

# **1D-FEM in Fortran: Steady State Heat Conduction**

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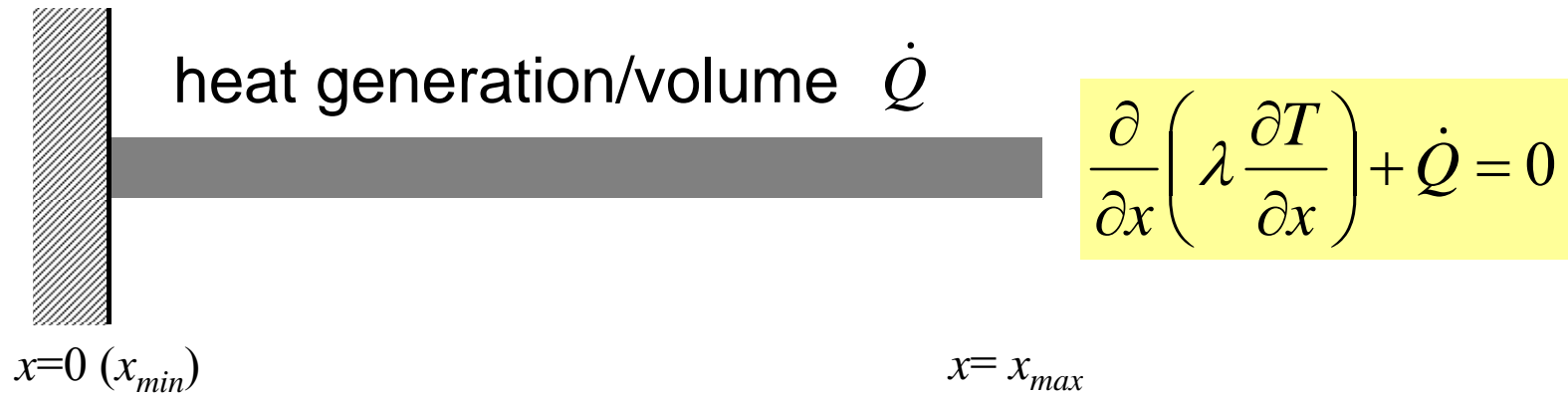
Programming for Parallel Computing (616-2057)  
Seminar on Advanced Computing (616-4009)

- 1D-code for Static Linear-Elastic Problems by Galerkin FEM
- Sparse Linear Solver
  - Conjugate Gradient Method
  - Preconditioning
- Storage of Sparse Matrices
- Program

# Keywords

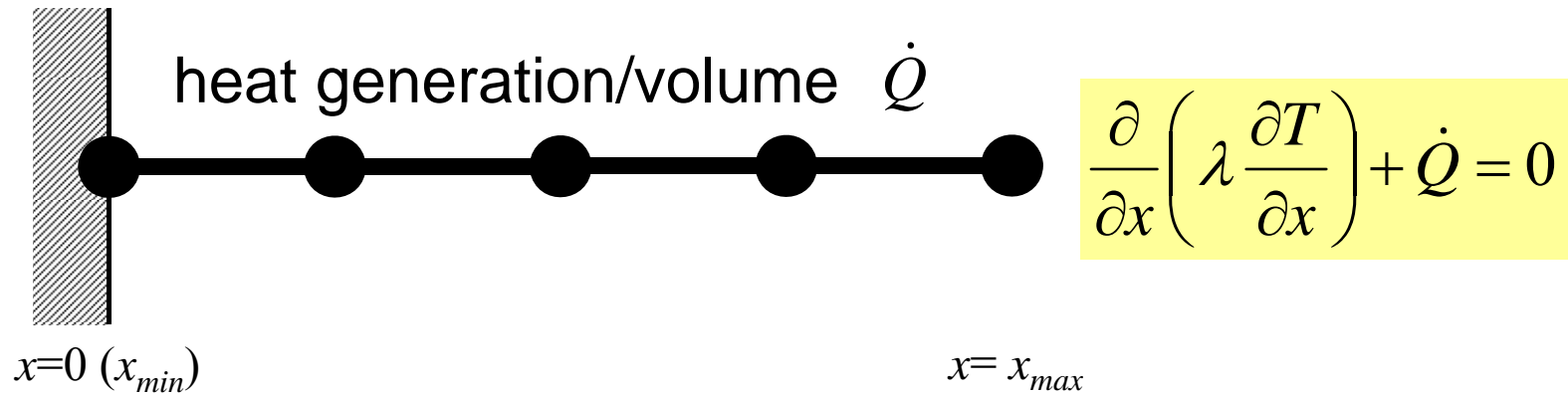
- 1D Steady State Heat Conduction Problems
- Galerkin Method
- Linear Element
- Preconditioned Conjugate Gradient Method

