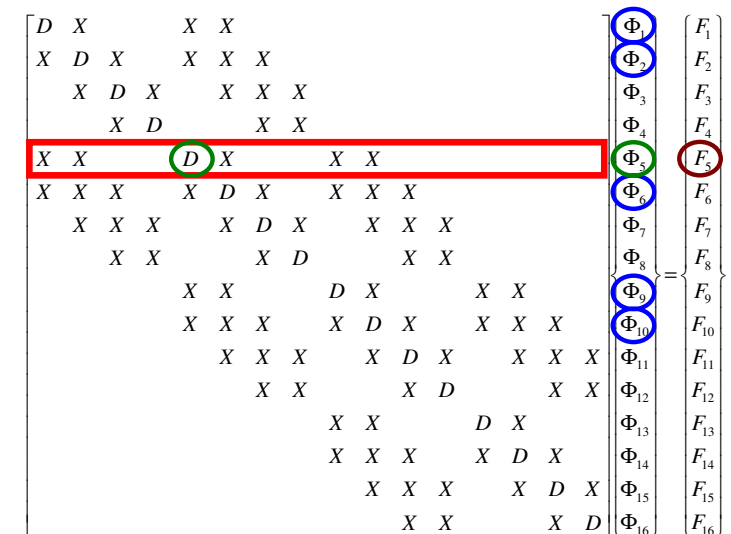
Mat-Vec. Multiplication for Sparse Matrix

Compressed Row Storage (CRS)

Diag [i] Diagonal Components (REAL, $i=1 \sim N$)
Index [i] Number of Non-Zero Off-Diagonals at Each ROW (INT, $i=0 \sim N$)
Item [k] Off-Diagonal Components (Corresponding Column ID)
 (INT, $k=0, \text{index}[N]$)
Amat [k] Off-Diagonal Components (Value)
 (REAL, $k=0, \text{index}[N]$)

$$\{Y\} = [A] \{X\}$$

```
for (i=0; i<N; i++) {
    Y[i] = Diag[i] * X[i];
    for (k=Index[i]; k<Index[i+1]; k++) {
        Y[i] += Amat[k]*X[Item[k]];
    }
}
```



Mat-Vec. Multiplication for Sparse Matrix

Compressed Row Storage (CRS)

```
{Q} = [A] {P}
```

```
for (i=0; i<N; i++) {  
    W[Q][i] = Diag[i] * W[P][i];  
    for (k=Index[i]; k<Index[i+1]; k++) {  
        W[Q][i] += AMat[k]*W[P][Item[k]];  
    }  
}
```

Mat-Vec. Multiplication for Dense Matrix

Very Easy, Straightforward

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1,N-1} & a_{1,N} \\ a_{21} & a_{22} & & a_{2,N-1} & a_{2,N} \\ \dots & & \dots & & \dots \\ a_{N-1,1} & a_{N-1,2} & & a_{N-1,N-1} & a_{N-1,N} \\ a_{N,1} & a_{N,2} & \dots & a_{N,N-1} & a_{N,N} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{N-1} \\ x_N \end{Bmatrix} = \begin{Bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{N-1} \\ y_N \end{Bmatrix}$$

$$\{Y\} = [A] \{X\}$$

```
for (j=0; j<N; j++) {
    Y[j] = 0.0;
    for (i=0; i<N; i++) {
        Y[j] += A[j][i]*X[i];
    }
}
```

Compressed Row Storage (CRS)

	1	2	3	4	5	6	7	8
1	1.1	2.4	0	0	3.2	0	0	0
2	4.3	3.6	0	2.5	0	3.7	0	9.1
3	0	0	5.7	0	1.5	0	3.1	0
4	0	4.1	0	9.8	2.5	2.7	0	0
5	3.1	9.5	10.4	0	11.5	0	4.3	0
6	0	0	6.5	0	0	12.4	9.5	0
7	0	6.4	2.5	0	0	1.4	23.1	13.1
8	0	9.5	1.3	9.6	0	3.1	0	51.3

Compressed Row Storage (CRS): C

Numbering starts from 0 in program

	0	1	2	3	4	5	6	7
0	1.1 ⊙	2.4 ①			3.2 ④			
1	4.3 ⊙	3.6 ①		2.5 ③		3.7 ⑤		9.1 ⑦
2			5.7 ②		1.5 ④		3.1 ⑥	
3		4.1 ①		9.8 ③	2.5 ④	2.7 ⑤		
4	3.1 ⊙	9.5 ①	10.4 ②		11.5 ④		4.3 ⑥	
5			6.5 ②			12.4 ⑤	9.5 ⑥	
6		6.4 ①	2.5 ②			1.4 ⑤	23.1 ⑥	13.1 ⑦
7		9.5 ①	1.3 ②	9.6 ③		3.1 ⑤		51.3 ⑦

N= 8

Diagonal Components

Diag[0]= 1.1

Diag[1]= 3.6

Diag[2]= 5.7

Diag[3]= 9.8

Diag[4]= 11.5

Diag[5]= 12.4

Diag[6]= 23.1

Diag[7]= 51.3

Compressed Row Storage (CRS)

	0	1	2	3	4	5	6	7
0	1.1 ⊙ ①		2.4 ①			3.2 ④		
1	3.6 ①	4.3 ⊙			2.5 ③		3.7 ⑤	9.1 ⑦
2	5.7 ②					1.5 ④		3.1 ⑥
3	9.8 ③		4.1 ①			2.5 ④	2.7 ⑤	
4	11.5 ④	3.1 ⊙	9.5 ①	10.4 ②				4.3 ⑥
5	12.4 ⑤			6.5 ②				9.5 ⑥
6	23.1 ⑥		6.4 ①	2.5 ②			1.4 ⑤	13.1 ⑦
7	51.3 ⑦		9.5 ①	1.3 ②	9.6 ③		3.1 ⑤	

Compressed Row Storage (CRS)

		# Non-Zero Off-Diag.	
0	1.1 ⊙ 2.4 ① 3.2 ④	2	Index[0] = 0
1	3.6 ① 4.3 ⊙ 2.5 ③ 3.7 ⑤ 9.1 ⑦	4	Index[1] = 2
2	5.7 ② 1.5 ④ 3.1 ⑥	2	Index[2] = 6
3	9.8 ③ 4.1 ① 2.5 ④ 2.7 ⑤	3	Index[3] = 8
4	11.5 ④ 3.1 ⊙ 9.5 ① 10.4 ② 4.3 ⑥	4	Index[4] = 11
5	12.4 ⑤ 6.5 ② 9.5 ⑥	2	Index[5] = 15
6	23.1 ⑥ 6.4 ① 2.5 ② 1.4 ⑤ 13.1 ⑦	4	Index[6] = 17
7	51.3 ⑦ 9.5 ① 1.3 ② 9.6 ③ 3.1 ⑤	4	Index[7] = 21
			Index[8] = 25

NPLU= 25
(=Index[N])

$(\text{Index}[i])^{\text{th}} \sim (\text{Index}[i+1])^{\text{th}} :$

Non-Zero Off-Diag. Components corresponding to i -th row.

Compressed Row Storage (CRS)

		# Non-Zero Off-Diag.	Index [0] = 0
0	1.1 ⊙ 2.4 ①,0 3.2 ④,1	2	Index [1] = 2
1	3.6 ① 4.3 ⊙,2 2.5 ③,3 3.7 ⑤,4 9.1 ⑦,5	4	Index [2] = 6
2	5.7 ② 1.5 ④,6 3.1 ⑥,7	2	<u>Index [3] = 8</u>
3	9.8 ③ 4.1 ①,8 2.5 ④,9 2.7 ⑤,10	3	<u>Index [4] = 11</u>
4	11.5 ④ 3.1 ⊙,11 9.5 ①,12 10.4 ②,13 4.3 ⑥,14	4	Index [5] = 15
5	12.4 ⑤ 6.5 ②,15 9.5 ⑥,16	2	Index [6] = 17
6	23.1 ⑥ 6.4 ①,17 2.5 ②,18 1.4 ⑤,19 13.1 ⑦,20	4	Index [7] = 21
7	51.3 ⑦ 9.5 ①,21 1.3 ②,22 9.6 ③,23 3.1 ⑤,24	4	Index [8] = 25

NPLU= 25
(=Index[N])

$(\text{Index}[i])^{\text{th}} \sim (\text{Index}[i+1])^{\text{th}}$:

Non-Zero Off-Diag. Components corresponding to i -th row.

Compressed Row Storage (CRS)

0	1.1 ⊙	2.4 ①,0	3.2 ④,1		
1	3.6 ①	4.3 ⊙,2	2.5 ③,3	3.7 ⑤,4	9.1 ⑦,5
2	5.7 ②	1.5 ④,6	3.1 ⑥,7		
3	9.8 ③	4.1 ①,8	2.5 ④,9	2.7 ⑤,10	
4	11.5 ④	3.1 ⊙,11	9.5 ①,12	10.4 ②,13	4.3 ⑥,14
5	12.4 ⑤	6.5 ②,15	9.5 ⑥,16		
6	23.1 ⑥	6.4 ①,17	2.5 ②,18	1.4 ⑤,19	13.1 ⑦,20
7	51.3 ⑦	9.5 ①,21	1.3 ②,22	9.6 ③,23	3.1 ⑤,24

Item[6] = 4, AMat[6] = 1.5

Item[18] = 2, AMat[18] = 2.5

Compressed Row Storage (CRS)

0	1.1 ⊙	2.4 ①,0	3.2 ④,1		
1	3.6 ①	4.3 ⊙,2	2.5 ③,3	3.7 ⑤,4	9.1 ⑦,5
2	5.7 ②	1.5 ④,6	3.1 ⑥,7		
3	9.8 ③	4.1 ①,8	2.5 ④,9	2.7 ⑤,10	
4	11.5 ④	3.1 ⊙,11	9.5 ①,12	10.4 ②,13	4.3 ⑥,14
5	12.4 ⑤	6.5 ②,15	9.5 ⑥,16		
6	23.1 ⑥	6.4 ①,17	2.5 ②,18	1.4 ⑤,19	13.1 ⑦,20
7	51.3 ⑦	9.5 ①,21	1.3 ②,22	9.6 ③,23	3.1 ⑤,24

Diag [i] Diagonal Components (REAL, i=0~N-1)
 Index [i] Number of Non-Zero Off-Diagonals at
 Each ROW (INT, i=0~N)
 Item [k] Off-Diagonal Components
 (Corresponding Column ID)
 (INT, k=0, index[N])
 Amat [k] Off-Diagonal Components (Value)
 (REAL, k=0, index[N])

$$\{Y\} = [A] \{X\}$$

```

for (i=0; i<N; i++) {
    Y[i] = Diag[i] * X[i];
    for (k=Index[i]; k<Index[i+1]; k++) {
        Y[i] += Amat[k]*X[Item[k]];
    }
}
  
```

Sparse Matrix: Only non-zero components are stored

Indirect Access: Memory-Bound

```
{Y}=[A] {X}
```

```
for (i=0; i<N; i++) {  
    Y[i] = Diag[i] * X[i];  
    for (k=Index[i]; k<Index[i+1]; k++) {  
        Y[i] += AMat[k]*X[Item[k]];  
    }  
}
```

Dense Matrix: Continuous Access

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1,N-1} & a_{1,N} \\ a_{21} & a_{22} & & a_{2,N-1} & a_{2,N} \\ \dots & & \dots & & \dots \\ a_{N-1,1} & a_{N-1,2} & & a_{N-1,N-1} & a_{N-1,N} \\ a_{N,1} & a_{N,2} & \dots & a_{N,N-1} & a_{N,N} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_{N-1} \\ x_N \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{N-1} \\ y_N \end{bmatrix}$$

$$\{Y\} = [A] \{X\}$$

```
for (j=0; j<N; j++) {
    Y[j] = 0.0;
    for (i=0; i<N; i++) {
        Y[j] += A[j][i]*X[i];
    }
}
```

- Background
 - Finite Volume Method
 - **Preconditioned Iterative Solvers**
- PCG Solver for Poisson's Equations
 - How to run
 - Data Structure
 - Program
 - Initialization
 - Coefficient Matrices
 - PCG

Large-Scale Linear Equations in Scientific Applications

- Solving large-scale linear equations $\mathbf{Ax}=\mathbf{b}$ is the most important and **expensive** part of various types of scientific computing.
 - for both linear and nonlinear applications
- Various types of methods proposed & developed.
 - for dense and sparse matrices
 - classified into **direct** and **iterative** methods
- Dense Matrices: 密行列: Globally Coupled Problems
 - BEM, Spectral Methods, MO/MD (gas, liquid)
- Sparse Matrices: 疎行列: Locally Defined Problems
 - FEM, FVM, FDM, DEM, MD (solid), BEM w/FMM

Direct Method

直接法

- Gaussian Elimination/LU Factorization
 - compute A^{-1} directly (or equivalent operations)

Good

- Robust for wide range of applications.
- Good for both dense and sparse matrices

Bad

- More expensive than iterative methods (memory, CPU)
 - not scalable

What is Iterative Method ?

反復法

Linear Equations
連立一次方程式

$$\begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{pmatrix}$$

A **x** **b**

Initial Solution
初期解

$$\mathbf{x}^{(0)} = \begin{pmatrix} x_1^{(0)} \\ x_2^{(0)} \\ \vdots \\ x_n^{(0)} \end{pmatrix}$$

Starting from a initial vector $\mathbf{x}^{(0)}$, iterative method obtains the final converged solutions by iterations

$$\mathbf{x}^{(1)}, \mathbf{x}^{(2)}, \cdots$$

Iterative Method

反復法

- Stationary Method

- Only $\mathbf{x}$ (solution vector) changes during iterations.
- SOR, Gauss-Seidel, Jacobi
- Generally slow, impractical

$$\mathbf{Ax} = \mathbf{b} \Rightarrow$$
$$\mathbf{x}^{(k+1)} = \mathbf{M}\mathbf{x}^{(k)} + \mathbf{Nb}$$

- Non-Stationary Method

- With restriction/optimization conditions
- Krylov-Subspace
- CG: Conjugate Gradient
- BiCGSTAB: Bi-Conjugate Gradient Stabilized
- GMRES: Generalized Minimal Residual

Iterative Method (cont.)

Good

- Less expensive than direct methods, especially in memory.
- Suitable for parallel and vector computing.

Bad

- Convergence strongly depends on problems, boundary conditions (condition number etc.)
- Preconditioning is required : Key Technology for Real-World Applications in Scientific Computing (FEM, FVM, FDM etc)

Non-Stationary/Krylov Subspace Method (1/2)

非定常法・クリロフ部分空間法

$$\mathbf{Ax} = \mathbf{b} \Rightarrow \mathbf{x} = \mathbf{b} + (\mathbf{I} - \mathbf{A})\mathbf{x}$$

Compute $\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_k$ by the following iterative procedures:

$$\begin{aligned}\mathbf{x}_k &= \mathbf{b} + (\mathbf{I} - \mathbf{A})\mathbf{x}_{k-1} \\ &= (\mathbf{b} - \mathbf{Ax}_{k-1}) + \mathbf{x}_{k-1}\end{aligned}$$

$$= \mathbf{r}_{k-1} + \mathbf{x}_{k-1}$$

where $\mathbf{r}_k = \mathbf{b} - \mathbf{Ax}_k$: residual



$$\mathbf{x}_k = \mathbf{x}_0 + \sum_{i=0}^{k-1} \mathbf{r}_i$$

$$\begin{aligned}\mathbf{r}_k &= \mathbf{b} - \mathbf{Ax}_k = \mathbf{b} - \mathbf{A}(\mathbf{r}_{k-1} + \mathbf{x}_{k-1}) \\ &= (\mathbf{b} - \mathbf{Ax}_{k-1}) - \mathbf{Ar}_{k-1} = \mathbf{r}_{k-1} - \mathbf{Ar}_{k-1} = (\mathbf{I} - \mathbf{A})\mathbf{r}_{k-1}\end{aligned}$$

Non-Stationary/Krylov Subspace Method (2/2)

非定常法・クリロフ部分空間法

$$\mathbf{x}_k = \mathbf{x}_0 + \sum_{i=0}^{k-1} \mathbf{r}_i = \mathbf{x}_0 + \mathbf{r}_0 + \sum_{i=0}^{k-2} (\mathbf{I} - \mathbf{A}) \mathbf{r}_i = \mathbf{x}_0 + \mathbf{r}_0 + \sum_{i=1}^{k-1} (\mathbf{I} - \mathbf{A})^i \mathbf{r}_0$$

$$\mathbf{z}_k = \mathbf{r}_0 + \sum_{i=1}^{k-1} (\mathbf{I} - \mathbf{A})^i \mathbf{r}_0 = \left[\mathbf{I} + \sum_{i=1}^{k-1} (\mathbf{I} - \mathbf{A})^i \right] \mathbf{r}_0$$



$\mathbf{z}_k$ is a vector which belongs to k^{th} Krylov Subspace (クリロフ部分空間), approximate solution vector $\mathbf{x}_k$ is derived by the Krylov Subspace:

$$\left[\mathbf{r}_0, \mathbf{A}\mathbf{r}_0, \mathbf{A}^2\mathbf{r}_0, \dots, \mathbf{A}^{k-1}\mathbf{r}_0 \right]$$

Conjugate Gradient Method

共役勾配法

- Conjugate Gradient: CG
 - Most popular “non-stationary” iterative method
- for Symmetric Positive Definite (SPD) Matrices
 - 対称正定
 - $\{x\}^T[A]\{x\} > 0$ for arbitrary $\{x\}$
 - All of diagonal components, eigenvalues and leading principal minors > 0 (主小行列式・首座行列式)
 - Matrices of Galerkin-based FEM & FVM: heat conduction, Poisson, static linear elastic problems
- Algorithm
 - “Steepest Descent Method”
 - $\mathbf{x}^{(i)} = \mathbf{x}^{(i-1)} + \alpha_i \mathbf{p}^{(i)}$
 - $\mathbf{x}^{(i)}$: solution, $\mathbf{p}^{(i)}$: search direction, α_i : coefficient
 - Solution $\{x\}$ minimizes $\{x-y\}^T[A]\{x-y\}$, where $\{y\}$ is exact solution.

$$\det \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & a_{24} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & a_{34} & \cdots & a_{3n} \\ a_{41} & a_{42} & a_{43} & a_{44} & \cdots & a_{4n} \\ \vdots & \vdots & \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & a_{n3} & a_{n4} & \cdots & a_{nn} \end{bmatrix}$$

Procedures of Conjugate Gradient

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
     $z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

- Mat-Vec. Multiplication
- Dot Products
- DAXPY (Double Precision: $a\{X\} + \{Y\}$)

$x^{(i)}$: Vector

α_i : Scalar

Procedures of Conjugate Gradient

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
     $z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

- Mat-Vec. Multiplication
- Dot Products
- DAXPY

$x^{(i)}$: Vector

α_i : Scalar

Procedures of Conjugate Gradient

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
     $z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} \cdot z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / (p^{(i)} \cdot q^{(i)})$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

- Mat-Vec. Multiplication
- Dot Products
- DAXPY

$x^{(i)}$: Vector

α_i : Scalar

Procedures of Conjugate Gradient

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
     $z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

- Mat-Vec. Multiplication
- Dot Products
- DAXPY
 - Double
 - $\{y\} = a\{x\} + \{y\}$

$x^{(i)}$: Vector

α_i : Scalar

Procedures of Conjugate Gradient

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
     $z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

$x^{(i)}$: Vector

α_i : Scalar

Derivation of CG Algorithm (1/5)

Solution x minimizes the following equation if y is the exact solution ($Ay=b$)

$$(x - y)^T [A](x - y)$$

$$\begin{aligned} (x - y)^T [A](x - y) &= (x, Ax) - (y, Ax) - (x, Ay) + (y, Ay) \\ &= (x, Ax) - 2(x, Ay) + (y, Ay) = (x, Ax) - 2(x, b) + \underline{(y, b)} \quad \text{Const.} \end{aligned}$$

Therefore, the solution x minimizes the following $f(x)$:

$$f(x) = \frac{1}{2}(x, Ax) - (x, b)$$

$$f(x + h) = f(x) + (h, Ax - b) + \frac{1}{2}(h, Ah) \quad \text{Arbitrary vector } h$$

$$f(x) = \frac{1}{2}(x, Ax) - (x, b)$$

$$f(x+h) = f(x) + (h, Ax-b) + \frac{1}{2}(h, Ah)$$

Arbitrary vector h

$$\begin{aligned} f(x+h) &= \frac{1}{2}(x+h, A(x+h)) - (x+h, b) \\ &= \frac{1}{2}(x+h, Ax) + \frac{1}{2}(x+h, Ah) - (x, b) - (h, b) \\ &= \frac{1}{2}(x, Ax) + \frac{1}{2}(h, Ax) + \frac{1}{2}(x, Ah) + \frac{1}{2}(h, Ah) - (x, b) - (h, b) \\ &= \frac{1}{2}(x, Ax) - (x, b) + (h, Ax) - (h, b) + \frac{1}{2}(h, Ah) \\ &= f(x) + (h, Ax-b) + \frac{1}{2}(h, Ah) \end{aligned}$$

Derivation of CG Algorithm (2/5)

CG method minimizes $f(x)$ at each iteration.

Start from initial solution $x^{(0)}$ and assume that approximate solution: $x^{(k)}$, and search direction vector $p^{(k)}$ is defined at k -th iter.

$$x^{(k+1)} = x^{(k)} + \alpha_k p^{(k)}$$

Minimization of $f(x^{(k+1)})$ is done as follows:

$$f(x^{(k)} + \alpha_k p^{(k)}) = \frac{1}{2} \alpha_k^2 (p^{(k)}, Ap^{(k)}) - \alpha_k (p^{(k)}, b - Ax^{(k)}) + f(x^{(k)})$$

$$\frac{\partial f(x^{(k)} + \alpha_k p^{(k)})}{\partial \alpha_k} = 0 \Rightarrow \alpha_k = \frac{(p^{(k)}, b - Ax^{(k)})}{(p^{(k)}, Ap^{(k)})} = \frac{(p^{(k)}, r^{(k)})}{(p^{(k)}, Ap^{(k)})} \quad (1)$$

$$r^{(k)} = b - Ax^{(k)} \text{ residual vector}$$

Derivation of CG Algorithm (3/5)

Residual vector at $(k+1)$ -th iteration:

$$r^{(k+1)} = r^{(k)} - \alpha_k A p^{(k)} \quad \underline{(2)}$$

$$r^{(k+1)} = b - Ax^{(k+1)}, r^{(k)} = b - Ax^{(k)}$$

$$r^{(k+1)} - r^{(k)} = -Ax^{(k+1)} + Ax^{(k)} = -\alpha_k A p^{(k)}$$

Search direction vector p is defined by the following recurrence formula:

$$p^{(k+1)} = r^{(k+1)} + \beta_k p^{(k)}, p^{(0)} = r^{(0)} \quad \underline{(3)}$$

It's lucky if we can get exact solution y at $(k+1)$ -th iteration:

$$y = x^{(k+1)} + \alpha_{k+1} p^{(k+1)}$$

Derivation of CG Algorithm (4/5)

BTW, we have the following (convenient) orthogonality relation:

$$(Ap^{(k)}, y - x^{(k+1)}) = 0$$

$$\begin{aligned} (Ap^{(k)}, y - x^{(k+1)}) &= (p^{(k)}, Ay - Ax^{(k+1)}) = (p^{(k)}, b - Ax^{(k+1)}) \\ &= (p^{(k)}, b - A[x^{(k)} + \alpha_k p^{(k)}]) = (p^{(k)}, b - Ax^{(k)} - \alpha_k Ap^{(k)}) \\ &= (p^{(k)}, r^{(k)} - \alpha_k Ap^{(k)}) = (p^{(k)}, r^{(k)}) - \alpha_k (p^{(k)}, Ap^{(k)}) = 0 \end{aligned}$$

$$\therefore \alpha_k = \frac{(p^{(k)}, r^{(k)})}{(p^{(k)}, Ap^{(k)})}$$

Thus, following relation is obtained:

$$(Ap^{(k)}, y - x^{(k+1)}) = (Ap^{(k)}, \alpha_{k+1} p^{(k+1)}) = 0 \Rightarrow (p^{(k+1)}, Ap^{(k)}) = 0$$

Derivation of CG Algorithm (5/5)

$$\begin{aligned} (p^{(k+1)}, Ap^{(k)}) &= (r^{(k+1)} + \beta_k p^{(k)}, Ap^{(k)}) = (r^{(k+1)}, Ap^{(k)}) + \beta_k (p^{(k)}, Ap^{(k)}) = 0 \\ \Rightarrow \beta_k &= \frac{-(r^{(k+1)}, Ap^{(k)})}{(p^{(k)}, Ap^{(k)})} \quad \textcolor{red}{(4)} \end{aligned}$$

$(p^{(k+1)}, Ap^{(k)}) = 0$ $p^{(k)}$ & $p^{(k+1)}$ are “conjugate (共役)” for matrix A

$p^{(k)}$: search direction vector, “gradient” vector

```

Compute  $p^{(0)} = r^{(0)} = b - [A]x^{(0)}$ 
for i= 1, 2, ...
    calc.  $\alpha_{i-1}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_{i-1}p^{(i-1)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_{i-1}[A]p^{(i-1)}$ 

    check convergence  $|r|$ 
    (if not converged)
    calc.  $\beta_{i-1}$ 
     $p^{(i)} = r^{(i)} + \beta_{i-1}p^{(i-1)}$ 
end
  
```

$$\alpha_{i-1} = \frac{(p^{(i-1)}, r^{(i-1)})}{(p^{(i-1)}, Ap^{(i-1)})}$$

$$\beta_{i-1} = \frac{-(r^{(i)}, Ap^{(i-1)})}{(p^{(i-1)}, Ap^{(i-1)})}$$

Properties of CG Algorithm

Following “conjugate(共役)” relationship is obtained for arbitrary (i,j) :

$$\left(p^{(i)}, Ap^{(j)}\right) = 0 \quad (i \neq j)$$

Following relationships are also obtained for $p^{(k)}$ and $r^{(k)}$:

$$\left(r^{(i)}, r^{(j)}\right) = 0 \quad (i \neq j), \quad \left(p^{(k)}, r^{(k)}\right) = \left(r^{(k)}, r^{(k)}\right)$$

In N-dimensional space, only N sets of orthogonal and linearly independent residual vector $r^{(k)}$. This means CG method converges after N iterations if number of unknowns is N. Actually, round-off error sometimes affects convergence.

Proof (1/3)

Mathematical Induction

数学的帰納法

$$\begin{aligned} (r^{(i)}, r^{(j)}) &= 0 \quad (i \neq j) \\ (p^{(i)}, Ap^{(j)}) &= 0 \quad (i \neq j) \end{aligned}$$

$$(1) \quad \alpha_k = \frac{(p^{(k)}, r^{(k)})}{(p^{(k)}, Ap^{(k)})}$$

$$(2) \quad r^{(k+1)} = r^{(k)} - \alpha_k Ap^{(k)}$$

$$(3) \quad p^{(k+1)} = r^{(k+1)} + \beta_k p^{(k)}, \quad r^{(0)} = p^{(0)}$$

$$(4) \quad \beta_k = \frac{-(r^{(k+1)}, Ap^{(k)})}{(p^{(k)}, Ap^{(k)})}$$

Proof (2/3)

Mathematical Induction

数学的帰納法

$$\begin{aligned} (r^{(i)}, r^{(j)}) &= 0 \quad (i \neq j) \\ (p^{(i)}, Ap^{(j)}) &= 0 \quad (i \neq j) \end{aligned} \quad (*)$$

(*) is satisfied for $i \leq k, j \leq k$ where $i \neq j$

$$\begin{aligned} \text{if } i < k \quad (r^{(k+1)}, r^{(i)}) &= (r^{(i)}, r^{(k+1)}) \stackrel{(2)}{=} (r^{(i)}, r^{(k)} - \alpha_k Ap^{(k)}) \\ &\stackrel{(*)}{=} -\alpha_k (r^{(i)}, Ap^{(k)}) \stackrel{(3)}{=} -\alpha_k (p^{(i)} - \beta_{i-1} p^{(i-1)}, Ap^{(k)}) \\ &= -\alpha_k (p^{(i)}, Ap^{(k)}) + \alpha_k \beta_{i-1} (p^{(i-1)}, Ap^{(k)}) \stackrel{(*)}{=} 0 \end{aligned}$$

$$\begin{aligned} \text{if } i = k \quad (r^{(k+1)}, r^{(k)}) &\stackrel{(2)}{=} (r^{(k)}, r^{(k)}) - (r^{(k)}, \alpha_k Ap^{(k)}) \\ &\stackrel{(3)}{=} (r^{(k)}, r^{(k)}) - (p^{(k)} - \beta_{k-1} p^{(k-1)}, \alpha_k Ap^{(k)}) \\ (1) \quad \alpha_k &= \frac{(p^{(k)}, r^{(k)})}{(p^{(k)}, Ap^{(k)})} \quad \stackrel{(*)}{=} (r^{(k)}, r^{(k)}) - \alpha_k (p^{(k)}, Ap^{(k)}) \stackrel{(1)}{=} (r^{(k)}, r^{(k)}) - (p^{(k)}, r^{(k)}) \\ (2) \quad r^{(k+1)} &= r^{(k)} - \alpha_k Ap^{(k)} \quad \stackrel{(3)}{=} (r^{(k)}, r^{(k)}) - (\beta_{k-1} p^{(k-1)} + r^{(k)}, r^{(k)}) \\ (3) \quad p^{(k+1)} &= r^{(k+1)} + \beta_k p^{(k)} \quad = -\beta_{k-1} (p^{(k-1)}, r^{(k)}) \stackrel{(2)}{=} -\beta_{k-1} (p^{(k-1)}, r^{(k-1)} - \alpha_{k-1} Ap^{(k-1)}) \\ (4) \quad \beta_k &= \frac{-(r^{(k+1)}, Ap^{(k)})}{(p^{(k)}, Ap^{(k)})} \quad = -\beta_{k-1} \{ (p^{(k-1)}, r^{(k-1)}) - \alpha_{k-1} (p^{(k-1)}, Ap^{(k-1)}) \} \stackrel{(1)}{=} 0 \end{aligned}$$

Proof (3/3)

Mathematical Induction 数学的帰納法

$$\begin{aligned} (r^{(i)}, r^{(j)}) &= 0 \quad (i \neq j) \\ (p^{(i)}, Ap^{(j)}) &= 0 \quad (i \neq j) \end{aligned} \quad (*)$$

$(*)$ is satisfied for $i \leq k, j \leq k$ where $i \neq j$

$$\begin{aligned} \text{if } i < k \quad (p^{(k+1)}, Ap^{(i)}) &\stackrel{(3)}{=} (r^{(k+1)} + \beta_k p^{(k)}, Ap^{(i)}) \\ &\stackrel{(*)}{=} (r^{(k+1)}, Ap^{(i)}) \\ &\stackrel{(2)}{=} \frac{1}{\alpha_i} (r^{(k+1)}, r^{(i)} - r^{(i+1)}) = 0 \end{aligned}$$

$$\begin{aligned} \text{if } i = k \quad (p^{(k+1)}, Ap^{(k)}) &\stackrel{(3)}{=} (r^{(k+1)}, Ap^{(k)}) + \beta_k (p^{(k)}, Ap^{(k)}) \\ &\stackrel{(4)}{=} 0 \end{aligned}$$

$$(1) \alpha_k = \frac{(p^{(k)}, r^{(k)})}{(p^{(k)}, Ap^{(k)})}$$

$$(2) r^{(k+1)} = r^{(k)} - \alpha_k Ap^{(k)}$$

$$(3) p^{(k+1)} = r^{(k+1)} + \beta_k p^{(k)}$$

$$(4) \beta_k = \frac{-(r^{(k+1)}, Ap^{(k)})}{(p^{(k)}, Ap^{(k)})}$$

$$\left(r^{(k+1)}, r^{(k)}\right) = 0$$

$$\left(r^{(k+1)}, r^{(k)}\right) \stackrel{(2)}{=} \left(r^{(k)}, r^{(k)}\right) - \left(r^{(k)}, \alpha_k A p^{(k)}\right)$$

$$\stackrel{(3)}{=} \left(r^{(k)}, r^{(k)}\right) - \left(p^{(k)} - \beta_{k-1} p^{(k-1)}, \alpha_k A p^{(k)}\right)$$

$$\stackrel{(*)}{=} \left(r^{(k)}, r^{(k)}\right) - \alpha_k \left(p^{(k)}, A p^{(k)}\right) \stackrel{(1)}{=} \left(r^{(k)}, r^{(k)}\right) - \left(p^{(k)}, r^{(k)}\right) = 0$$

$$\therefore \left(r^{(k)}, r^{(k)}\right) = \left(p^{(k)}, r^{(k)}\right)$$

$$(1) \alpha_k = \frac{\left(p^{(k)}, r^{(k)}\right)}{\left(p^{(k)}, A p^{(k)}\right)}$$

$$(2) r^{(k+1)} = r^{(k)} - \alpha_k A p^{(k)}$$

$$(3) p^{(k+1)} = r^{(k+1)} + \beta_k p^{(k)}$$

$$(4) \beta_k = \frac{-\left(r^{(k+1)}, A p^{(k)}\right)}{\left(p^{(k)}, A p^{(k)}\right)}$$

$$\alpha_k, \beta_k$$

Usually, we use simpler definitions of α_k, β_k as follows:

$$\alpha_k = \frac{(p^{(k)}, b - Ax^{(k)})}{(p^{(k)}, Ap^{(k)})} = \frac{(p^{(k)}, r^{(k)})}{(p^{(k)}, Ap^{(k)})} = \frac{(r^{(k)}, r^{(k)})}{(p^{(k)}, Ap^{(k)})}$$

$$\because (p^{(k)}, r^{(k)}) = (r^{(k)}, r^{(k)})$$

$$\beta_k = \frac{-(r^{(k+1)}, Ap^{(k)})}{(p^{(k)}, Ap^{(k)})} = \frac{(r^{(k+1)}, r^{(k+1)})}{(r^{(k)}, r^{(k)})}$$

$$\because (r^{(k+1)}, Ap^{(k)}) = \frac{(r^{(k+1)}, r^{(k)} - r^{(k+1)})}{\alpha_k} = -\frac{(r^{(k+1)}, r^{(k+1)})}{\alpha_k}$$

Procedures of Conjugate Gradient

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
     $z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} \cdot z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} \cdot q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

$\mathbf{x}^{(i)}$: Vector

α_i : Scalar

$$\beta_{i-1} = \frac{\left(r^{(i-1)}, r^{(i-1)} \right)}{\left(r^{(i-2)}, r^{(i-2)} \right)} \quad \left(= \rho_{i-1} \right)$$

$$\alpha_i = \frac{\left(r^{(i-1)}, r^{(i-1)} \right)}{\left(p^{(i)}, Ap^{(i)} \right)} \quad \left(= \rho_{i-1} \right)$$

Preconditioning for Iterative Solvers

- Convergence rate of iterative solvers strongly depends on the spectral properties (eigenvalue distribution) of the coefficient matrix A .
 - Eigenvalue distribution is small, eigenvalues are close to 1
 - In “ill-conditioned” problems, “condition number” (ratio of max/min eigenvalue if A is symmetric) is large (条件数) .
- A preconditioner M (whose properties are similar to those of A) transforms the linear system into one with more favorable spectral properties (前处理)
 - M transforms $Ax=b$ into $A'x=b'$ where $A'=M^{-1}A$, $b'=M^{-1}b$
 - If $M \sim A$, $M^{-1}A$ is close to identity matrix.
 - If $M^{-1}=A^{-1}$, this is the best preconditioner (Gaussian Elim.)
 - Generally, $A'x'=b'$ where $A'=M_L^{-1}AM_R^{-1}$, $b'=M_L^{-1}b$, $x'=M_Rx$
 - M_L/M_R : Left/Right Preconditioning (左／右前处理)

Preconditioned CG Solver

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
    solve  $[M]z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} \cdot z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} \cdot q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

$$[M] = [M_1] [M_2]$$

$$[A'] x' = b'$$

$$[A'] = [M_1]^{-1} [A] [M_2]^{-1}$$

$$x' = [M_2] x, \quad b' = [M_1]^{-1} b$$

$$p' \Rightarrow [M_2] p, \quad r' \Rightarrow [M_1]^{-1} r$$

$$p'^{(i)} = r'^{(i-1)} + \beta'_{i-1} p'^{(i-1)}$$

$$[M_2] p^{(i)} = [M_1]^{-1} r^{(i-1)} + \beta'_{i-1} [M_2] p^{(i-1)}$$

$$p^{(i)} = [M_2]^{-1} [M_1]^{-1} r^{(i-1)} + \beta'_{i-1} p^{(i-1)}$$

$$p^{(i)} = [M]^{-1} r^{(i-1)} + \beta'_{i-1} p^{(i-1)}$$

$$\beta'_{i-1} = \frac{([M]^{-1} r^{(i-1)}, r^{(i-1)})}{([M]^{-1} r^{(i-2)}, r^{(i-2)})}$$

$$\alpha'_{i-1} = \frac{([M]^{-1} r^{(i-1)}, r^{(i-1)})}{(p^{(i-1)}, [A] p^{(i-1)})}$$

In CG method, preconditioner usually satisfies $[M_2] = [M_1]^T$, such as Incomplete Cholesky/Incomplete Modified Cholesky Factorizations. In this problem, let us define $[M_1]$ and $[M_2]$ as follows:

$$[M_1] = [X]^T, [M_2] = [X], [M] = [M_1][M_2]$$

$$[A']x' = b'$$

$$[A'] = [M_1]^{-1}[A][M_2]^{-1} = [[X]^T]^{-1}[A][X]^{-1} = [X]^{-T}[A][X]^{-1}$$

$$x' = [X]x, \quad b' = [X]^{-T}b, \quad r' = [X]^{-T}r$$

$$\begin{aligned} \alpha'_{i-1} &= \frac{(r'^{(i-1)}, r'^{(i-1)})}{(p'^{(i-1)}, A' p'^{(i-1)})} = \frac{([X]^{-T} r^{(i-1)}, [X]^{-T} r^{(i-1)})}{([X] p^{(i-1)}, [X]^{-T} [A][X]^{-1} [X] p^{(i-1)})} \\ &= \frac{\left(([X]^{-T} r^{(i-1)})^T, [X]^{-T} r^{(i-1)} \right)}{\left(([X] p^{(i-1)})^T, [X]^{-T} [A] p^{(i-1)} \right)} = \frac{\left((r^{(i-1)})^T [X]^{-1}, [X^T]^{-1} r^{(i-1)} \right)}{\left((p^{(i-1)})^T [X]^T, [X]^{-T} [A] p^{(i-1)} \right)} \\ &= \frac{\left(r^{(i-1)}, [[X^T][X]]^{-1} r^{(i-1)} \right)}{(p^{(i-1)}, [A] p^{(i-1)})} = \frac{(r^{(i-1)}, [M]^{-1} r^{(i-1)})}{(p^{(i-1)}, [A] p^{(i-1)})} = \frac{(r^{(i-1)}, z^{(i-1)})}{(p^{(i-1)}, [A] p^{(i-1)})} \end{aligned}$$

$$\begin{aligned}
\beta'_{i-1} &= \frac{\left(r^{(i-1)}, r^{(i-1)} \right)}{\left(r^{(i-2)}, r^{(i-2)} \right)} = \frac{\left([\mathbf{X}]^{-T} r^{(i-1)}, [\mathbf{X}]^{-T} r^{(i-1)} \right)}{\left([\mathbf{X}]^{-T} r^{(i-2)}, [\mathbf{X}]^{-T} r^{(i-2)} \right)} \\
&= \frac{\left(\left([\mathbf{X}]^{-T} r^{(i-1)} \right)^T, [\mathbf{X}]^{-T} r^{(i-1)} \right)}{\left(\left([\mathbf{X}]^{-T} r^{(i-2)} \right)^T, [\mathbf{X}]^{-T} r^{(i-2)} \right)} = \frac{\left(\left(r^{(i-1)} \right)^T [\mathbf{X}]^{-1}, [\mathbf{X}^T]^{-1} r^{(i-1)} \right)}{\left(\left(r^{(i-2)} \right)^T [\mathbf{X}]^{-1}, [\mathbf{X}^T]^{-1} r^{(i-2)} \right)} \\
&= \frac{\left(r^{(i-1)}, \left[[\mathbf{X}^T] [\mathbf{X}] \right]^{-1} r^{(i-1)} \right)}{\left(r^{(i-2)}, \left[[\mathbf{X}^T] [\mathbf{X}] \right]^{-1} r^{(i-2)} \right)} = \frac{\left(r^{(i-1)}, [\mathbf{M}]^{-1} r^{(i-1)} \right)}{\left(r^{(i-2)}, [\mathbf{M}]^{-1} r^{(i-2)} \right)} = \frac{\left(r^{(i-1)}, z^{(i-1)} \right)}{\left(r^{(i-2)}, z^{(i-2)} \right)}
\end{aligned}$$

Preconditioned Conjugate Gradient Method (PCG)

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
    solve  $[M]z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} \cdot z^{(i-1)}$ 
    if  $i = 1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} \cdot q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

Solving the following equation:

$$\{z\} = [M]^{-1} \{r\}$$

“Approximate Inverse Matrix”

$$[M]^{-1} \approx [A]^{-1}, \quad [M] \approx [A]$$

Ultimate Preconditioning:

Inverse Matrix

$$[M]^{-1} = [A]^{-1}, \quad [M] = [A]$$

Diagonal Scaling: Simple but weak

$$[M]^{-1} = [D]^{-1}, \quad [M] = [D]$$

Diagonal Scaling, Point-Jacobi

$$[M] = \begin{bmatrix} D_1 & 0 & \dots & 0 & 0 \\ 0 & D_2 & & 0 & 0 \\ \dots & & \dots & & \dots \\ 0 & 0 & & D_{N-1} & 0 \\ 0 & 0 & \dots & 0 & D_N \end{bmatrix}$$

- **solve** $[M] \mathbf{z}^{(i-1)} = \mathbf{r}^{(i-1)}$ is very easy.
- Provides fast convergence for simple problems.