# 1D Steady State Heat Conduction



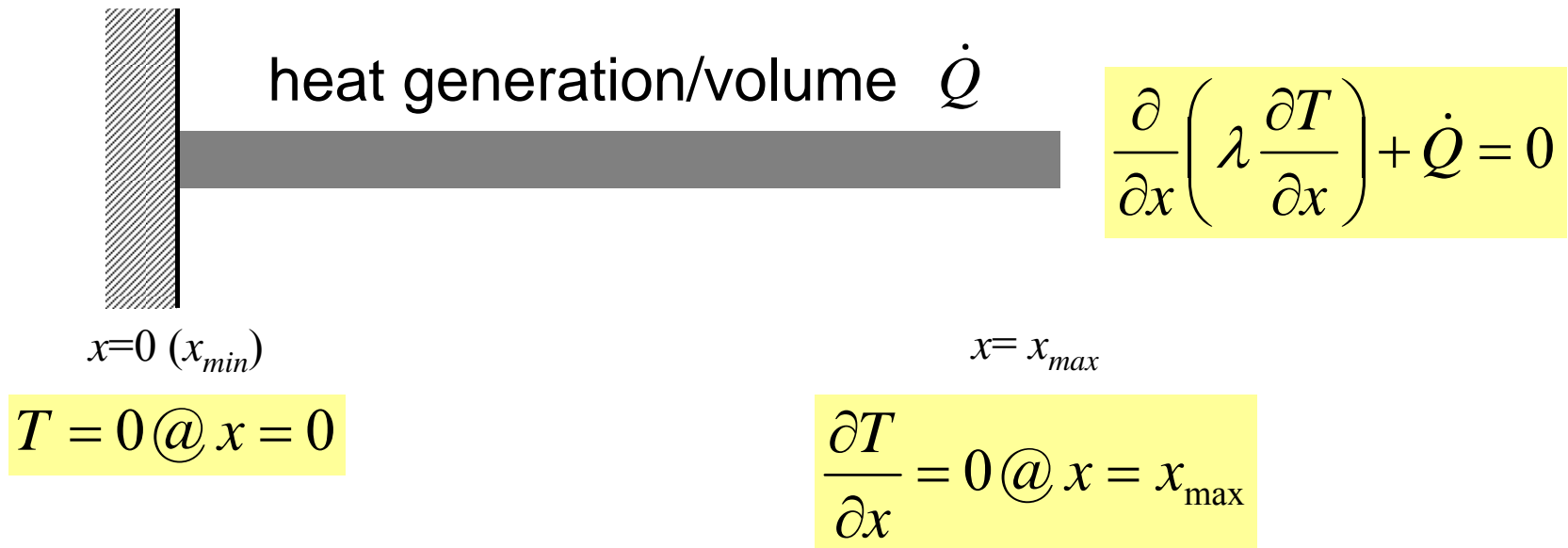
- **Uniform: Sectional Area:  $A$ , Thermal Conductivity:  $\lambda$**
- Heat Generation Rate/Volume/Time [ $QL^{-3}T^{-1}$ ]  $\dot{Q}$
- Boundary Conditions
  - $x=0$  :  $T=0$  (Fixed Temperature)
  - $x=x_{max}$  :  $\frac{\partial T}{\partial x} = 0$  (Insulated)

# 1D Steady State Heat Conduction



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# Analytical Solution



$$\lambda T'' = -\dot{Q}$$

$$\lambda T' = -\dot{Q}x + C_1 \Rightarrow C_1 = \dot{Q}x_{max}, \quad T' = 0 @ x = x_{max}$$

$$\lambda T = -\frac{1}{2}\dot{Q}x^2 + C_1x + C_2 \Rightarrow C_2 = 0, \quad T = 0 @ x = 0$$

$$\therefore T = -\frac{1}{2\lambda}\dot{Q}x^2 + \frac{\dot{Q}x_{max}}{\lambda}x$$