ILU(0), IC(0)

- Widely used Preconditioners for Sparse Matrices
 - Incomplete LU Factorization
 - Incomplete Cholesky Factorization (for Symmetric Matrices)
- Incomplete Direct Method
 - Even if original matrix is sparse, inverse matrix is not necessarily sparse.
 - fill-in
 - ILU(0)/IC(0) without fill-in have same non-zero pattern with the original (sparse) matrices
- **More details are shown in the later stage of this class**

- Background
 - Finite Volume Method
 - Preconditioned Iterative Solvers
- **PCG Solver for Poisson's Equations**
 - **How to run**
 - **Data Structure**
 - Program
 - Initialization
 - Coefficient Matrices
 - PCG

Target Application

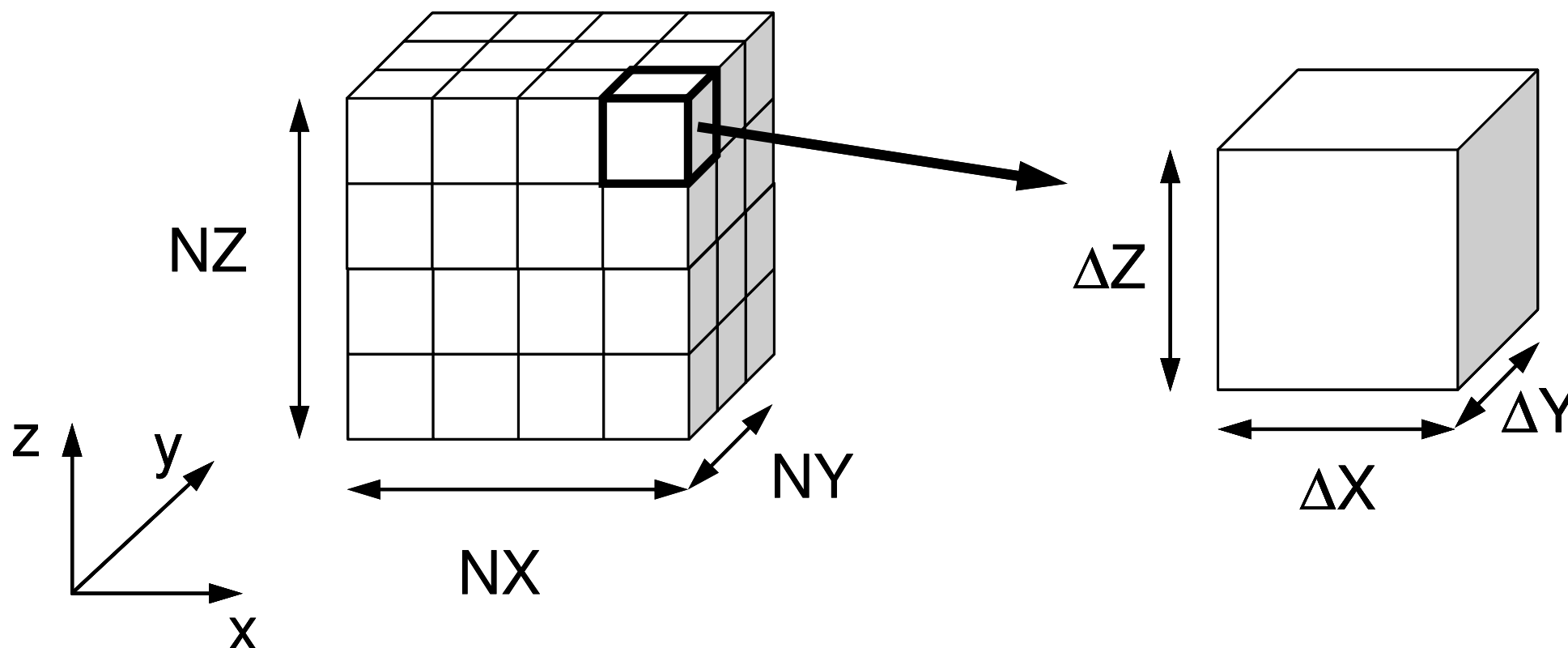
- 3D Poisson Equation/Poisson's Equation

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

- Finite Volume Method (FVM)
 - Arbitrary Shape Meshes, Cell-Centered
 - “Direct” Finite Difference Method
- Boundary Conditions (B.C.) etc.
 - Dirichlet B.C., Volume Flux
- Preconditioned Iterative Solvers
 - Conjugate Gradient + Preconditioner

3D Structured Mesh

Internal data structure is “unstructured”



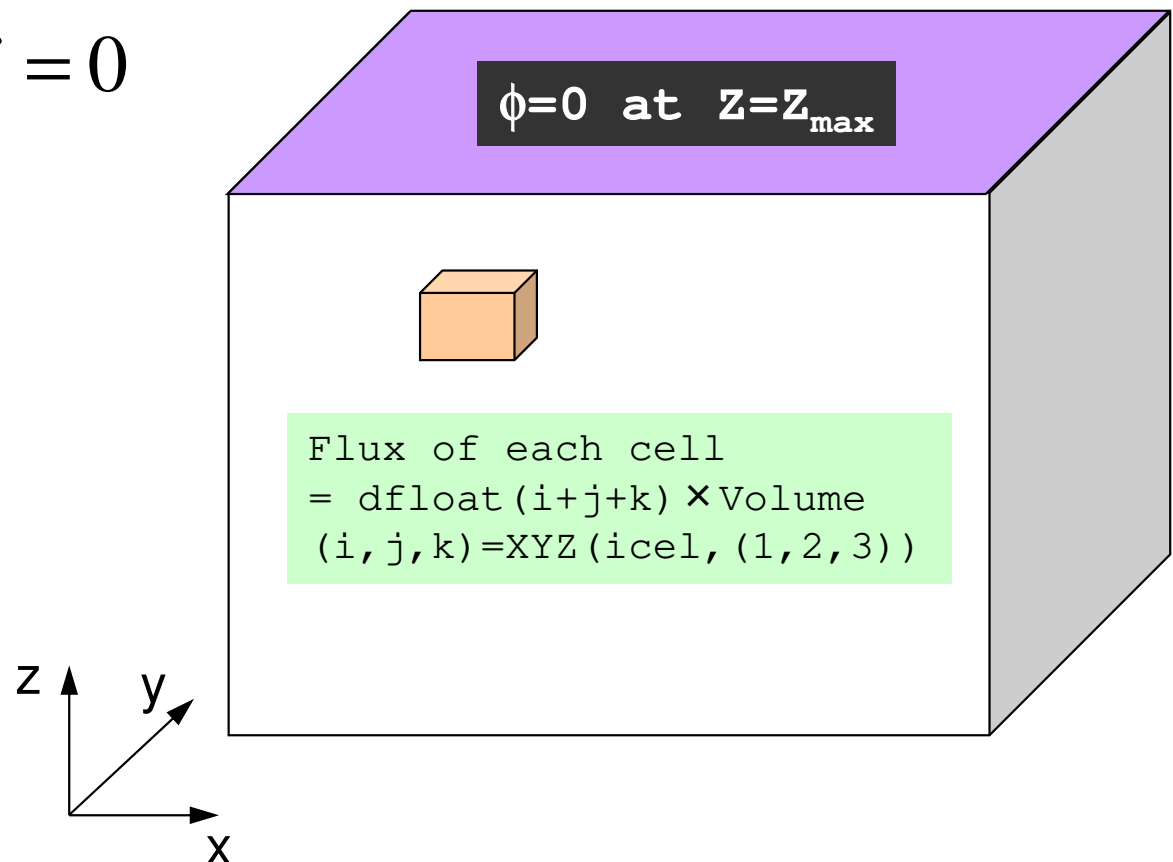
Target Problem: Variables are defined at cell-center's

Poisson Equation/ Poisson's Equation

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

Boundary Conditions (B.C.) etc.

- Volume Flux
- $\phi=0 @ Z=Z_{\max}$

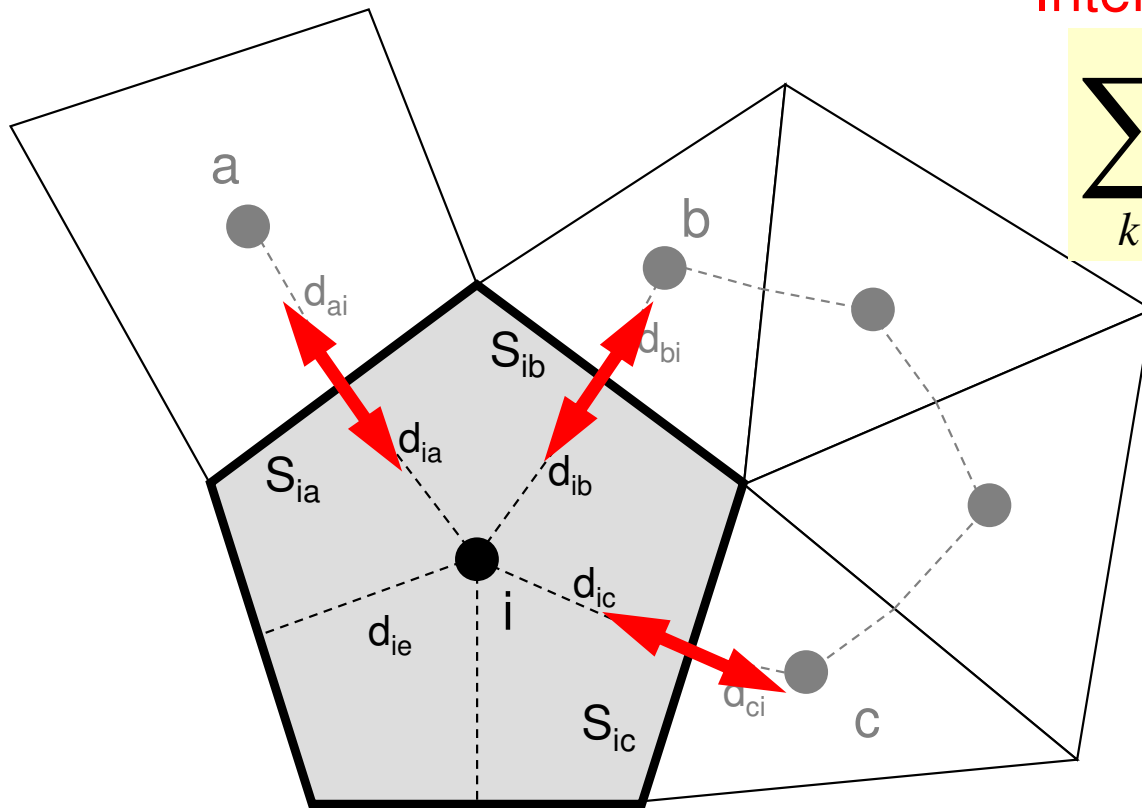


Poisson Equation by Finite Volume Method (FVM)

Conservation of Fluxes through
Surfaces

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

Diffusion:
Interaction with Neighbors



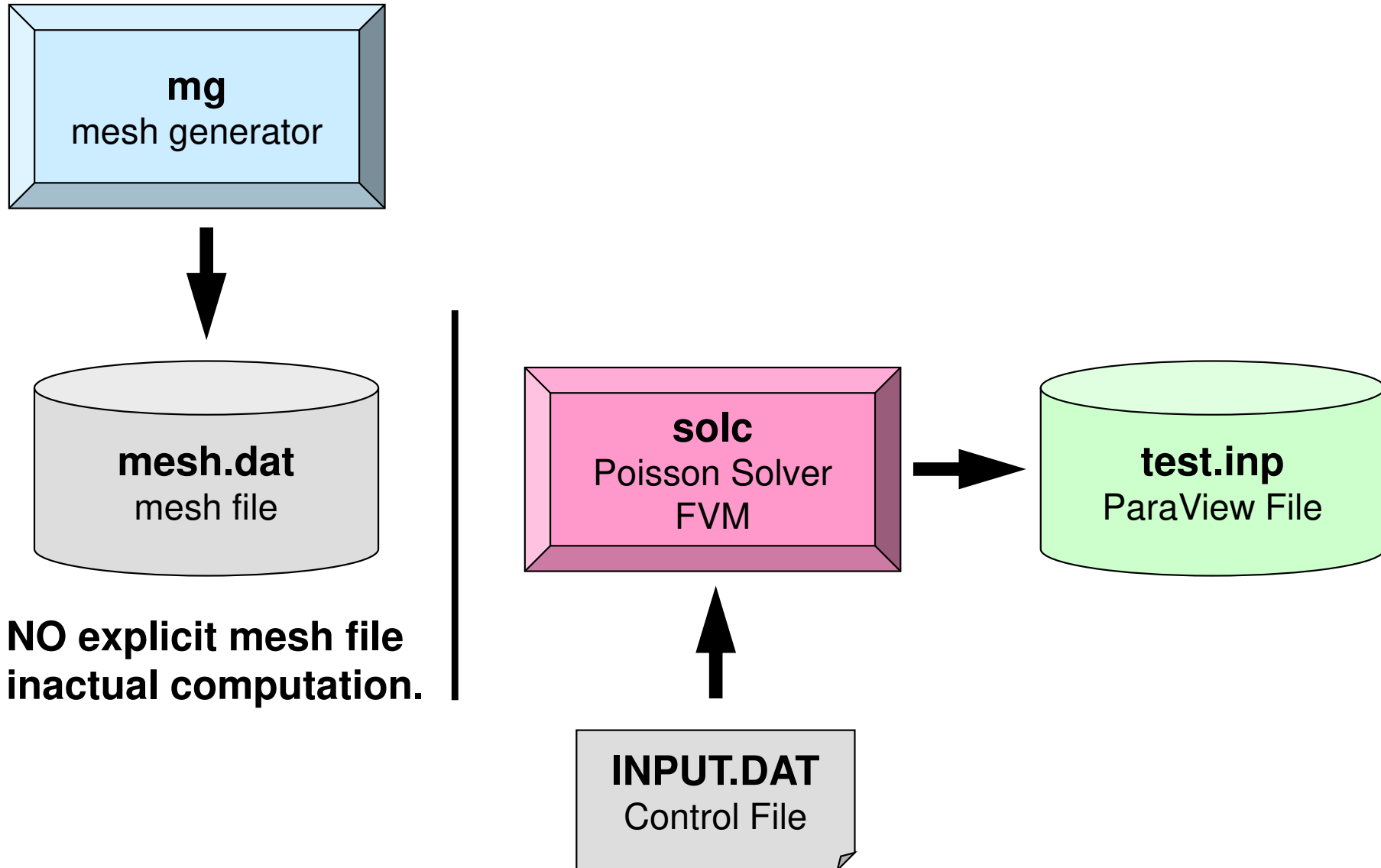
$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

Volume Flux

- V_i : Volume
- S : Surface Area
- d_{ij} : Distance between
Cell-Center &
Surface
- Q : Volume Flux

Running the Program:

<\$P-FVM>/run



Running the Program

Compiling

```
$> cd <$P-FVM>/run
```

```
$> cc -O mg.c -o mg
```

```
$> ls mg
```

mg

Mesh Generator: **mg**

```
$> cd ../src-c
```

```
$> make
```

```
$> ls ../run/solc
```

solc

Poisson Solver (FVM): **solc**

Running the Program

Mesh Generation

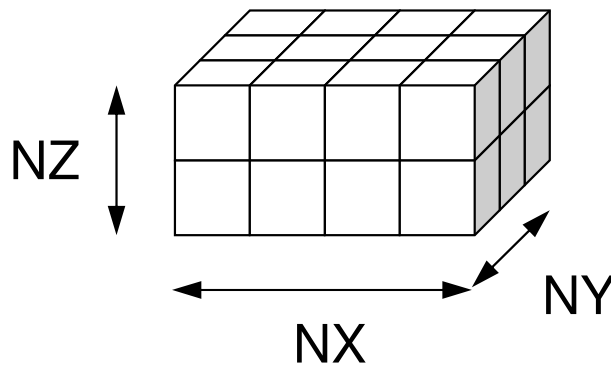
```
$> cd ../run
```

```
$> ./mg
```

```
4 3 2
```

NX, NY, NZ

```
$> ls mesh.dat  
mesh.dat
```



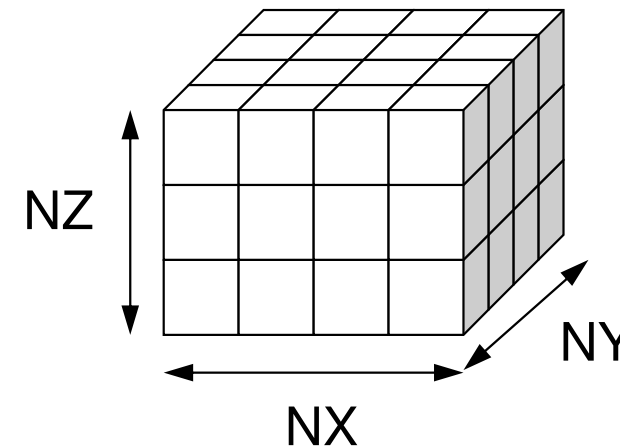
mesh.dat (1/5)

4	3	2							
24									
1	0	2	0	5	0	13	1	1	1
2	1	3	0	6	0	14	2	1	1
3	2	4	0	7	0	15	3	1	1
4	3	0	0	8	0	16	4	1	1
5	0	6	1	9	0	17	1	2	1
6	5	7	2	10	0	18	2	2	1
7	6	8	3	11	0	19	3	2	1
8	7	0	4	12	0	20	4	2	1
9	0	10	5	0	0	21	1	3	1
10	9	11	6	0	0	22	2	3	1
11	10	12	7	0	0	23	3	3	1
12	11	0	8	0	0	24	4	3	1
13	0	14	0	17	1	0	1	1	2
14	13	15	0	18	2	0	2	1	2
15	14	16	0	19	3	0	3	1	2
16	15	0	0	20	4	0	4	1	2
17	0	18	13	21	5	0	1	2	2
18	17	19	14	22	6	0	2	2	2
19	18	20	15	23	7	0	3	2	2
20	19	0	16	24	8	0	4	2	2
21	0	22	17	0	9	0	1	3	2
22	21	23	18	0	10	0	2	3	2
23	22	24	19	0	11	0	3	3	2
24	23	0	20	0	12	0	4	3	2

```
read (21,'(10i10)') NX , NY , NZ
read (21,'(10i10)') ICELTOT
```

```
do i= 1, ICELTOT
  read (21, '(10i10)' ) ii, (NEIBcell(i,k), k= 1, 6), (XYZ(i,j), j= 1, 3)
enddo
```

mesh.dat (2/5)



**Number of meshes
in X/Y/Z directions**

4	3	2							
24									
1	0	2	0	5	0	13	1	1	1
2	1	3	0	6	0	14	2	1	1
3	2	4	0	7	0	15	3	1	1
4	3	0	0	8	0	16	4	1	1
5	0	6	1	9	0	17	1	2	1
6	5	7	2	10	0	18	2	2	1
7	6	8	3	11	0	19	3	2	1
8	7	0	4	12	0	20	4	2	1
9	0	10	5	0	0	21	1	3	1
10	9	11	6	0	0	22	2	3	1
11	10	12	7	0	0	23	3	3	1
12	11	0	8	0	0	24	4	3	1
13	0	14	0	17	1	0	1	1	2
14	13	15	0	18	2	0	2	1	2
15	14	16	0	19	3	0	3	1	2
16	15	0	0	20	4	0	4	1	2
17	0	18	13	21	5	0	1	2	2
18	17	19	14	22	6	0	2	2	2
19	18	20	15	23	7	0	3	2	2
20	19	0	16	24	8	0	4	2	2
21	0	22	17	0	9	0	1	3	2
22	21	23	18	0	10	0	2	3	2
23	22	24	19	0	11	0	3	3	2
24	23	0	20	0	12	0	4	3	2

```
read (21,'(10i10)') NX , NY , NZ
read (21,'(10i10)') ICELTOT
```

```
do i= 1, ICELTOT
  read (21, '(10i10)') ii, (NEIBcell(i,k), k= 1, 6), (XYZ(i,j), j= 1, 3)
enddo
```

mesh.dat (3/5)

Number of Meshes (Cells)
= NX x NY x NZ

4	3	2							
24	1	0	2	0	5	0	13	1	1
	2	1	3	0	6	0	14	2	1
	3	2	4	0	7	0	15	3	1
	4	3	0	0	8	0	16	4	1
	5	0	6	1	9	0	17	1	2
	6	5	7	2	10	0	18	2	2
	7	6	8	3	11	0	19	3	2
	8	7	0	4	12	0	20	4	2
	9	0	10	5	0	0	21	1	3
	10	9	11	6	0	0	22	2	3
	11	10	12	7	0	0	23	3	3
	12	11	0	8	0	0	24	4	3
	13	0	14	0	17	1	0	1	1
	14	13	15	0	18	2	0	2	1
	15	14	16	0	19	3	0	3	1
	16	15	0	0	20	4	0	4	1
	17	0	18	13	21	5	0	1	2
	18	17	19	14	22	6	0	2	2
	19	18	20	15	23	7	0	3	2
	20	19	0	16	24	8	0	4	2
	21	0	22	17	0	9	0	1	3
	22	21	23	18	0	10	0	2	3
	23	22	24	19	0	11	0	3	3
	24	23	0	20	0	12	0	4	3

```
read (21,'(10i10)') NX , NY , NZ
```

```
read (21,'(10i10)') ICELTOT
```

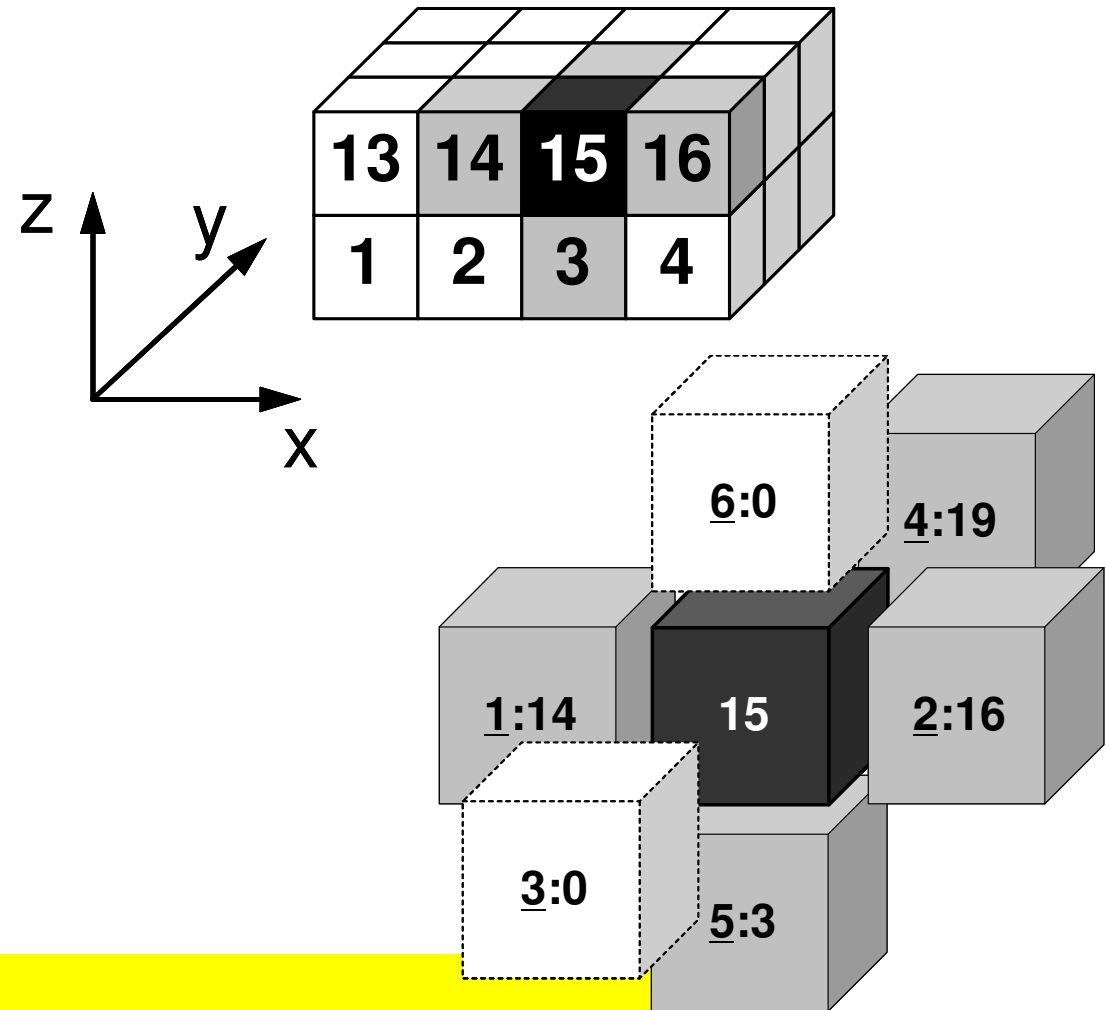
```
do i= 1, ICELTOT
```

```
  read (21, '(10i10)') ii, (NEIBcell(i,k), k= 1, 6), (XYZ(i,j), j= 1, 3)
```

```
enddo
```

mesh.dat (4/5)

Neighboring Cells: NEIBcell(i,k)



4	3	2							
24									
1	0	2	0	5	0	13	1	1	1
2	1	3	0	6	0	14	2	1	1
3	2	4	0	7	0	15	3	1	1
4	3	0	0	8	0	16	4	1	1
5	0	6	1	9	0	17	1	2	1
6	5	7	2	10	0	18	2	2	1
7	6	8	3	11	0	19	3	2	1
8	7	0	4	12	0	20	4	2	1
9	0	10	5	0	0	21	1	3	1
10	9	11	6	0	0	22	2	3	1
11	10	12	7	0	0	23	3	3	1
12	11	0	8	0	0	24	4	3	1
13	0	14	0	17	1	0	1	1	2
14	13	15	0	18	2	0	2	1	2
15	14	16	0	19	3	0	3	1	2
16	15	0	0	20	4	0	4	1	2
17	0	18	13	21	5	0	1	2	2
18	17	19	14	22	6	0	2	2	2
19	18	20	15	23	7	0	3	2	2
20	19	0	16	24	8	0	4	2	2
21	0	22	17	0	9	0	1	3	2
22	21	23	18	0	10	0	2	3	2
23	22	24	19	0	11	0	3	3	2
24	23	0	20	0	12	0	4	3	2

```
read (21,'(10i10)') NX , NY , NZ
read (21,'(10i10)') ICELTOT
```

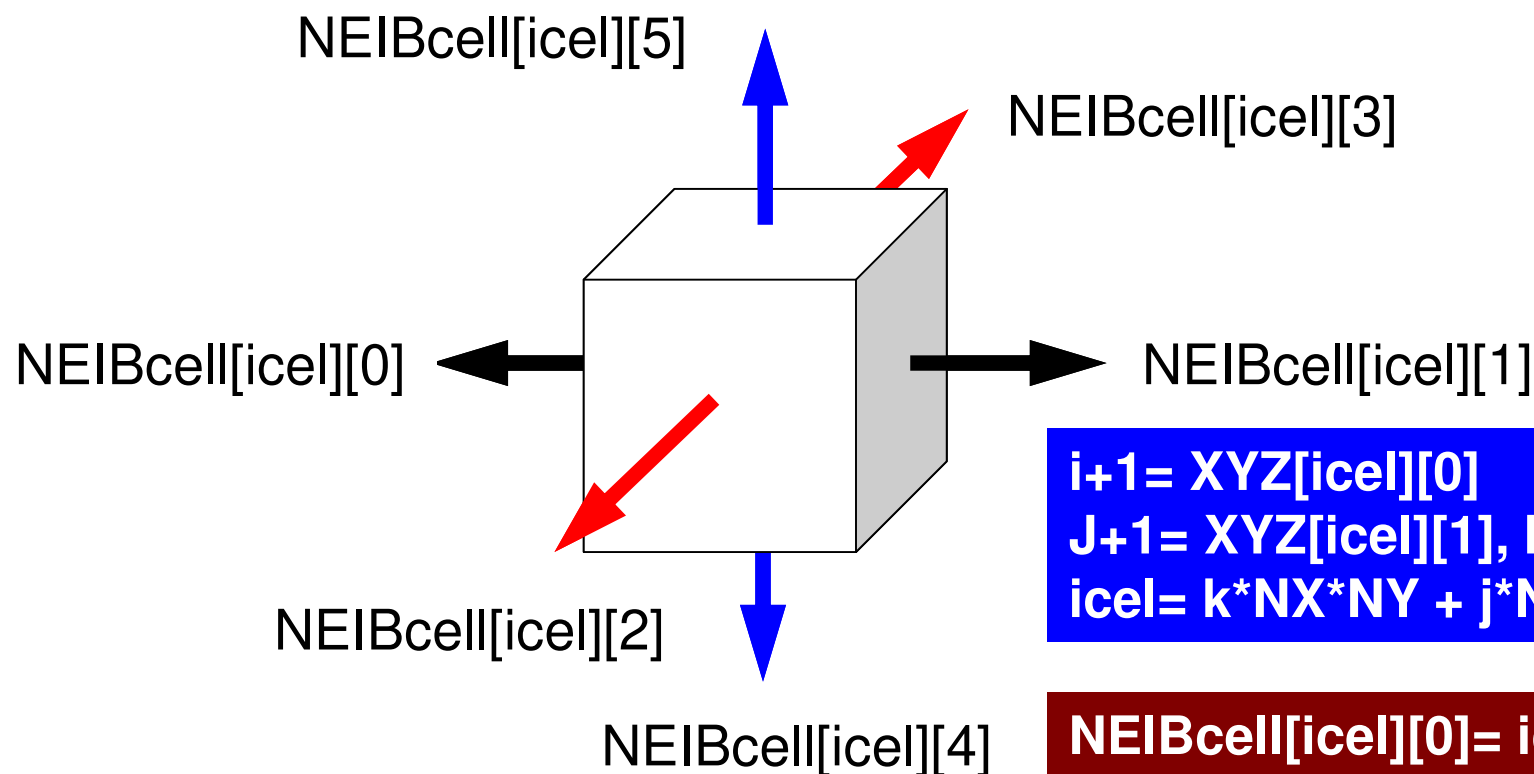
```
do i= 1, ICELTOT
```

```
  read (21, '(10i10)') ii, (NEIBcell(i,k), k= 1, 6), (XYZ(i,j), j= 1, 3)
enddo
```

1st Col.: Global ID of the Cell

NEIBcell: ID of Neighboring Mesh/Cell

=0: for Boundary Surface



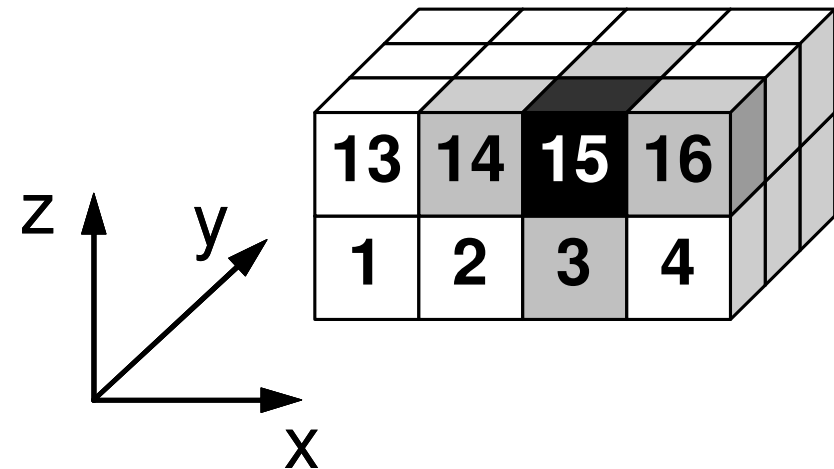
$i+1 = \text{XYZ}[\text{icel}][0]$
 $J+1 = \text{XYZ}[\text{icel}][1], k+1 = \text{XYZ}[\text{icel}][2]$
 $\text{icel} = k * \text{NX} * \text{NY} + j * \text{NX} + i$

$\text{NEIBcell}[\text{icel}][0] = \text{icel} - 1 \quad + 1$
 $\text{NEIBcell}[\text{icel}][1] = \text{icel} + 1 \quad + 1$
 $\text{NEIBcell}[\text{icel}][2] = \text{icel} - \text{NX} \quad + 1$
 $\text{NEIBcell}[\text{icel}][3] = \text{icel} + \text{NX} \quad + 1$
 $\text{NEIBcell}[\text{icel}][4] = \text{icel} - \text{NX} * \text{NY} + 1$
 $\text{NEIBcell}[\text{icel}][5] = \text{icel} + \text{NX} * \text{NY} + 1$

mesh.dat (5/5)

Location in X,Y,Z-directions: XYZ(i,j)

	4	3	2						
24	1	0	2	0	5	0	13	1	1
	2	1	3	0	6	0	14	2	1
	3	2	4	0	7	0	15	3	1
	4	3	0	0	8	0	16	4	1
	5	0	6	1	9	0	17	1	2
	6	5	7	2	10	0	18	2	2
	7	6	8	3	11	0	19	3	2
	8	7	0	4	12	0	20	4	2
	9	0	10	5	0	0	21	1	3
10	9	11	6	0	0	0	22	2	3
11	10	12	7	0	0	0	23	3	3
12	11	0	8	0	0	0	24	4	3
13	0	14	0	17	1	0		1	1
14	13	15	0	18	2	0		2	1
15	14	16	0	19	3	0		3	1
16	15	0	0	20	4	0		4	1
17	0	18	13	21	5	0		1	2
18	17	19	14	22	6	0		2	2
19	18	20	15	23	7	0		3	2
20	19	0	16	24	8	0		4	2
21	0	22	17	0	9	0		1	3
22	21	23	18	0	10	0		2	3
23	22	24	19	0	11	0		3	3
24	23	0	20	0	12	0		4	3

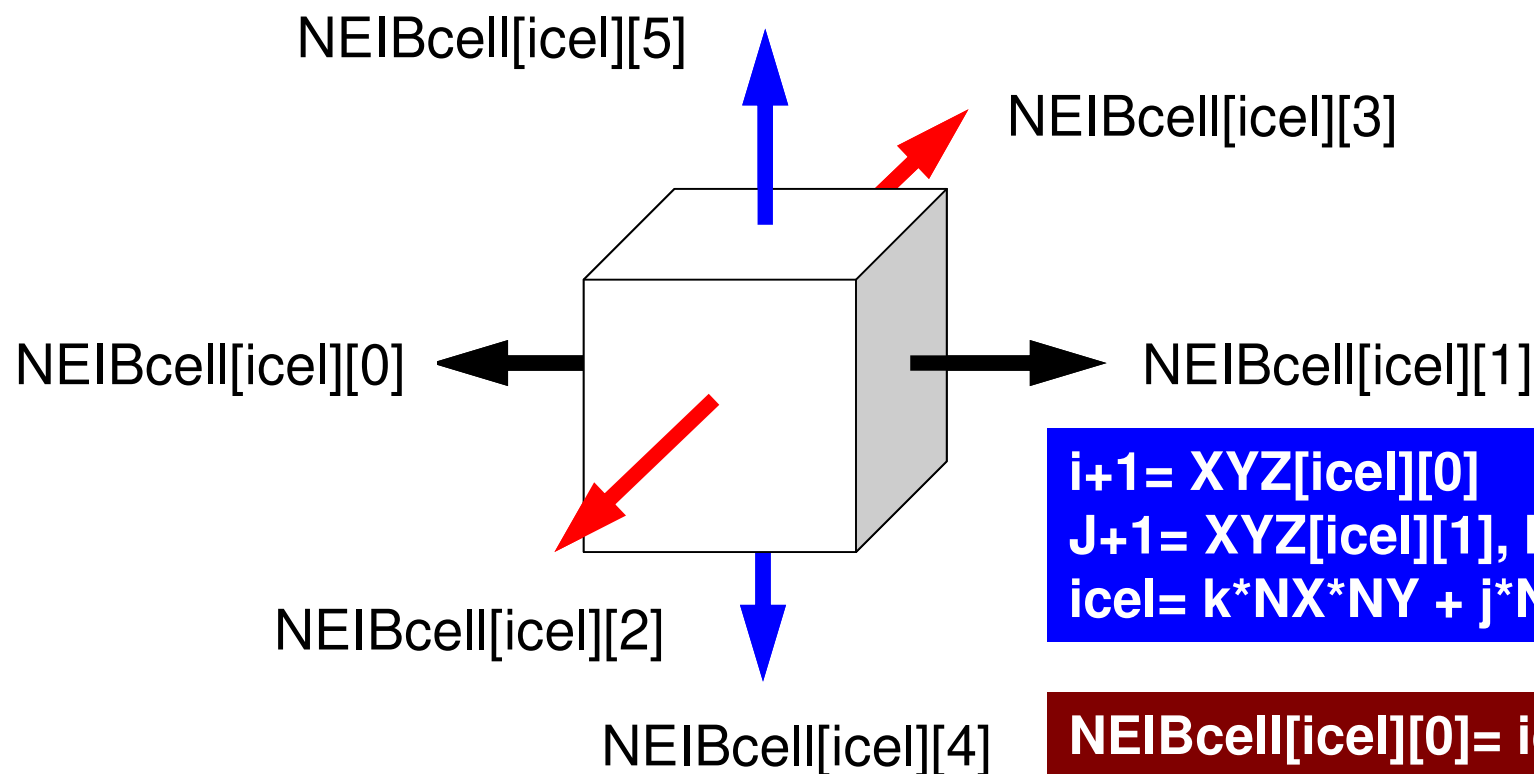


```
read (21,'(10i10)') NX , NY , NZ
read (21,'(10i10)') ICELTOT
```

```
do i= 1, ICELTOT
  read (21, '(10i10)') ii, (NEIBcell(i,k), k= 1, 6), (XYZ(i,j), j= 1, 3)
enddo
```

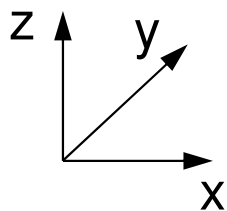
NEIBcell: ID of Neighboring Mesh/Cell

=0: for Boundary Surface



$i+1 = XYZ[icel][0]$
 $J+1 = XYZ[icel][1], k+1 = XYZ[icel][2]$
 $icel = k * NX * NY + j * NX + i$

$NEIBcell[icel][0] = icel - 1 \quad + 1$
 $NEIBcell[icel][1] = icel + 1 \quad + 1$
 $NEIBcell[icel][2] = icel - NX \quad + 1$
 $NEIBcell[icel][3] = icel + NX \quad + 1$
 $NEIBcell[icel][4] = icel - NX * NY + 1$
 $NEIBcell[icel][5] = icel + NX * NY + 1$

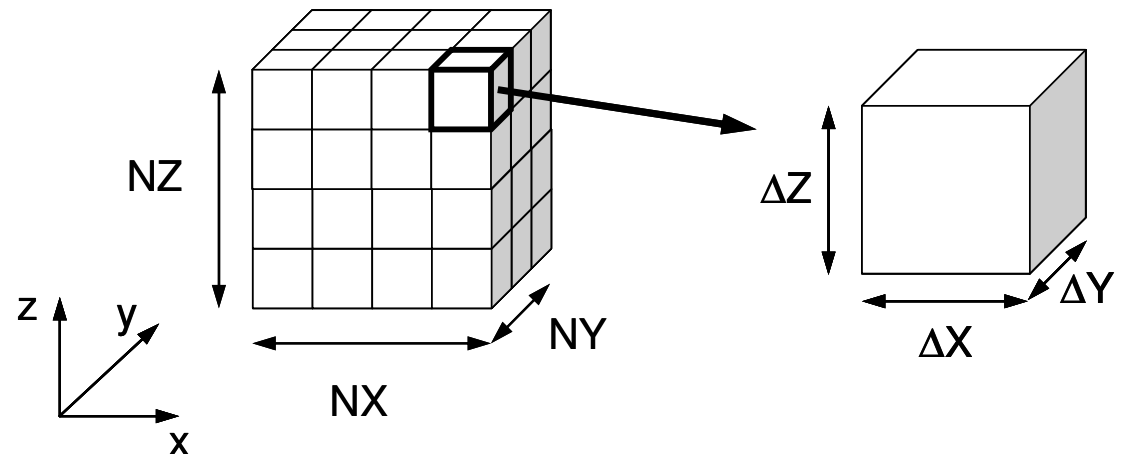


Running the Program

Control Data: <\$P-FVM>/run/INPUT.DAT

32	32	32	NX/NY/NZ
1.00e-00	1.00e-00	1.00e-00	DX/DY/DZ
1.0e-08			EPSICCG

- NX, NY, NZ
 - Number of meshes in X/Y/Z dir.
- DX, DY, DZ
 - Size of meshes
- EPSICCG
 - Convergence Criteria for PCG



Running the Program

```
$> cd <$P-FVM>/run
```

```
$> ./solc
```

```
    32    32    32      NX,NY,NZ  
    1  4.409359E+00  Residual at the 1st Iteration
```

```
   101  1.807571E-02
```

```
   201  2.194680E-08
```

```
   208  9.354536E-09  Residual at convergence (<10-8)
```

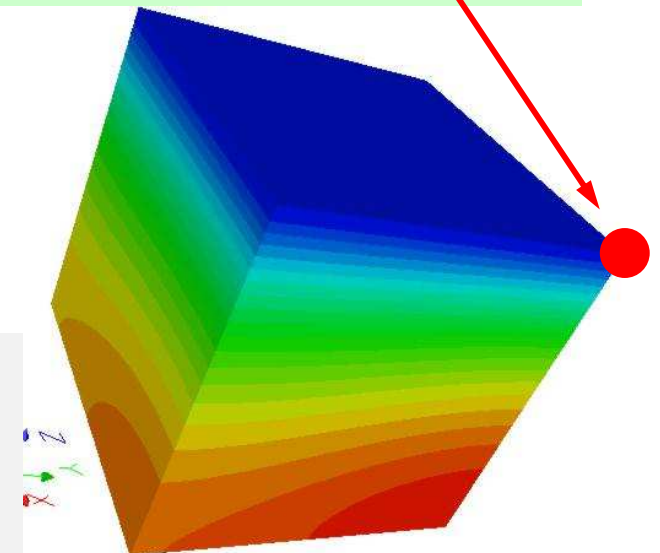
```
##ANSWER      32768      9.297409E+02  Result at ●-point
```

```
$> ls test.inp  
test.inp
```

ParaView

<http://www.paraview.org/>

<http://nkl.cc.u-tokyo.ac.jp/22s/ParaView.pdf>



UCD Format (1/2)

Unstructured Cell Data

要素の種類

点

線

三角形

四角形

四面体

角錐

三角柱

六面体

二次要素

線2

三角形2

四角形2

四面体2

角錐2

三角柱2

六面体2

キーワード

pt

line

tri

quad

tet

pyr

prism

hex

line2

tri2

quad2

tet2

pyr2

prism2

hex2

点

0

線

0 1

線2

0 2 1

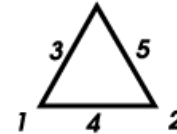
三角形

0

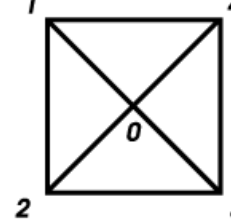


三角形2

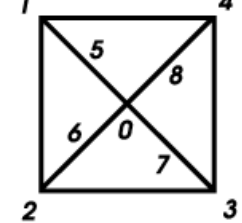
0



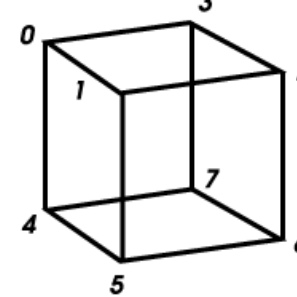
四角錐



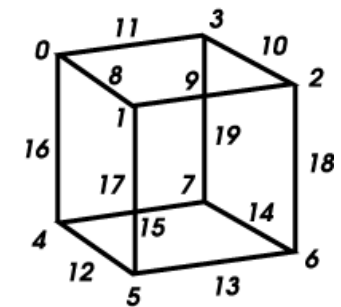
四角錐2



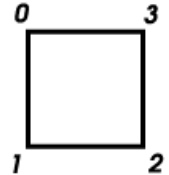
六面体



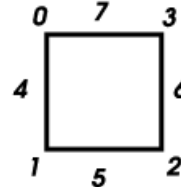
六面体2



四角形

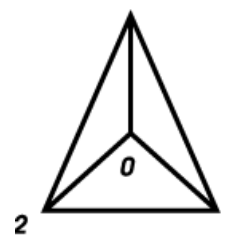


四角形2



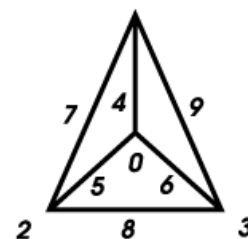
三角錐

1

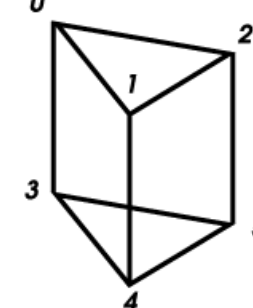


三角錐2

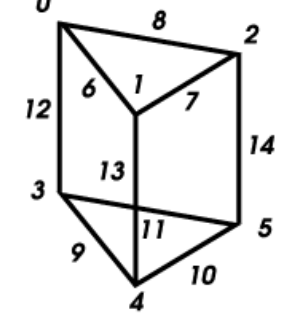
1



三角柱



三角柱2



UCD Format (2/2)

- Originally for AVS, microAVS
- Extension of the UCD file is “inp”
- There are two types of formats. Only old type can be read by ParaView.