# 1D Linear Element (1/4)

## 一次元線形要素

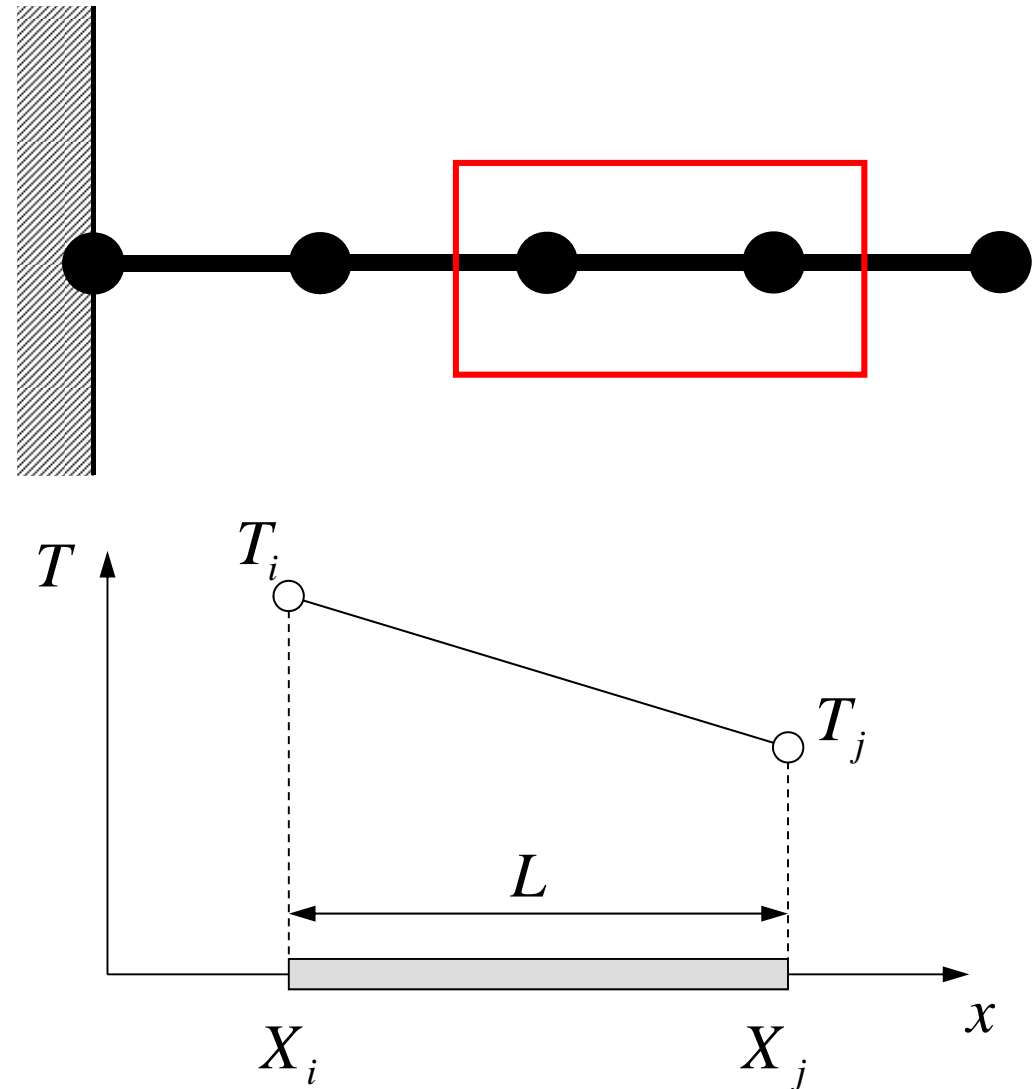
- 1D Linear Element

- Length =  $L$

- Node (Vertex)
- Element

- $T_i$  Temperature at  $i$
- $T_j$  Temperature at  $j$
- Temperature  $T$  on each element is linear function of  $x$  (Piecewise Linear):