- Background
 - Finite Volume Method
 - Preconditioned Iterative Solvers
- **PCG Solver for Poisson's Equations**
 - How to run
 - Data Structure
 - **Program**
 - **Initialization**
 - **Coefficient Matrices**
 - PCG

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <errno.h>
#include "struct.h"
#include "pcg.h"
#include "input.h" ...
```

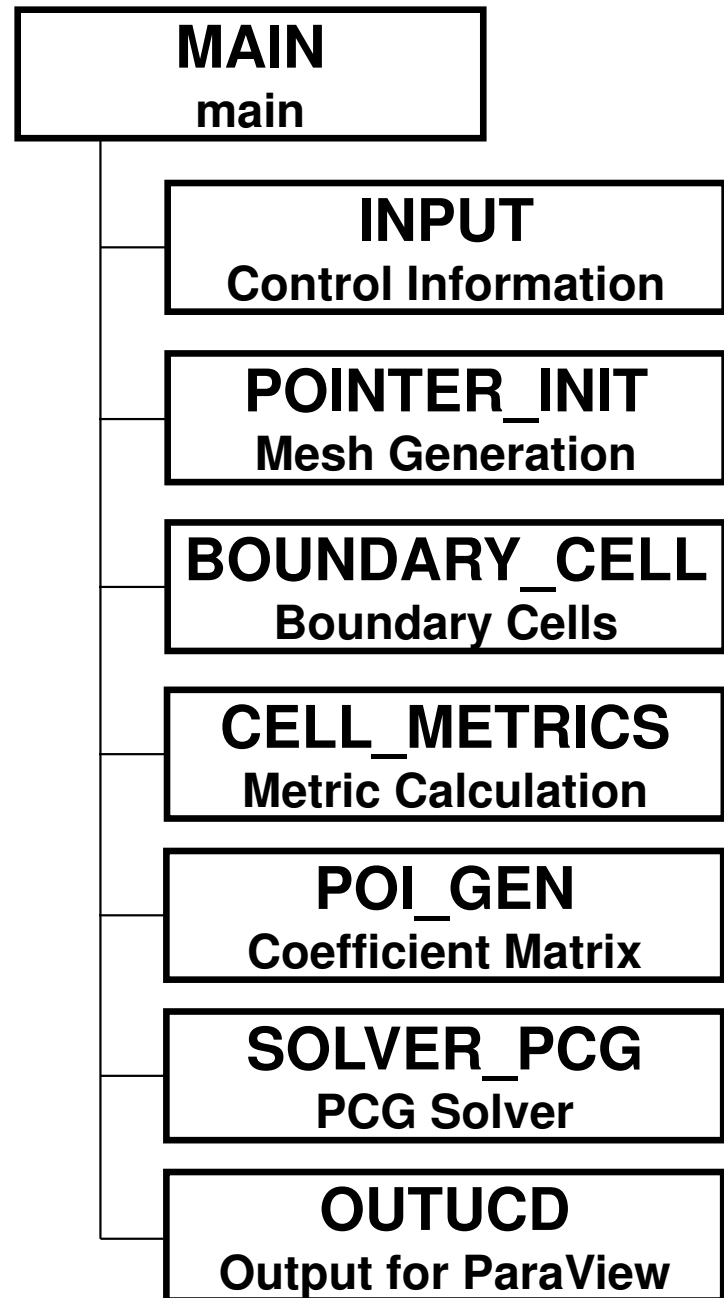
```
int
main()
{
    double *WK;
    int NPL, NPU; ISET, ITR, IER; icel, ic0, i;
    double xN, xL, xU; Stime, Etime;

    if(INPUT()) goto error;
    if(POINTER_INIT()) goto error;
    if(BOUNDARY_CELL()) goto error;
    if(CELL_METRICS()) goto error;
    if(POI_GEN()) goto error;

    memset(PHI, 0.0, sizeof(double)*ICELTOT);
    ISET = 0;
    WK = (double *)malloc(sizeof(double)*ICELTOT);
    if(solve_PCG(...)) goto error;}

    if(OUTUCD()) goto error;
    return 0;
error:
    return -1;
}
```

Structure of the Program



struct.h: Variables/Arrays for FVM

```
#ifndef __H_STRUCT
#define __H_STRUCT

#include <omp.h>

int ICELTOT, ICELTOTp, N;
int NX, NY, NZ, NXP1, NYP1, NZP1, IBNODTOT;
int NXc, NYc, NZc;

double DX, DY, DZ, XAREA, YAREA, ZAREA;
double RDX, RDY, RDZ, RDX2, RDY2, RDZ2, R2DX, R2DY, R2DZ;
double *VOLCEL, *VOLNOD, *RVC, *RVN;

int **XYZ, **NEIBcell;

int ZmaxCELtot;
int *BC_INDEX, *BC_NOD;
int *ZmaxCEL;

int **IWKX;
double **FCV;

int my_rank, PETOT, PEsmptOT;

#endif /* __H_STRUCT */
```

struct.h: Variables/Arrays for FVM

Name	Type	Size	Content
NX, NY, NZ	I		Number of meshes in x/y/z directions
ICELTOT	I		Total number of meshes (NX x NY x NZ)
N	I		Total number of nodes (for visualization)
NXP1,NYP1, NZP1	I		Number of nodes in x/y/z directions (for visualization)
IBNODTOT	I		$NXP1 \times NYP1$
XYZ	I	[ICELTOT][3]	Location of meshes
NEIBcell	I	[ICELTOT][6]	Neighboring meshes

pcg.h: Variables/Arrays for Sparse Matrix

```
#ifndef __H_PCG
#define __H_PCG

    int N2;
    int NLUmax, NLU;
    int METHOD, ORDER_METHOD;

    double EPSICCG;

    double *D, *PHI, *BFORCE;
    double *AMAT;

    int *INLU, *indexLU, *itemLU;
    int NPLU;

#endif /* __H_PCG */
```

pcg.h: Variables/Arrays for Sparse Matrix

Name	Type	Size	Content
NLU	I		Maximum number of neighbors for each mesh (=6)
EPSICCG	R		Convergence Criteria for PCG
NPLU	I		Total number of non-zero off-diagonal components (CRS)
indexLU	I	[ICELTOT+1]	Number of non-zero off-diagonal components (CRS)
itemLU	I	[NPLU]	Column ID of non-zero off-diagonal components (CRS)
PHI	R	[ICELTOT]	Unknown vector
BFORCE	R	[ICELTOT]	RHS vector
D	R	[ICELTOT]	Diagonal components of the matrix
AMAT	R	[NPLU]	Non-zero off-diagonal components of the matrix (CRS)
INLU	I	[ICELTOT]	Number of non-zero off-diagonal components for each mesh, to be used for calculation of NPLU

Mat-Vec Multiplication: $\{q\}=[A]\{p\}$

```
for (i=0; i<N; i++) {  
    q[i]= D[i] * p[i];  
    for (j=indexLU[i]; j<indexLU[i+1]; j++) {  
        q[i] += AMAT[j] * p[itemLU[j]];  
    }  
}
```

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <errno.h>
#include "struct.h"
#include "pcg.h"
#include "input.h" ...
```

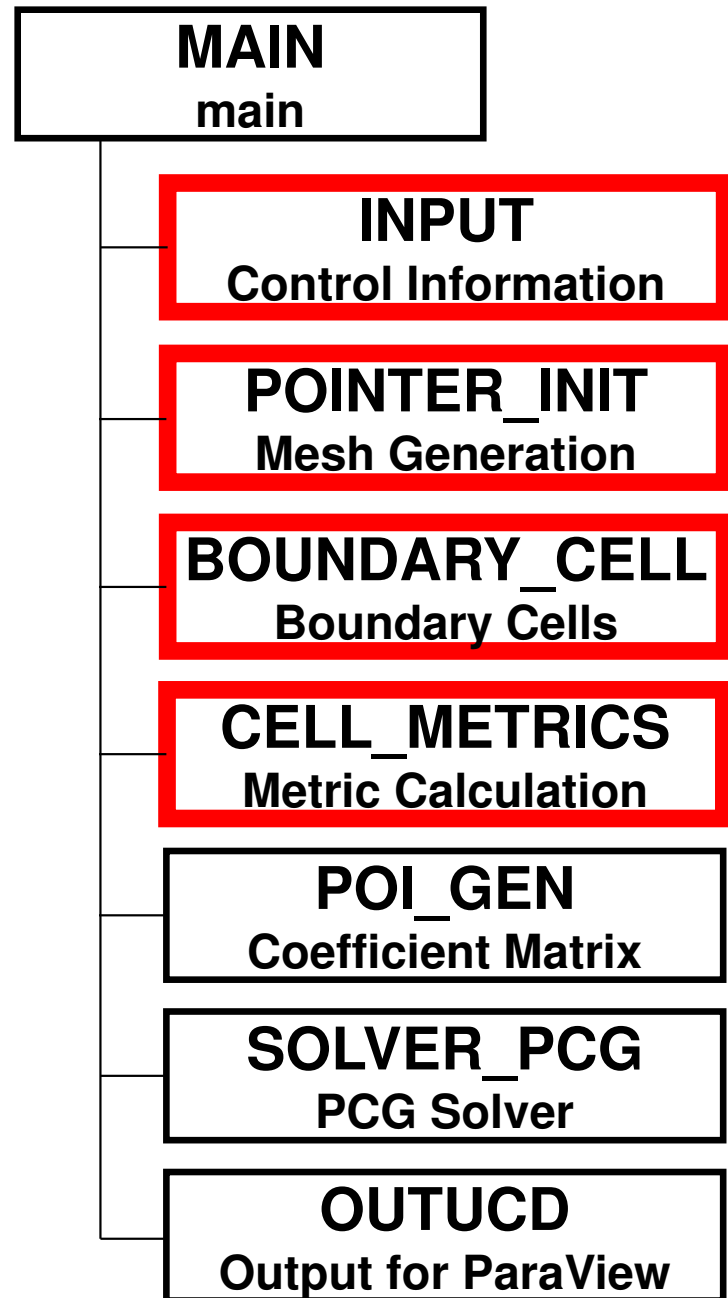
```
int
main()
{
    double *WK;
    int NPL, NPU; ISET, ITR, IER; icel, ic0, i;
    double xN, xL, xU; Stime, Etime;

    if(INPUT()) goto error;
    if(POINTER_INIT()) goto error;
    if(BOUNDARY_CELL()) goto error;
    if(CELL_METRICS()) goto error;
    if(POI_GEN()) goto error;

    memset(PHI, 0.0, sizeof(double)*ICELTOT);
    ISET = 0;
    WK = (double *)malloc(sizeof(double)*ICELTOT);
    if(solve_PCG(...)) goto error;}

    if(OUTUCD()) goto error;
    return 0;
error:
    return -1;
}
```

Structure of the Program



input: reading “INPUT.DAT”

```
#include <stdio.h>; <stdlib.h>; <string.h>; <errno.h>
#include "struct_ext.h"; "pcg_ext.h"; "input.h"

extern int
INPUT(void)
{
#define BUF_SIZE 1024
    char line[BUF_SIZE];
    char CNTFIL[81];
    double OMEGA;
    FILE *fp11;

    if((fp11 = fopen("INPUT.DAT", "r")) == NULL) {
        fprintf(stderr, "Error: %s\n", strerror(errno));
        return -1;
    }
    fgets(line, BUF_SIZE, fp11); sscanf(line, "%d%d%d", &NX, &NY, &NZ);
    fgets(line, BUF_SIZE, fp11); sscanf(line, "%le%le%le", &DX, &DY, &DZ);
    fgets(line, BUF_SIZE, fp11); sscanf(line, "%le", &EPSICCG);
    fgets(line, BUF_SIZE, fp11);

    fclose(fp11);
    return 0;
}
```

32 32 32

NX/NY/NZ

1.00e-00 1.00e-00 1.00e-00

DX/DY/DZ

1.0e-08

EPSICCG

pointer_init (1/3): “mesh.dat”

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <errno.h>

#include "struct_ext.h"
#include "pcg_ext.h"
#include "pointer_init.h"
#include "allocate.h"

extern int
POINTER_INIT(void)
{
    int icel, ipe, i, j, k;

    /*
     * INIT.
     */

    ICELTOT = NX * NY * NZ;

    NXP1 = NX + 1;
    NYP1 = NY + 1;
    NZP1 = NZ + 1;

    NEIBcell =
        (int **)allocate_matrix(sizeof(int), ICELTOT, 6);

    XYZ =
        (int **)allocate_matrix(sizeof(int), ICELTOT, 3);
```

NX, NY, NZ :

Number of meshes in x/y/z directions

NXP1, NYP1, NZP1 :

Number of nodes in x/y/z directions
(for visualization)

ICELTOT :

Number of meshes (NX x NY x NZ)

XYZ[ICELTOT][3] :

Location of meshes

NEIBcell[ICELTOT][6] :

Neighboring meshes

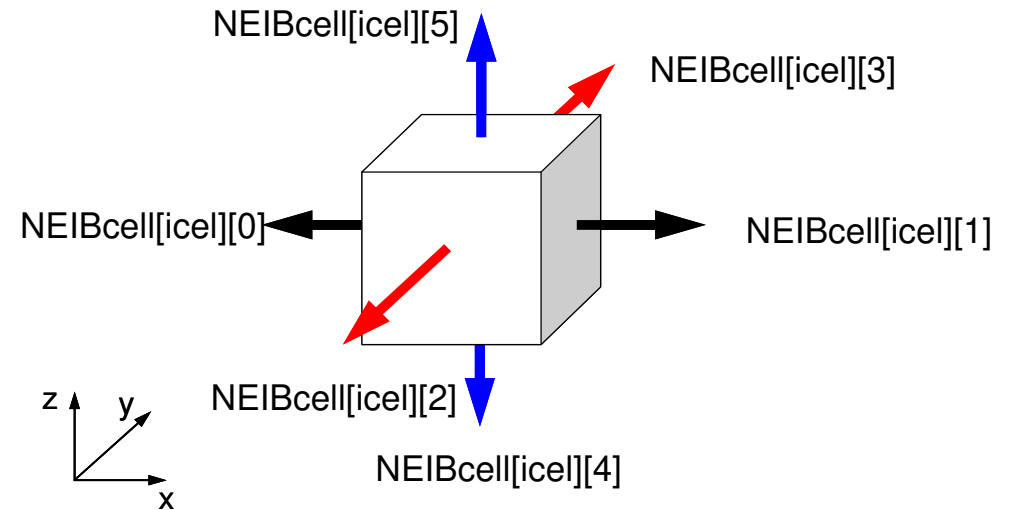
allocate/deallocate

pointer_init (2/3): “mesh.dat”

```

for(k=0; k<NZ; k++) {
  for(j=0; j<NY; j++) {
    for(i=0; i<NX; i++) {
      icel = k * NX * NY + j * NX + i;
      NEIBcell[icel][0] = icel - 1;
      NEIBcell[icel][1] = icel + 1;
      NEIBcell[icel][2] = icel - NX;
      NEIBcell[icel][3] = icel + NX;
      NEIBcell[icel][4] = icel - NX * NY + 1;
      NEIBcell[icel][5] = icel + NX * NY + 1;
      if(i == 0) NEIBcell[icel][0] = 0;
      if(i == NX-1) NEIBcell[icel][1] = 0;
      if(j == 0) NEIBcell[icel][2] = 0;
      if(j == NY-1) NEIBcell[icel][3] = 0;
      if(k == 0) NEIBcell[icel][4] = 0;
      if(k == NZ-1) NEIBcell[icel][5] = 0;
      XYZ[icel][0] = i + 1;
      XYZ[icel][1] = j + 1;
      XYZ[icel][2] = k + 1;
    }
  }
}

```



$i+1 = \text{XYZ}[\text{icel}][0]$
 $j+1 = \text{XYZ}[\text{icel}][1], k+1 = \text{XYZ}[\text{icel}][2]$
 $\text{icel} = k * \text{NX} * \text{NY} + j * \text{NX} + i$

“icel” starts at 0

12	13	14	15
8	9	10	11
4	5	6	7
0	1	2	3

“NEIBcell” starts at 1

13	14	15	16
9	10	11	12
5	6	7	8
1	2	3	4

$\text{NEIBcell}[\text{icel}][0] = \text{icel} - 1$
 $\text{NEIBcell}[\text{icel}][1] = \text{icel} + 1$
 $\text{NEIBcell}[\text{icel}][2] = \text{icel} - \text{NX}$
 $\text{NEIBcell}[\text{icel}][3] = \text{icel} + \text{NX}$
 $\text{NEIBcell}[\text{icel}][4] = \text{icel} - \text{NX} * \text{NY} + 1$
 $\text{NEIBcell}[\text{icel}][5] = \text{icel} + \text{NX} * \text{NY} + 1$

pointer_init (3/3): “mesh.dat”

```
if(DX <= 0.0) {  
    DX = 1.0 / (double)NX;  
    DY = 1.0 / (double)NY;  
    DZ = 1.0 / (double)NZ;  
}
```

```
NXP1 = NX + 1;  
NYP1 = NY + 1;  
NZP1 = NZ + 1;
```

```
IBNODTOT = NXP1 * NYP1;  
N         = NXP1 * NYP1 * NZP1;
```

```
return 0;
```

```
}
```

if DX is no larger than 0.0

pointer_init (3/3): “mesh.dat”

```

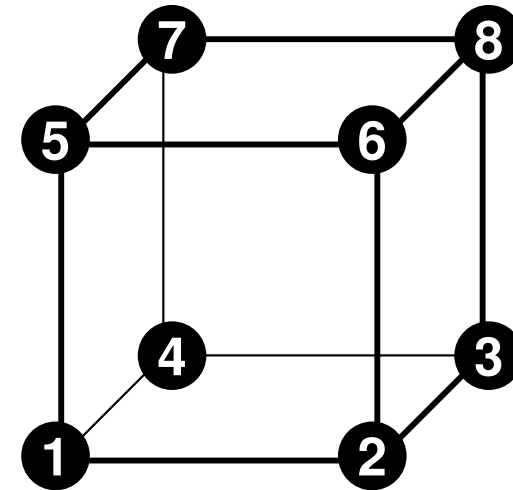
if(DX <= 0.0) {
    DX = 1.0 / (double)NX;
    DY = 1.0 / (double)NY;
    DZ = 1.0 / (double)NZ;
}

NXP1 = NX + 1;
NYP1 = NY + 1;
NZP1 = NZ + 1;

IBNODTOT = NXP1 * NYP1;
N         = NXP1 * NYP1 * NZP1;

return 0;
}

```



NXP1, NYP1, NZP1:

Number of nodes in x/y/z directions

IBNODTOT:

= NXP1 x NYP1

N:

Number of modes
meshes (for visualization)

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <errno.h>
```

```
#include "struct_ext.h"
#include "boundary_cell.h"
#include "allocate.h"
```

```
extern int
BOUNDARY_CELL(void)
{
    int IFACTOT;
    int icou, icel, i, j, k;
```

```
    IFACTOT = NX * NY;
    ZmaxCEltot = IFACTOT;
```

```
    ZmaxCEL =
        (int *)allocate_vector(sizeof(int), ZmaxCEltot);
```

```
    icou = 0;
    k = NZ - 1;
```

```
    for(j=0; j<NY; j++) {
        for(i=0; i<NX; i++) {
            icel = k*IFACTOT + j*NX + i+1;
            ZmaxCEL[icou] = icel;
            icou++;
        }
    }
```

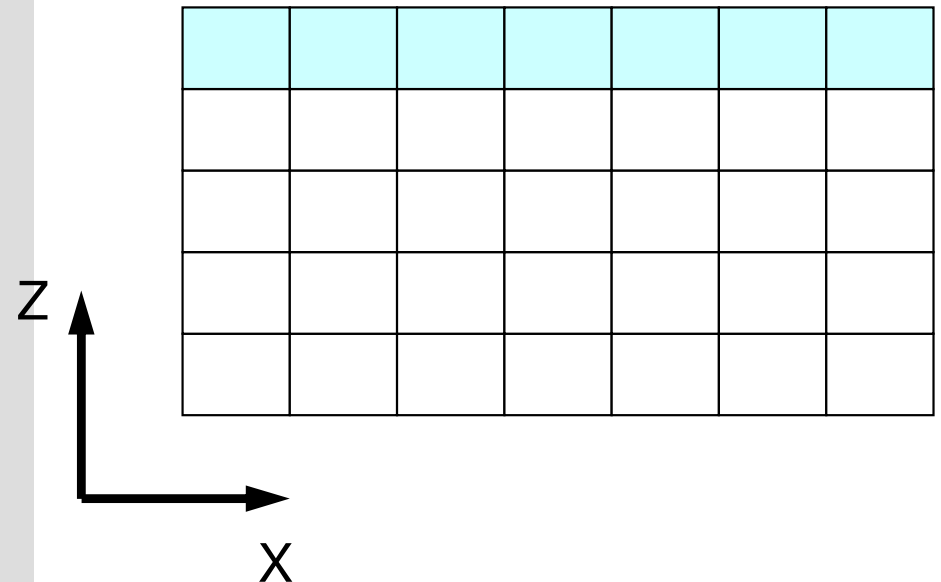
```
    return 0;
}
```

boundary_cell

Meshes @ $Z=Z_{\max}$

Number: $Z_{\max} \text{CEL}_{\text{tot}}$

Mesh ID: $Z_{\max} \text{CEL}[:]$



```
/******
allocate vector
*****/
```

allocate.c

```
void* allocate_vector(int size, int m)
{
    void *a;
    if ( ( a=(void *)malloc( m * size ) ) == NULL ) {
        fprintf(stdout, "Error:Memory does not enough! in vector %n");
        exit(1);
    }
    return a;
}
```

```
#include <stdio.h> ...
```

```
extern int
CELL_METRICS(void)
{
    double V0, RVO;
    int i;
    VOLCEL =
    (double*) allocate_vector(sizeof(double), ICELTOT);
    RVC =
    (double*) allocate_vector(sizeof(double), ICELTOT);
```

```
XAREA = DY * DZ;
YAREA = DZ * DX;
ZAREA = DX * DY;
```

```
RDY = 1.0 / DY;
RDZ = 1.0 / DZ;
```

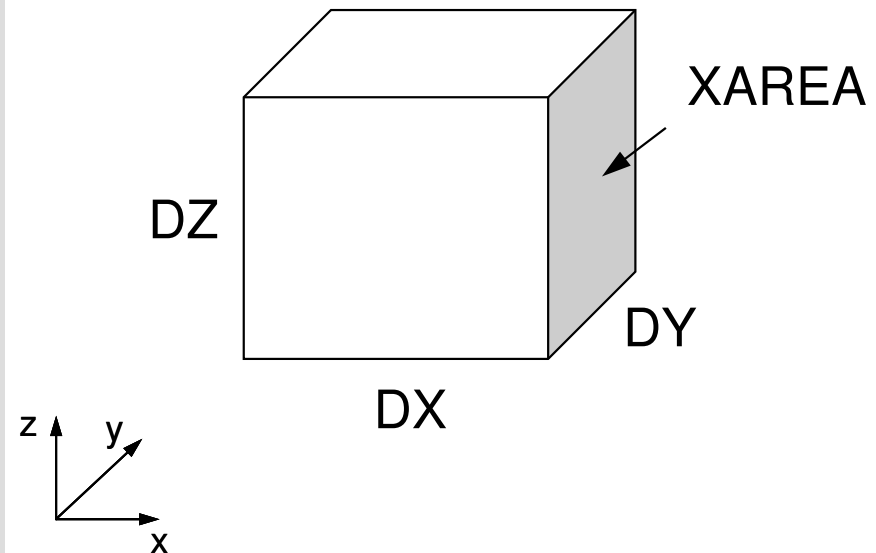
```
RDY2 = 1.0 / (pow(DY, 2.0));
RDZ2 = 1.0 / (pow(DZ, 2.0));
R2DX = 1.0 / (0.5 * DX);
R2DY = 1.0 / (0.5 * DY);
R2DZ = 1.0 / (0.5 * DZ);
```

```
V0 = DX * DY * DZ;
RVO = 1.0 / V0;
```

```
for(i=0; i<ICELTOT; i++) {
    VOLCEL[i] = V0;
    RVC[i] = RVO;
}
return 0; }
```

cell_metrics

Parameters for Computations



$$XAREA = \Delta Y \times \Delta Z, \quad YAREA = \Delta Z \times \Delta X, \\ ZAREA = \Delta X \times \Delta Y$$

$$RDY = \frac{1}{\Delta Y}, \quad RDZ = \frac{1}{\Delta Z}$$

```
#include <stdio.h> ...
```

```
extern int
CELL_METRICS(void)
{
    double V0, RV0;
    int i;

    VOLCEL =
    (double*) allocate_vector(sizeof(double), ICELTOT);
    RVC =
    (double*) allocate_vector(sizeof(double), ICELTOT);
```

```
XAREA = DY * DZ;
YAREA = DZ * DX;
ZAREA = DX * DY;
```

```
RDY = 1.0 / DY;
RDZ = 1.0 / DZ;
```

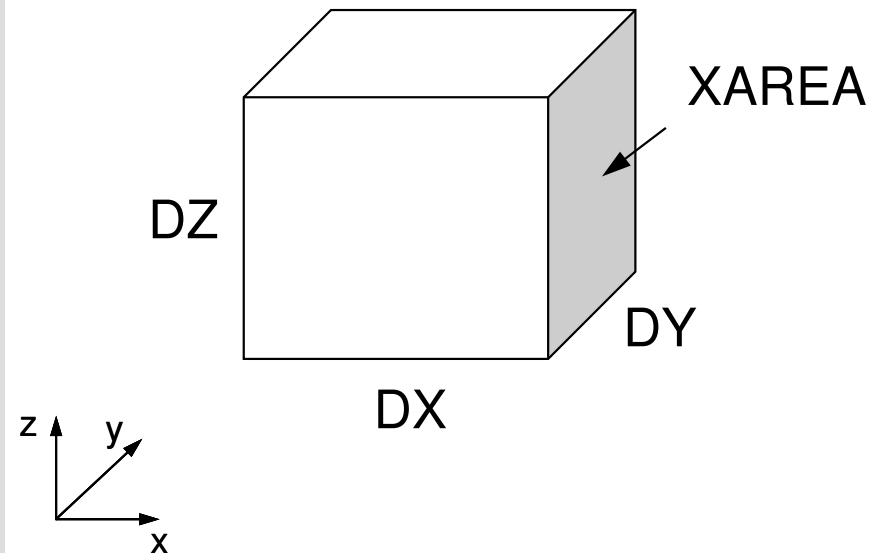
```
RDY2 = 1.0 / (pow(DY, 2.0));
RDZ2 = 1.0 / (pow(DZ, 2.0));
R2DY = 1.0 / (0.5 * DY);
R2DZ = 1.0 / (0.5 * DZ);
```

```
V0 = DX * DY * DZ;
RV0 = 1.0 / V0;
```

```
for(i=0; i<ICELTOT; i++) {
    VOLCEL[i] = V0;
    RVC[i] = RV0;
}
return 0; }
```

cell_metrics

Parameters for Computations



$$RDY2 = \frac{1}{\Delta Y^2}, \quad RDZ2 = \frac{1}{\Delta Z^2}, \quad RDZ2 = \frac{1}{\Delta Z^2}$$

$$R2DY = \frac{1}{0.5 \times \Delta Y}, \quad R2DZ = \frac{1}{0.5 \times \Delta Z}$$

$$R2DZ = \frac{1}{0.5 \times \Delta Z}$$

```
#include <stdio.h> ...
```

```
extern int
CELL_METRICS(void)
{
    double V0, RV0;
    int i;
```

```
VOLCEL =
(double*) allocate_vector (sizeof(double), ICELTOT);
RVC =
(double*) allocate_vector (sizeof(double), ICELTOT);
```

```
XAREA = DY * DZ;
YAREA = DZ * DX;
ZAREA = DX * DY;
```

```
RDY = 1.0 / DY;
RDZ = 1.0 / DZ;
```

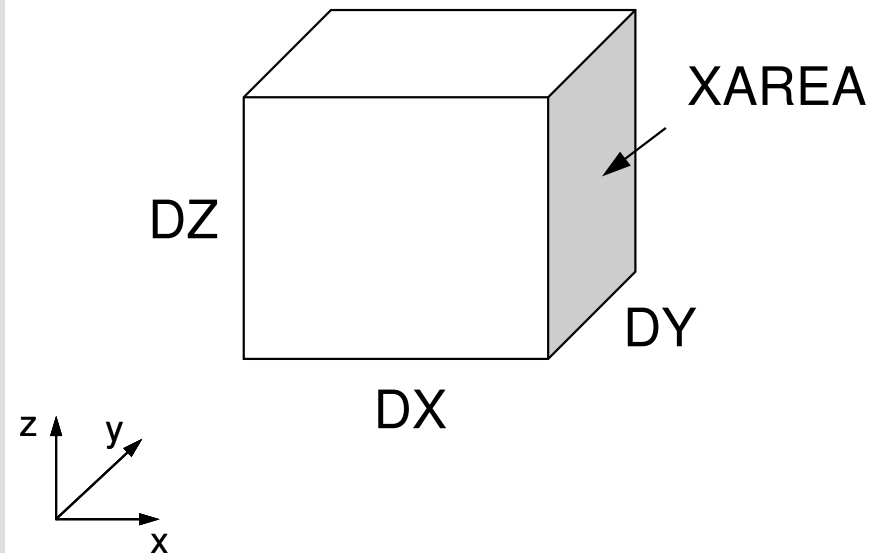
```
RDY2 = 1.0 / (pow(DY, 2.0));
RDZ2 = 1.0 / (pow(DZ, 2.0));
R2DX = 1.0 / (0.5 * DX);
R2DY = 1.0 / (0.5 * DY);
R2DZ = 1.0 / (0.5 * DZ);
```

```
V0 = DX * DY * DZ;
RV0 = 1.0 / V0;
```

```
for(i=0; i<ICELTOT; i++) {
    VOLCEL[i] = V0;
    RVC[i] = RV0;
}
return 0; }
```

cell_metrics

Parameters for Computations



$$VOLCEL = V0 = \Delta X \times \Delta Y \times \Delta Z$$

$$RV0 = RVC = \frac{1}{VOLCEL}$$

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <errno.h>
#include "struct.h"
#include "pcg.h"
#include "input.h" ...
```

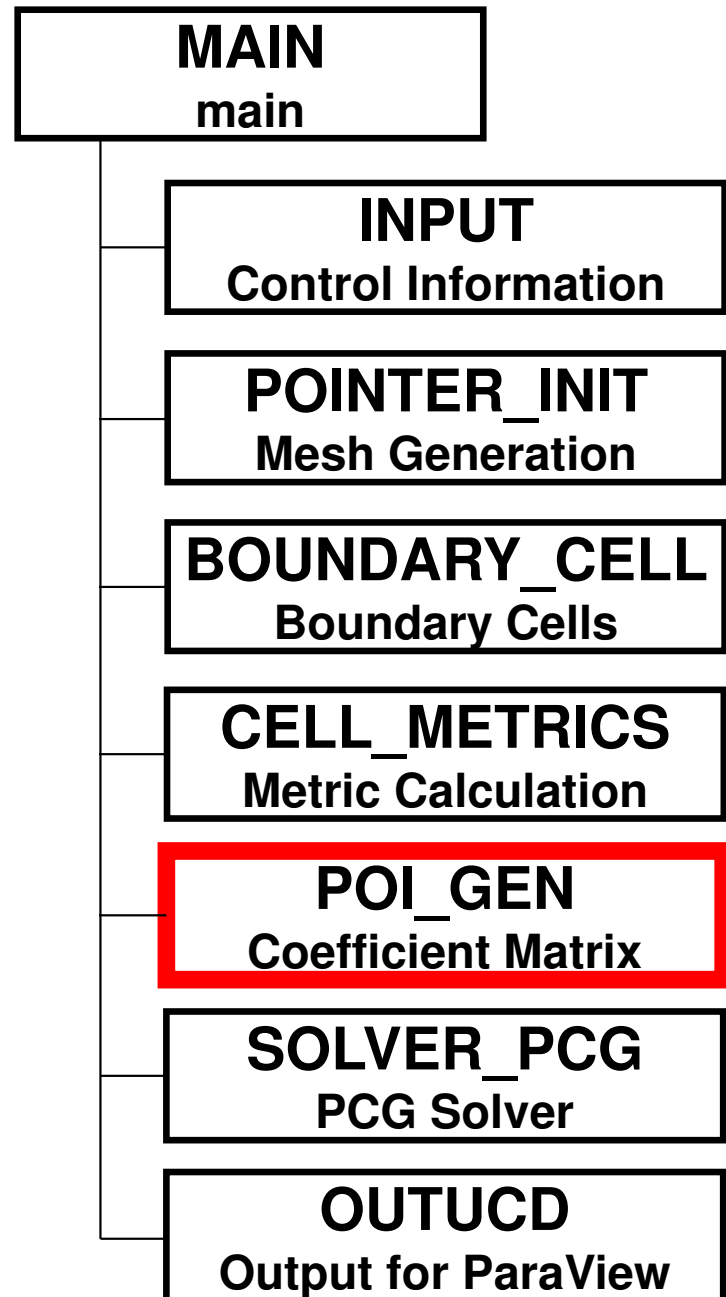
```
int
main()
{
    double *WK;
    int NPL, NPU; ISET, ITR, IER; icel, ic0, i;
    double xN, xL, xU; Stime, Etime;

    if(INPUT()) goto error;
    if(POINTER_INIT()) goto error;
    if(BOUNDARY_CELL()) goto error;
    if(CELL_METRICS()) goto error;
    if(POI_GEN()) goto error;

    memset(PHI, 0.0, sizeof(double)*ICELTOT);
    ISET = 0;
    WK = (double *)malloc(sizeof(double)*ICELTOT);
    if(solve_PCG(...)) goto error;}

    if(OUTUCD()) goto error;
    return 0;
error:
    return -1;
}
```

Structure of the Program



poi_gen (1/6)

```
extern int
POI_GEN(void)
{
    int nn;
    int ic0, icN1, icN2, icN3, icN4, icN5, icN6;
    int i, j, k, ib, ic, ip, icel, icou, icol, icouG;
    int ii, jj, kk, nn1, num, nr, j0, j1;
    double coef, VOL0, S1t, E1t;
    int isLU;
```

```
NLU= 6;
```

```
BFORCE =(double *)allocate_vector(sizeof(double), ICELTOT);
D       =(double *)allocate_vector(sizeof(double), ICELTOT);
PHI     =(double *)allocate_vector(sizeof(double), ICELTOT);
INLU    =(int *)allocate_vector(sizeof(int), ICELTOT);
indexLU=(int *)allocate_vector(sizeof(int), ICELTOT+1);
```

```
for (i = 0; i < ICELTOT ; i++) {
    BFORCE[i]=0.0;
    D[i]     =0.0;
    PHI[i]=0.0;
    INLU[i]  = 0;
}

for (i = 0; i <= ICELTOT ; i++) {
    indexLU[i] = 0;
}
```

```
/******  
    allocate matrix  
*****/  
void** allocate_matrix(int size, int m, int n)  
{  
    void **aa;  
    int i;  
    if ( ( aa=(void **)malloc( m * sizeof(void*) ) ) == NULL ) {  
        fprintf(stdout, "Error:Memory does not enough! aa in matrix %n");  
        exit(1);  
    }  
    if ( ( aa[0]=(void *)malloc( m * n * size ) ) == NULL ) {  
        fprintf(stdout, "Error:Memory does not enough! in matrix %n");  
        exit(1);  
    }  
    for(i=1; i<m; i++) aa[i]=(char*)aa[i-1]+size*n;  
    return aa;  
}
```

pcg.h: Variables/Arrays for Sparse Matrix

Name	Type	Size	Content
NLU	I		Maximum number of neighbors for each mesh (=6)
EPSICCG	R		Convergence Criteria for PCG
NPLU	I		Total number of non-zero off-diagonal components (CRS)
indexLU	I	[ICELTOT+1]	Number of non-zero off-diagonal components (CRS)
itemLU	I	[NPLU]	Column ID of non-zero off-diagonal components (CRS)
PHI	R	[ICELTOT]	Unknown vector
BFORCE	R	[ICELTOT]	RHS vector
D	R	[ICELTOT]	Diagonal components of the matrix
AMAT	R	[NPLU]	Non-zero off-diagonal components of the matrix (CRS)
INLU	I	[ICELTOT]	Number of non-zero off-diagonal components for each mesh, to be used for calculation of NPLU

```
for (icel=0; icel<ICELTOT; icel++) {
```

```
    icN1 = NEIBcell[[icel]][0];
    icN2 = NEIBcell[[icel]][1];
    icN3 = NEIBcell[[icel]][2];
    icN4 = NEIBcell[[icel]][3];
    icN5 = NEIBcell[[icel]][4];
    icN6 = NEIBcell[[icel]][5];
```

```
    if (icN5 != 0) {
        INLU[icel] = INLU[icel] + 1;
    }
```

```
    if (icN3 != 0) {
        INLU[icel] = INLU[icel] + 1;
    }
```

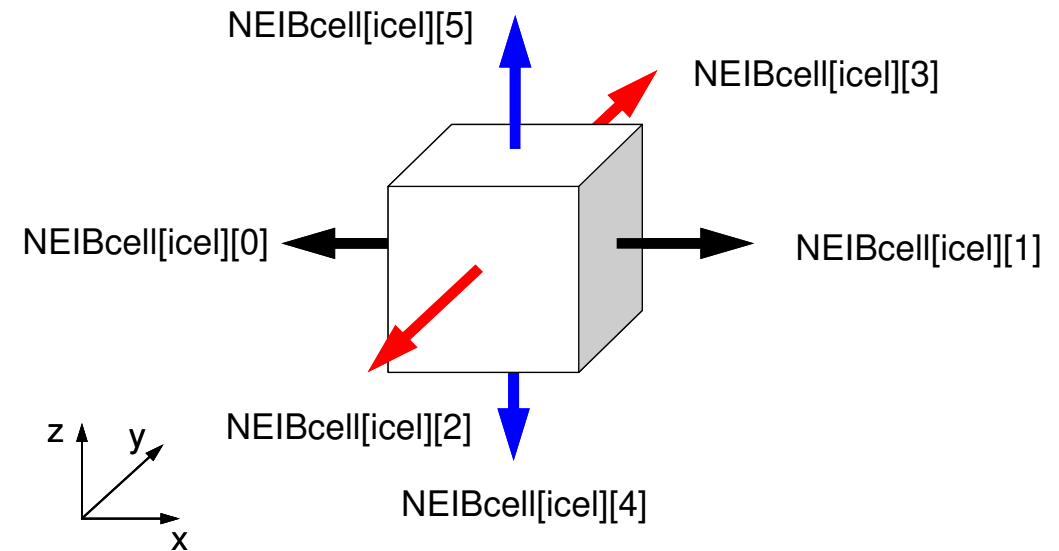
```
    if (icN1 != 0) {
        INLU[icel] = INLU[icel] + 1;
    }
```

```
    if (icN2 != 0) {
        INLU[icel] = INLU[icel] + 1;
    }
```

```
    if (icN4 != 0) {
        INLU[icel] = INLU[icel] + 1;
    }
```

```
    if (icN6 != 0) {
        INLU[icel] = INLU[icel] + 1;
    }
```

poi_gen (2/6)



Lower Triangular Components

$$\text{NEIBcell[icel][4]} = \text{icel} - \text{NX} * \text{NY} + 1$$

$$\text{NEIBcell[icel][2]} = \text{icel} - \text{NX} + 1$$

$$\text{NEIBcell[icel][0]} = \text{icel} - 1 + 1$$

Upper Triangular Components

$$\text{NEIBcell[icel][1]} = \text{icel} + 1 + 1$$

$$\text{NEIBcell[icel][3]} = \text{icel} + \text{NX} + 1$$

$$\text{NEIBcell[icel][5]} = \text{icel} + \text{NX} * \text{NY} + 1$$

poi_gen (3/6)

```

for(i=0; i<ICELTOT; i++) {
    indexLU[i+1]=indexLU[i] + INLU[i];
}

NPLU= indexLU[ICELTOT];

itemLU= (int*)allocate_vector(sizeof(int), NPLU);
AMAT = (double*)allocate_vector(sizeof(double), NPLU);

for(i=0; i<ICELTOT; i++) {
    for(j=indexLU[i]; j<indexLU[i+1]; j++) {
        itemLU[j]=0;
        AMAT[j]=0.0;
    }
}
free(INLU);

```

```

for (i=0; i<N; i++) {
    q[i]= D[i] * p[i];
    for (j=indexLU[i]; j<indexLU[i+1]; j++) {
        q[i] += AMAT[j] * p[itemLU[j]];
    }
}

```

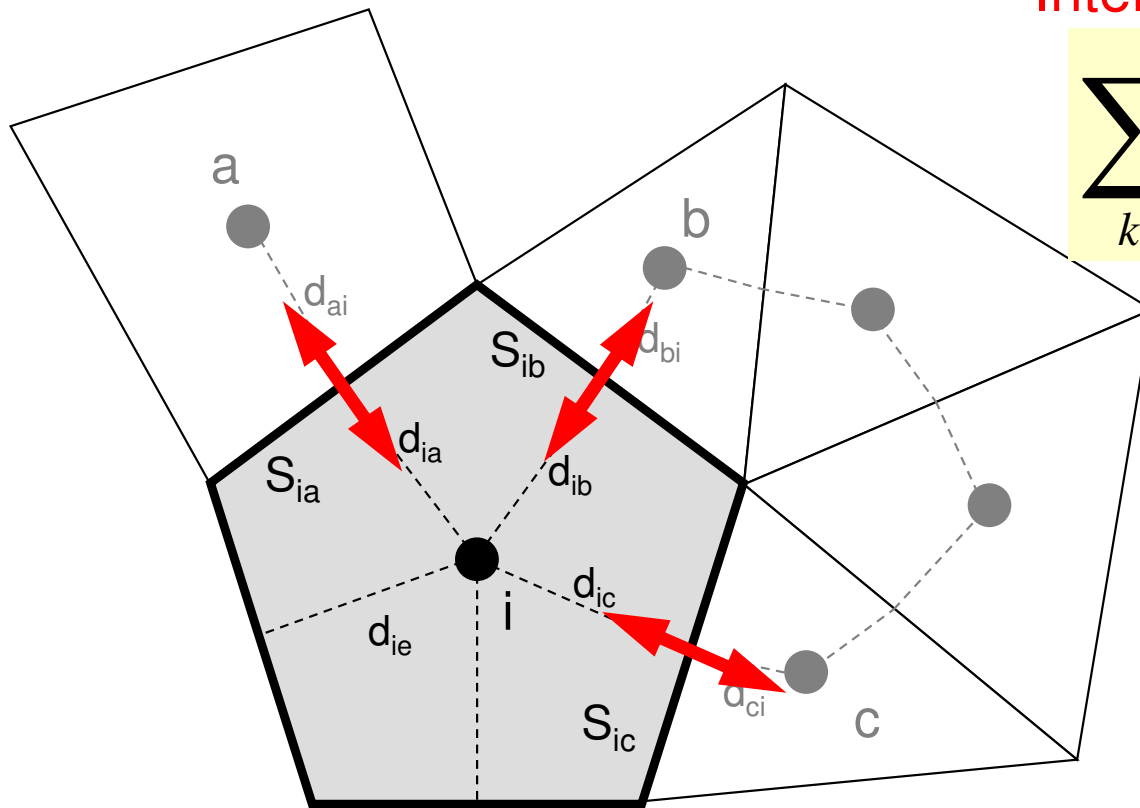
Name	Type	Size	Content
NLU	I		Maximum number of neighbors for each mesh (=6)
EPSICCG	R		Convergence Criteria for PCG
NPLU	I		Total number of non-zero off-diagonal components (CRS)
indexLU	I	[ICELTOT+1]	Number of non-zero off-diagonal components (CRS)
itemLU	I	[NPLU]	Column ID of non-zero off-diagonal components (CRS)
PHI	R	[ICELTOT]	Unknown vector
BFORCE	R	[ICELTOT]	RHS vector
D	R	[ICELTOT]	Diagonal components of the matrix
AMAT	R	[NPLU]	Non-zero off-diagonal components of the matrix (CRS)
INLU	I	[ICELTOT]	Number of non-zero off-diagonal components for each mesh, to be used for calculation of NPLU

Poisson Equation by Finite Volume Method (FVM)

Conservation of Fluxes through
Surfaces

$$\frac{\partial^2 \phi}{\partial x^2} + \frac{\partial^2 \phi}{\partial y^2} + \frac{\partial^2 \phi}{\partial z^2} + f = 0$$

Diffusion:
Interaction with Neighbors



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

Volume Flux

V_i : Volume
 S : Surface Area
 d_{ij} : Distance between
 Cell-Center &
 Surface
 Q : Volume Flux

Constructing Coefficient Matrix

Conservation for i-th mesh

$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$+ \sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k - \sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_i = -V_i \dot{Q}_i$$

$$- \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

D (diagonal)

**AMAT
(off-diag.)**

**BFORCE
(RHS)**

```

for (icel=0; icel<ICELTOT; icel++) {
    icN1 = NEIBcell[icel][0];
    icN2 = NEIBcell[icel][1];
    icN3 = NEIBcell[icel][2];
    icN4 = NEIBcell[icel][3];
    icN5 = NEIBcell[icel][4];
    icN6 = NEIBcell[icel][5];
    VOL0 = VOLCEL[icel];
    isLU = indexLU[icel];

```

```

    icou= 0;
    if(icN5 != 0) {
        coef = RDZ * ZAREA;
        D[icel] -= coef;
        itemLU[icou+isLU]= icN5 - 1
        AMAT [icou+isLU]= ccoef;
        icou= icou + 1;
    }

```

```

    if(icN3 != 0) {
        coef = RDY * YAREA;
        D[icel] -= coef;
        itemLU[icou+isLU]= icN3 - 1
        AMAT [icou+isLU]= ccoef;
        icou= icou + 1;
    }

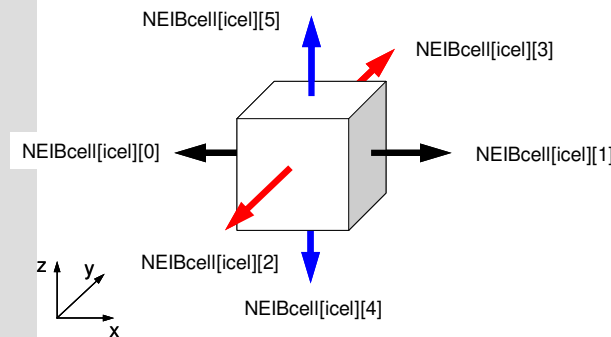
```

```

    if(icN1 != 0) {
        coef = RDX * XAREA;
        D[icel] -= coef;
        itemLU[icou+isLU]= icN1 - 1
        AMAT [icou+isLU]= ccoef;
        icou= icou + 1;
    }

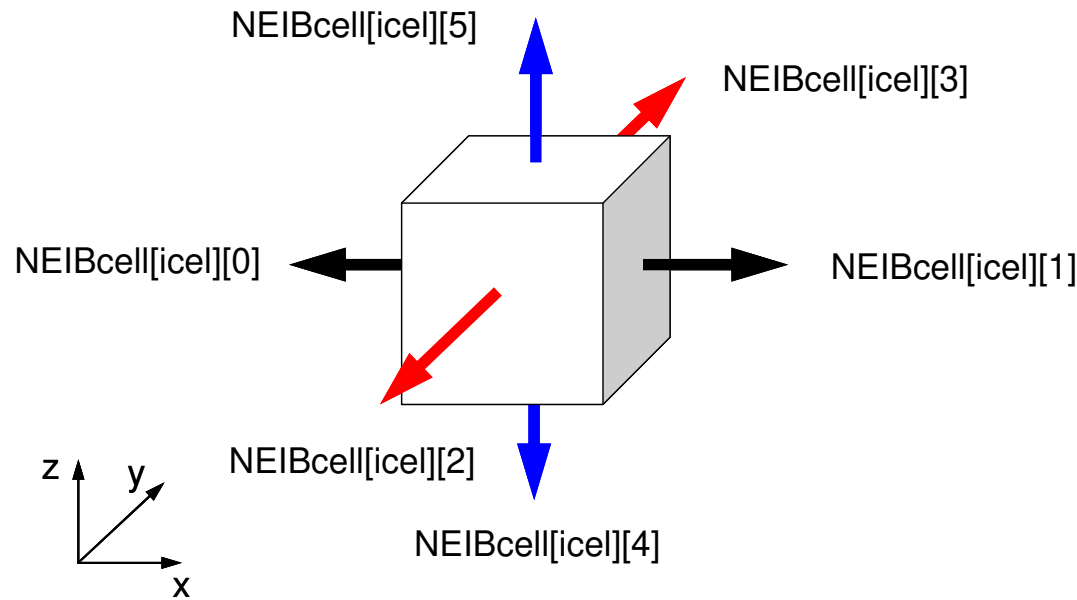
```

poi_gen (4/6)



$$\begin{aligned}
 & \frac{\phi_{neib[icel][0]} - \phi_{icel}}{\Delta x} \Delta y \Delta z + \frac{\phi_{neib[icel][1]} - \phi_{icel}}{\Delta x} \Delta y \Delta z + \\
 & \frac{\phi_{neib[icel][2]} - \phi_{icel}}{\Delta y} \Delta z \Delta x + \frac{\phi_{neib[icel][3]} - \phi_{icel}}{\Delta y} \Delta z \Delta x + \\
 & \boxed{\frac{\phi_{neib[icel][4]} - \phi_{icel}}{\Delta z} \Delta x \Delta y} + \frac{\phi_{neib[icel][5]} - \phi_{icel}}{\Delta z} \Delta x \Delta y + f_{icel} \Delta x \Delta y \Delta z = 0
 \end{aligned}$$

Calculations of Coefficients



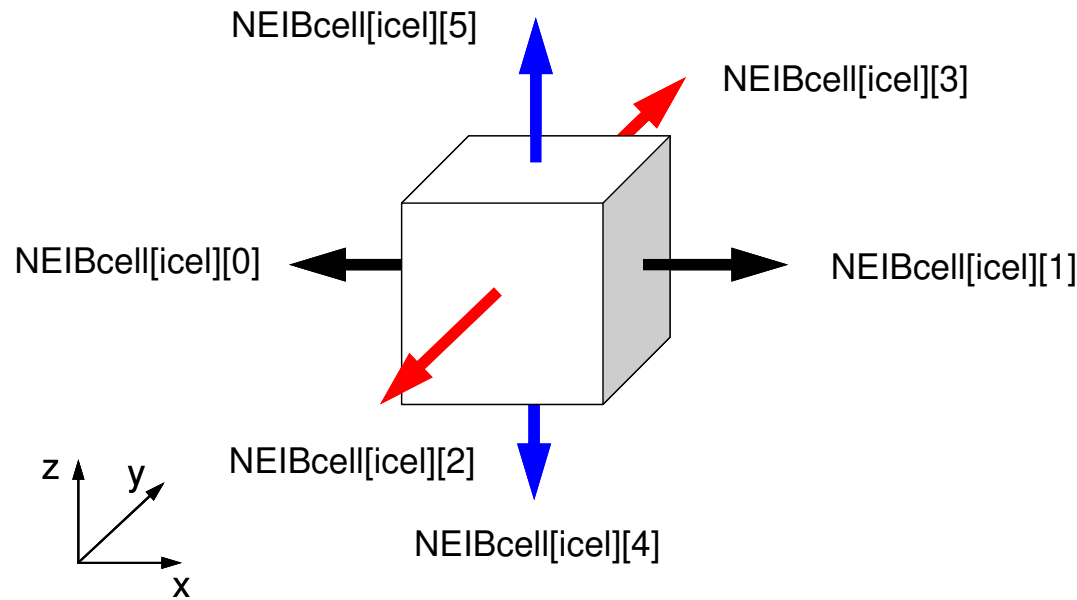
```
if(icN5 != 0) {
    coef = RDZ * ZAREA;
    D[icel] -= coef;
    itemLU[icou+isLU] = icN5-1;
    AMAT [icou+isLU] = coef;
    icou = icou + 1;
}
```

$$\frac{\phi_{neib[icel][0]} - \phi_{icel}}{\Delta x} \Delta y \Delta z + \frac{\phi_{neib[icel][1]} - \phi_{icel}}{\Delta x} \Delta y \Delta z +$$

$$\frac{\phi_{neib[icel][2]} - \phi_{icel}}{\Delta y} \Delta z \Delta x + \frac{\phi_{neib[icel][3]} - \phi_{icel}}{\Delta y} \Delta z \Delta x +$$

$$\frac{\phi_{neib[icel][4]} - \phi_{icel}}{\Delta z} \Delta x \Delta y + \frac{\phi_{neib[icel][5]} - \phi_{icel}}{\Delta z} \Delta x \Delta y + f_{icel} \Delta x \Delta y \Delta z = 0$$

Calculations of Coefficients



```
if(icN5 != 0) {
    coef = RDZ * ZAREA;
    D[icel] -= coef;
    itemLU[icou+isLU] = icN5-1;
    AMAT [icou+isLU] = coef;
    icou = icou + 1;
}
```

$$\frac{\Delta y \Delta z}{\Delta x} (\phi_{neib[icel][0]} - \phi_{icel}) + \frac{\Delta y \Delta z}{\Delta x} (\phi_{neib[icel][1]} - \phi_{icel}) +$$

$$\frac{\Delta z \Delta x}{\Delta y} (\phi_{neib[icel][2]} - \phi_{icel}) + \frac{\Delta z \Delta x}{\Delta y} (\phi_{neib[icel][3]} - \phi_{icel}) +$$

ZAREA

RDZ

$$\frac{\Delta x \Delta y}{\Delta z} (\phi_{neib[icel][4]} - \phi_{icel}) + \frac{\Delta x \Delta y}{\Delta z} (\phi_{neib[icel][5]} - \phi_{icel}) + f_{icel} \Delta x \Delta y \Delta z = 0$$

poi_gen (5/6)

```

if(icN2 != 0) {
    coef = RDX * XAREA;
    D[icel] -= coef;
    itemLU[icou+isLU]= icN2 - 1
    AMAT [icou+isLU]= conef;
    icou= icou + 1;
}

if(icN4 != 0) {
    coef = RDY * YAREA;
    D[icel] -= coef;
    itemLU[icou+isLU]= icN4 - 1
    AMAT [icou+isLU]= conef;
    icou= icou + 1;
}

if(icN6 != 0) {
    coef = RDZ * ZAREA;
    D[icel] -= coef;
    itemLU[icou+isLU]= icN6 - 1
    AMAT [icou+isLU]= conef;
    icou= icou + 1;
}

ii = XYZ[icel][0];
jj = XYZ[icel][1];
kk = XYZ[icel][2];

BFORCE[icel]= -(double)(ii+jj+kk) *
               VOLCEL[icel];
}

```

poi_gen (5/6)

Volume Flux

```

if(icN2 != 0) {
    coef = RDX * XAREA;
    D[icel] -= coef;
    itemLU[icou+isLU] = icN2 - 1;
    AMAT[icou+isLU] = conef;
    icou = icou + 1;
}

```

```

if(icN2 != 0) {
    coef = RDY * YAREA;
    D[icel] -= coef;
    itemLU[icou+isLU] = icN2 - 1;
    AMAT[icou+isLU] = conef;
    icou = icou + 1;
}

```

```

if(icN6 != 0) {
    coef = RDZ * ZAREA;
    D[icel] -= coef;
    itemLU[icou+isLU] = icN6 - 1;
    AMAT[icou+isLU] = conef;
    icou = icou + 1;
}

```

```

ii = XYZ[icel][0];
jj = XYZ[icel][1];
kk = XYZ[icel][2];

```

```

BFORCE[icel] = -(double)(ii+jj+kk) *
                VOLCEL[icel];

```

$$f = dfloat(i_0 + j_0 + k_0)$$

$$i_0 = XYZ[icel][0],$$

$$j_0 = XYZ[icel][1],$$

$$k_0 = XYZ[icel][2]$$

$XYZ[icel][k]$ (k=0,1,2)

Index for location of finite-difference mesh in X-/Y-/Z-axis.

```
/* TOP SURFACE */
```

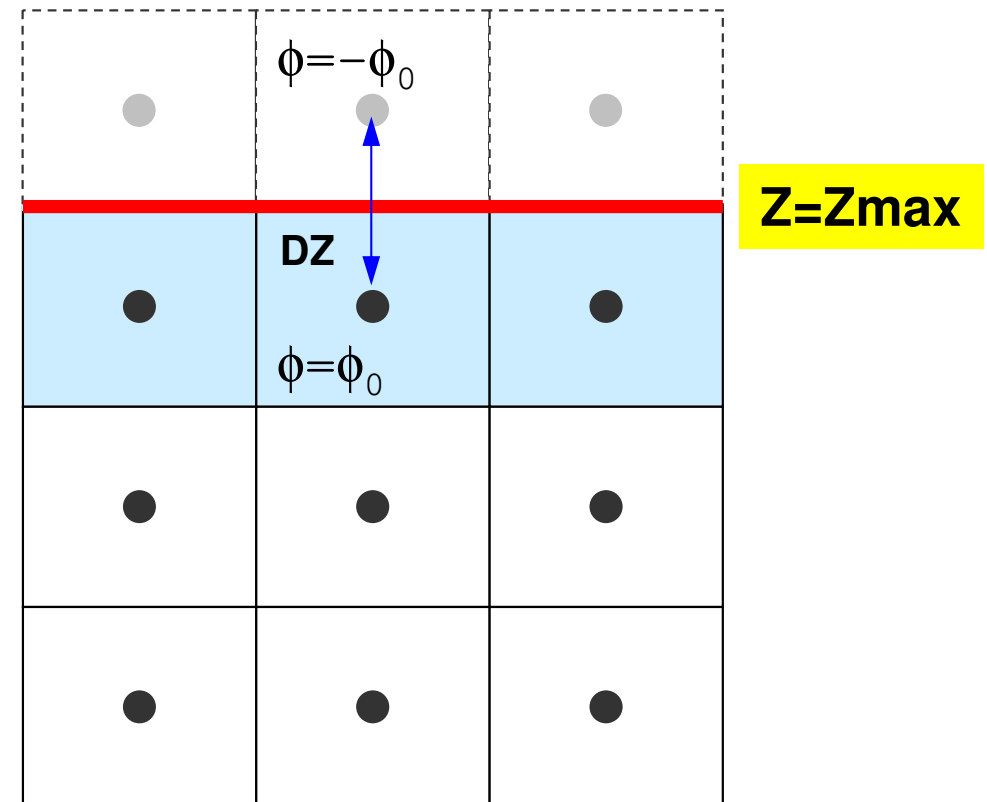
```
for(ib=0; ib<ZmaxCELtot; ib++) {  
    icel = ZmaxCEL[ib];  
    coef = 2.0 * RDZ * ZAREA;  
    D[icel-1] -= coef;  
}
```

```
return 0;
```

```
}
```

poi_gen (6/6)

Calculation of Coefficients
on Boundary Surface @ $Z=Z_{\max}$



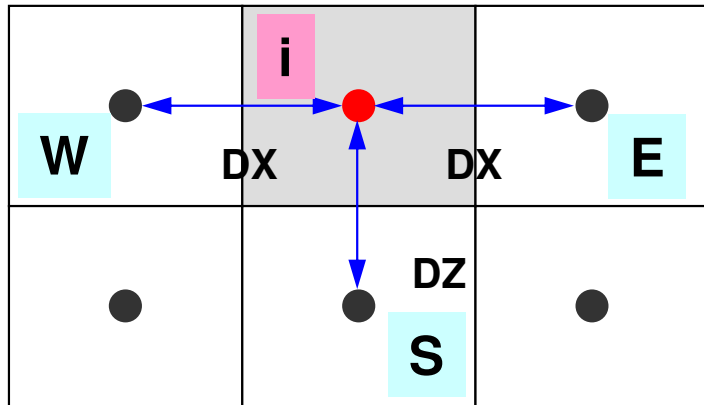
1st Order Approximation:

Mirror Image according to $Z=Z_{\max}$ surface.

$\phi = -\phi_0$ at the center of the (virtual) mesh

$\phi = 0$ @ $Z=Z_{\max}$ surface

Dirichlet B.C.



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

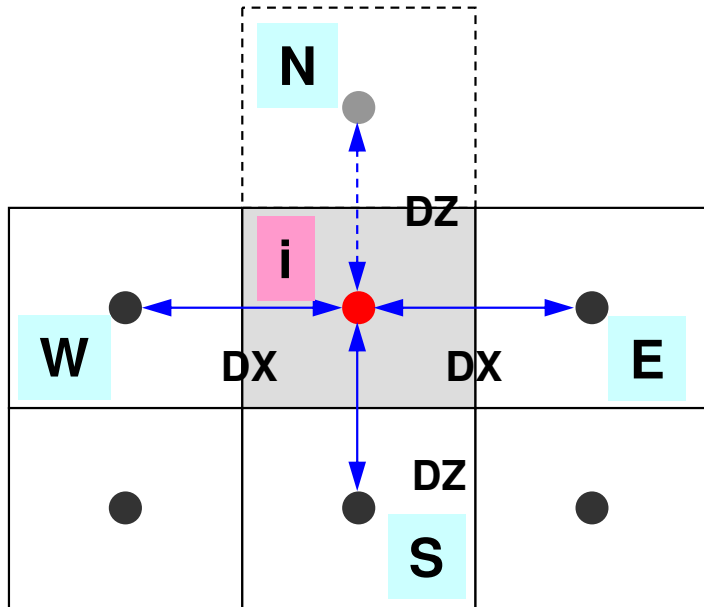
$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

D (diagonal)

**AMAT
(off-diag.)**

**BFORCE
(RHS)**

Dirichlet B.C.



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

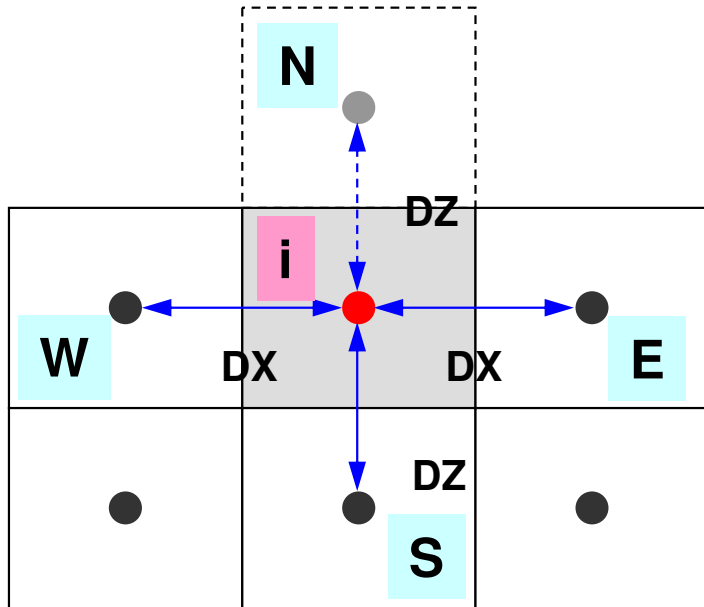
D (diagonal)

**AMAT
(off-diag.)**

**BFORCE
(RHS)**

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + \frac{\phi_N - \phi_i}{\Delta z} \Delta x \Delta y = -V_i \dot{Q}_i, \quad \phi_N = -\phi_i$$

Dirichlet B.C.



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

D (diagonal)

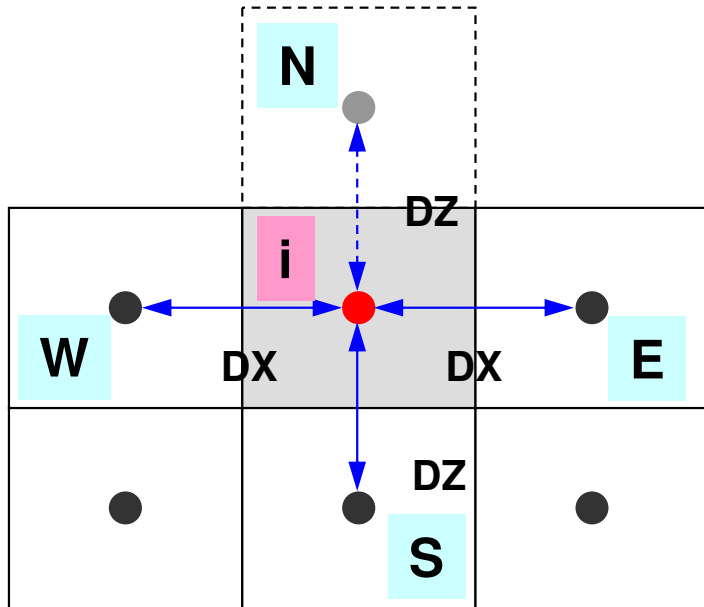
**AMAT
(off-diag.)**

**BFORCE
(RHS)**

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + \frac{\phi_N - \phi_i}{\Delta z} \Delta x \Delta y = -V_i \dot{Q}_i, \quad \phi_N = -\phi_i$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + \frac{-\phi_i - \phi_i}{\Delta z} \Delta x \Delta y = -V_i \dot{Q}_i$$

Dirichlet B.C.



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

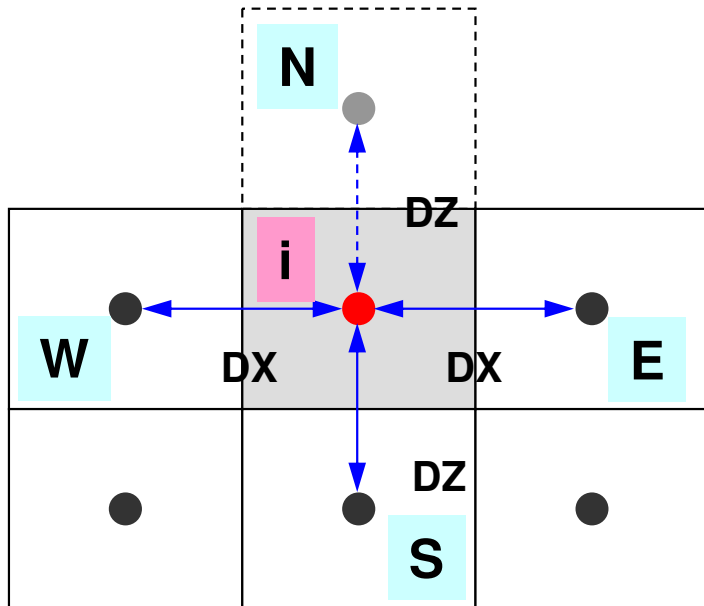
D (diagonal)

**AMAT
(off-diag.)**

**BFORCE
(RHS)**

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + \frac{-2\phi_i}{\Delta z} \Delta x \Delta y = +V_i \dot{Q}_i$$

Dirichlet B.C.



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

D (diagonal)

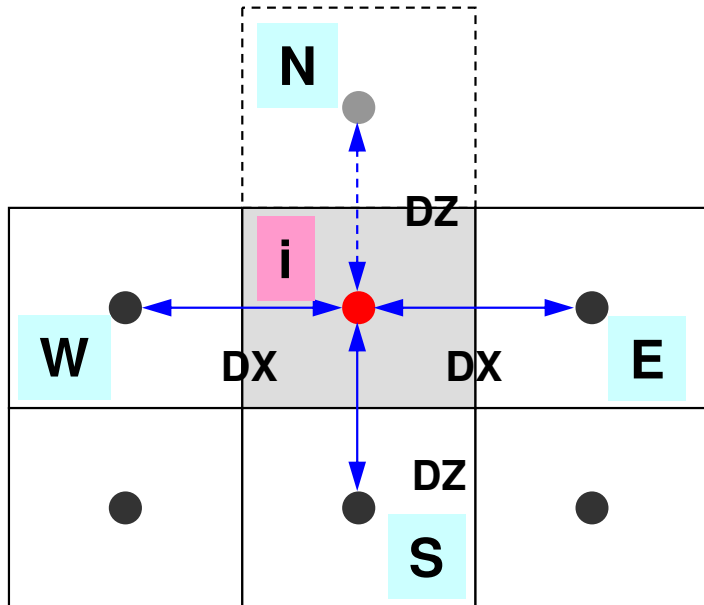
**AMAT
(off-diag.)**

**BFORCE
(RHS)**

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + \frac{-2\phi_i}{\Delta z} \Delta x \Delta y = +V_i \dot{Q}_i$$

$$\left[-\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} - \frac{2}{\Delta z} \Delta x \Delta y \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + = -V_i \dot{Q}_i$$

Dirichlet B.C.



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

D (diagonal)

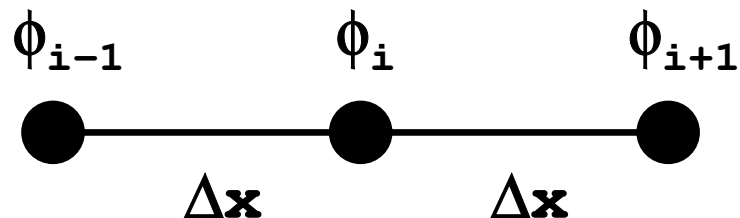
**AMAT
(off-diag.)**

**BFORCE
(RHS)**

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] - \left[-\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} - \frac{2}{\Delta z} \Delta x \Delta y \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + = -V_i \dot{Q}_i$$

```
for (ib=0; ib<ZmaxCELtot; ib++) {
    icel = ZmaxCEL[ib];
    coef = 2.0 * RDZ * ZAREA;
    D[icel-1] -= coef;
}
```

Taylor Series Expansion



$$\phi_{i+1} = \phi_i + \Delta x \left(\frac{\partial \phi}{\partial x} \right)_i + \frac{(\Delta x)^2}{2!} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i + \frac{(\Delta x)^3}{3!} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_i \dots$$

$$\phi_{i-1} = \phi_i - \Delta x \left(\frac{\partial \phi}{\partial x} \right)_i + \frac{(\Delta x)^2}{2!} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i - \frac{(\Delta x)^3}{3!} \left(\frac{\partial^3 \phi}{\partial x^3} \right)_i \dots$$

$$\phi_{i-1} + \phi_{i+1} = 2\phi_i + 2 \times \frac{(\Delta x)^2}{2!} \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i + 2 \times \frac{(\Delta x)^4}{4!} \left(\frac{\partial^4 \phi}{\partial x^4} \right)_i \dots$$

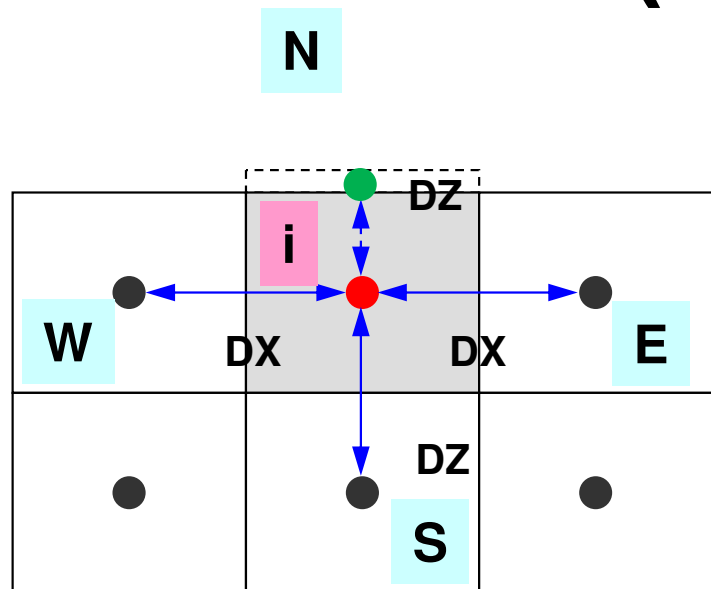
$$\frac{\phi_{i-1} - 2\phi_i + \phi_{i+1}}{(\Delta x)^2} = \left(\frac{\partial^2 \phi}{\partial x^2} \right)_i + \frac{(\Delta x)^2}{12} \left(\frac{\partial^4 \phi}{\partial x^4} \right)_i \dots$$

**Truncation Err.: 2nd Order
2nd Order Accuracy**

**If Δx is not uniform: 1st or
Lower Order Accuracy**

Dirichlet B.C. “N” is very thin ($=\varepsilon$)

1st order (or lower) Accuracy



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

D (diagonal)

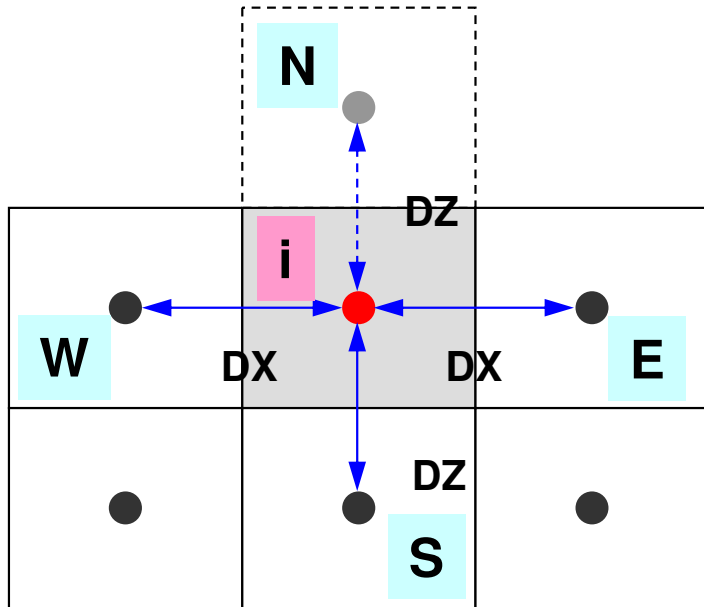
**AMAT
(off-diag.)**

**BFORCE
(RHS)**

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + \frac{\phi_N - \phi_i}{\left(\frac{\Delta z}{2} + \frac{\varepsilon}{2} \right)} \Delta x \Delta y = -V_i \dot{Q}_i, \quad \phi_N = 0, \quad \varepsilon \sim 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] - \frac{2\phi_i}{\Delta z} \Delta x \Delta y = -V_i \dot{Q}_i$$

Dirichlet B.C. using Mirror Image



$$\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} (\phi_k - \phi_i) + V_i \dot{Q}_i = 0$$

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] = -V_i \dot{Q}_i$$

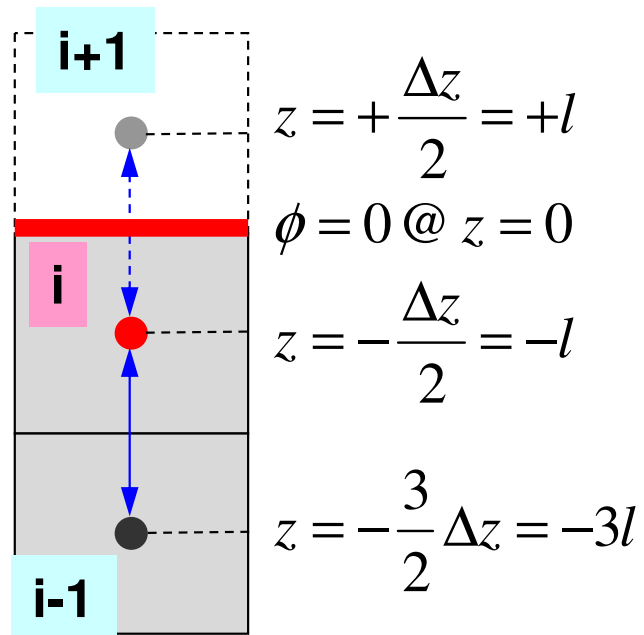
D (diagonal)

**AMAT
(off-diag.)**

**BFORCE
(RHS)**

$$-\left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + \frac{-2\phi_i}{\Delta z} \Delta x \Delta y = +V_i \dot{Q}_i$$

$$\left[-\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} - \frac{2}{\Delta z} \Delta x \Delta y \right] \phi_i + \left[\sum_k \frac{S_{ik}}{d_{ik} + d_{ki}} \phi_k \right] + = -V_i \dot{Q}_i$$



Higher Order Approximation for Dirichlet B.C. in 1D Problem

**more complicated in
2D/3D cases**

$$\phi = az^2 + bz + c$$

$$\phi(z = 0) = c = 0$$

$$\phi_i = al^2 - bl + c = al^2 - bl, \quad \phi_{i-1} = 9al^2 - 3bl + c = 9al^2 - 3bl$$

$$a = \frac{\phi_{i-1} - 3\phi_i}{6l^2}, \quad b = \frac{\phi_{i-1} - 9\phi_i}{6l} \Rightarrow \phi_{i+1} = al^2 + bl = \frac{1}{3}\phi_{i-1} - 2\phi_i$$

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <errno.h>
#include "struct.h"
#include "pcg.h"
#include "input.h" ...
```

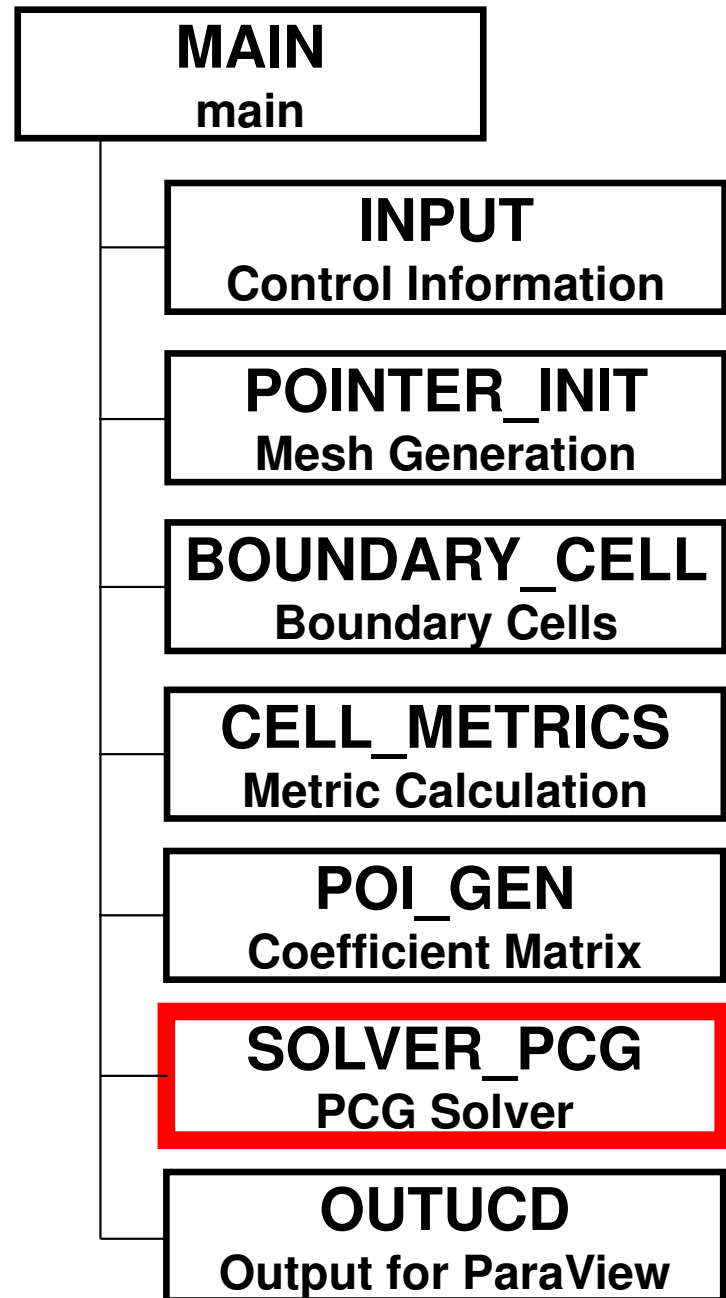
```
int
main()
{
    double *WK;
    int NPL, NPU; ISET, ITR, IER; icel, ic0, i;
    double xN, xL, xU; Stime, Etime;

    if(INPUT()) goto error;
    if(POINTER_INIT()) goto error;
    if(BOUNDARY_CELL()) goto error;
    if(CELL_METRICS()) goto error;
    if(POI_GEN()) goto error;

    memset(PHI, 0.0, sizeof(double)*ICELTOT);
    ISET = 0;
    WK = (double *)malloc(sizeof(double)*ICELTOT);
    if(solve_PCG(...)) goto error;

    if(OUTUCD()) goto error;
    return 0;
error:
    return -1;
}
```

Structure of the Program



- Background
 - Finite Volume Method
 - Preconditioned Iterative Solvers
- **PCG Solver for Poisson's Equations**
 - How to run
 - Data Structure
 - **Program**
 - Initialization
 - Coefficient Matrices
 - **PCG**

Solving Linear Equations

- Conjugate Gradient, CG
- Preconditioning
 - Point Jacobi, Diagonal Scaling
- PCG

$$[M] = \begin{bmatrix} D_1 & 0 & \dots & 0 & 0 \\ 0 & D_2 & & 0 & 0 \\ \dots & & \dots & & \dots \\ 0 & 0 & & D_{N-1} & 0 \\ 0 & 0 & \dots & 0 & D_N \end{bmatrix}$$

solve_PCG (1/6)

```
#include <stdio.h>
#include <stdlib.h>
#include <string.h>
#include <errno.h>
#include <math.h> etc.
```

```
#include "solver_PCG.h"
```

```
extern int
solve_PCG (int N, int *indexLU, int *itemLU,
           double *D, double *B, double *X, double *AMAT,
           double EPS, int *ITR, int *IER)
```

```
{
    double **W;
    double VAL, BNRM2, WVAL, SW, RHO, BETA, RHO1, C1, DNRM2;
    double ALPHA, ERR;
```

```
    int i, j, ic, ip, L, ip1;
```

```
    int R = 0;
    int Z = 1;
    int Q = 1;
    int P = 2;
    int DD = 3;
```

ICELTOT → N

BFORCE → B

PHI → X

EPSICCG → EPS

$W[0][i] = W[R][i] \Rightarrow \{r\}$

$W[1][i] = W[Z][i] \Rightarrow \{z\}$

$W[1][i] = W[Q][i] \Rightarrow \{q\}$

$W[2][i] = W[P][i] \Rightarrow \{p\}$

$W[3][i] = W[DD][i] \Rightarrow \{1/d\}$

solve_PCG (2/6)

```

W = (double **)malloc(sizeof(double *)*4);
if(W == NULL) {
    fprintf(stderr, "Error: %s\n", strerror(errno));
    return -1;
}

for(i=0; i<4; i++) {
    W[i] = (double *)malloc(sizeof(double)*N);
    if(W[i] == NULL) {
        fprintf(stderr, "Error: %s\n", strerror(errno));
        return -1;
    }
}

for(i=0; i<N; i++) {
    X[i] = 0.0;
    W[1][i] = 0.0;
    W[2][i] = 0.0;
    W[3][i] = 0.0;
}

for(i=0; i<N; i++) {
    W[DD][i] = 1.0 / D[i];
}

```

Reciprocal numbers (倒数) of diagonal components are stored in $W[DD][i]$. Computational cost for division is usually expensive.

Although it was said (division):(+,-,*) is 10:1 before, the difference is much smaller now. Generally, multiplying is still faster than division.

solve_PCG (3/6)

```

for (i=0; i<N; i++) {
    VAL = D[i] * X[i];
    for (j=indexLU[i]; j<indexLU[i+1]; j++)
    {
        VAL += AMAT[j] * X[itemLU[j]-1];
    }
    W[R][i] = B[i] - VAL;
}

```

```

BNRM2 = 0.0;
for (i=0; i<N; i++) {
    BNRM2 += B[i]*B[i];
}

```

Compute $\mathbf{r}^{(0)} = \mathbf{b} - [\mathbf{A}]\mathbf{x}^{(0)}$

```

for i= 1, 2, ...
    solve [M] z(i-1) = r(i-1)
    ρi-1 = r(i-1) z(i-1)
    if i=1
        p(1) = z(0)
    else
        βi-1 = ρi-1 / ρi-2
        p(i) = z(i-1) + βi-1 p(i-1)
    endif
    q(i) = [A] p(i)
    αi = ρi-1 / p(i) q(i)
    x(i) = x(i-1) + αi p(i)
    r(i) = r(i-1) - αi q(i)
    check convergence |r|
end

```

$\text{BNRM2} = \|\mathbf{b}\|^2$

to be used for convergence evaluation

```

*ITR = N;
for (L=0; L<(*ITR); L++) {
/*****
* {z} = [Minv]{r} *
*****/
    for (i=0; i<N; i++) {
        W[Z][i] = W[R][i]*W[DD][i];
    }
/*****
* RHO = {r}{z} *
*****/
    RHO = 0.0;
    for (i=0; i<N; i++) {
        RHO += W[R][i] * W[Z][i];
    }
/*****
* {p} = {z} if ITER=0 *
* BETA = RHO / RHO1 otherwise *
*****/
    if (L == 0) {
        for (i=0; i<N; i++) {
            W[P][i] = W[Z][i];
        }
    } else {
        BETA = RHO / RHO1;
        for (i=0; i<N; i++) {
            W[P][i] = W[Z][i] + BETA * W[P][i];
        }
    }
}

```

solve_PCG (4/6)

Compute $r^{(0)} = b - [A]x^{(0)}$
for $i = 1, 2, \dots$
solve $[M]z^{(i-1)} = r^{(i-1)}$
 $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$
if $i=1$
 $p^{(1)} = z^{(0)}$
else
 $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$
 $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$
endif
 $q^{(i)} = [A]p^{(i)}$
 $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$
 $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$
 $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$
 check convergence $|r|$
end

```

/*****
* {q} = [A] {p} *
*****/
for (i=0; i<N; i++) {
    VAL = D[i] * W[P][i];
    for (j=indexLU[i]; j<indexLU[i+1]; j++) {
        VAL += AMAT[j] * W[P][itemLU[j]];
    }
    W[Q][i] = VAL;
}

/*****
* ALPHA = RHO / {p} {q} *
*****/
C1 = 0.0;
for (i=0; i<N; i++) {
    C1 += W[P][i] * W[Q][i];
}
ALPHA = RHO / C1;

/*****
* {x} = {x} + ALPHA * {p} *
* {r} = {r} - ALPHA * {q} *
*****/
for (i=0; i<N; i++) {
    X[i] += ALPHA * W[P][i];
    W[R][i] -= ALPHA * W[Q][i];
}

```

solve_PCG

(5/6)

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for i = 1, 2, ...
    solve  $[M]z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    if i=1
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

solve_PCG (6/6)

```

DNRM2 = 0.0;
for(i=0; i<N; i++) {
    DNRM2 += W[R][i]*W[R][i];
}

ERR = sqrt(DNRM2/BNRM2);
if((L+1)%100 ==1) {
    fprintf(stderr, "%5d%16.6e\n", L+1, ERR);
}
if(ERR < EPS) {
    *IER = 0; goto N900;
} else {
    RH01 = RH0;
}
}
*IER = 1;

```

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
    solve  $[M]z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    if  $i=1$ 
         $p^{(1)} = z^{(0)}$ 
    else
         $\beta_{i-1} = \rho_{i-1} / \rho_{i-2}$ 
         $p^{(i)} = z^{(i-1)} + \beta_{i-1} p^{(i-1)}$ 
    endif
     $q^{(i)} = [A]p^{(i)}$ 
     $\alpha_i = \rho_{i-1} / p^{(i)} q^{(i)}$ 
     $x^{(i)} = x^{(i-1)} + \alpha_i p^{(i)}$ 
     $r^{(i)} = r^{(i-1)} - \alpha_i q^{(i)}$ 
    check convergence  $|r|$ 
end

```

solve_PCG (6/6)

```

DORM2 = 0.0;
for(i=0; i<N; i++) {
    DORM2 += W[R][i]*W[R][i];
}

ERR = sqrt(DORM2/BNORM2);
if((L+1)%100 ==1) {
    fprintf(stderr, "%5d%16.6e\n", L+1, ERR);
}
if(ERR < EPS) {
    *IER = 0; goto N900;
} else {
    RH01 = RH0;
}
}
*IER = 1;

```

$$ERR = \sqrt{\frac{DNorm2}{BNorm2}} = \frac{|r|}{|b|} = \frac{|b - Ax|}{|b|} \leq Eps$$

$$\begin{aligned}
 r &= b - [A]x \\
 DORM2 &= |r|^2 \\
 BNORM2 &= |b|^2 \\
 ERR &= |r| / |b|
 \end{aligned}$$

```

Compute r(0) = b - [A]x(0)
for i= 1, 2, ...
    solve [M] z(i-1) = r(i-1)
    ρi-1 = r(i-1) z(i-1)
    βi-1 = ρi-1 / ρi-2
    p(i) = z(i-1) + βi-1 p(i-1)
endif
q(i) = [A] p(i)
αi = ρi-1 / p(i) q(i)
x(i) = x(i-1) + αi p(i)
r(i) = r(i-1) - αi q(i)
check convergence |r|
end

```

$$|r|, |b| : 2 / L2 / Euclidean - norm \quad (\|r\|_2, \|b\|_2)$$

$$Ax = b \Rightarrow \alpha Ax = \alpha b$$

$$r = b - Ax \Rightarrow R = \alpha b - \alpha Ax = \alpha r$$

In next two weeks ...

- You will receive 2 e-mail's from me, if you have officially registered for this class
- Please check your mail-box
- Please contact me at nakajima(at)cc.u-tokyo.ac.jp if you HAVE NOT received those at noon May 10 (T).
 - Only officially registered students will receive this info.
- **Subject: Info-1**
 - **t80XYZ** **User ID for Odyssey Supercomputer**
- **Subject: Info-2**
 - **Initial Password with 8 characters**