$$T = \alpha_1 + \alpha_2 x$$



# 1D Linear Element (1/4)

## 一次元線形要素

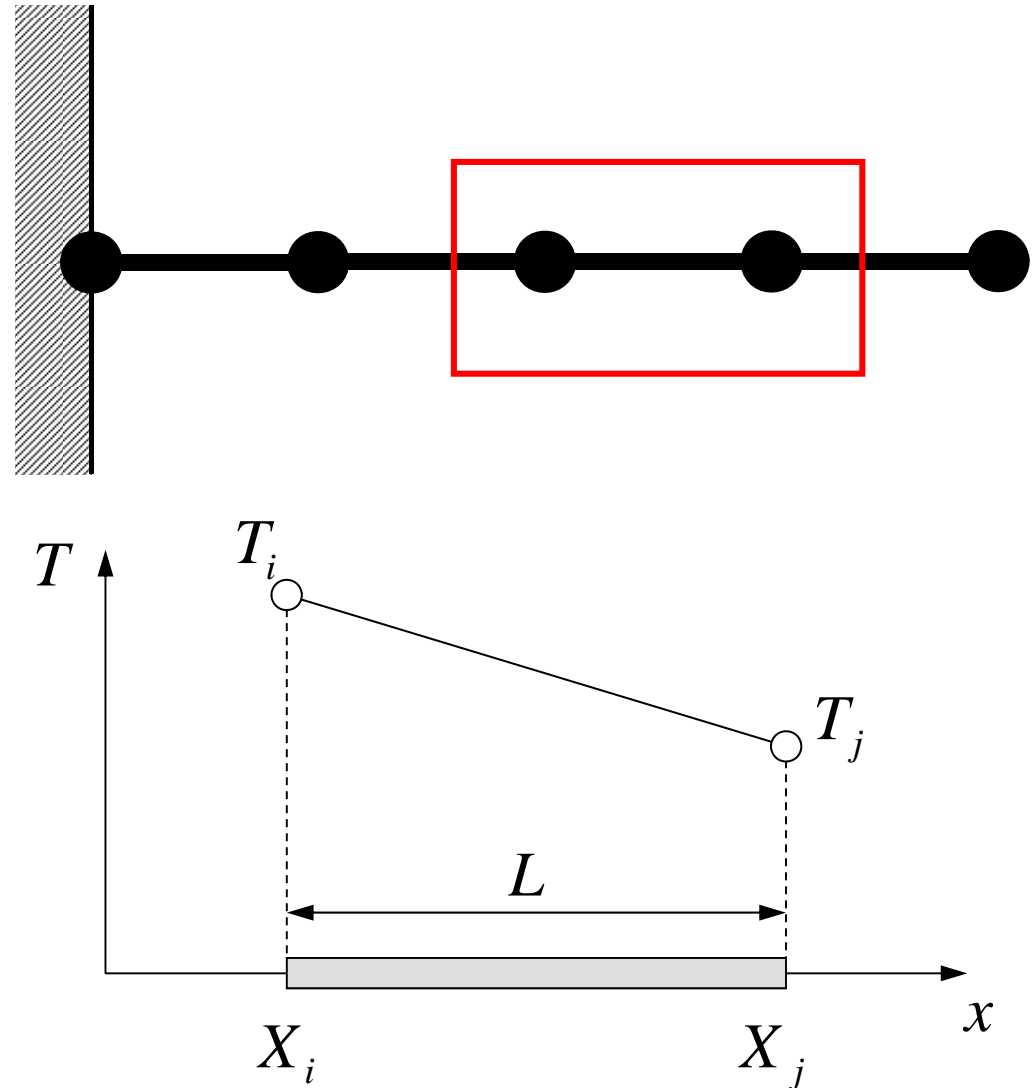
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- Node (Vertex)
- Element

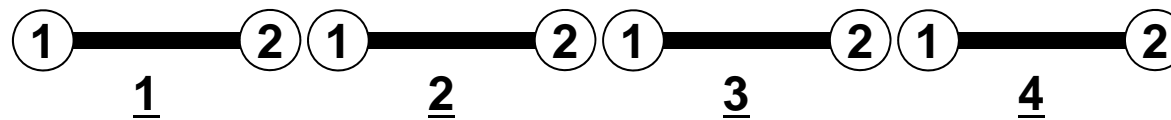
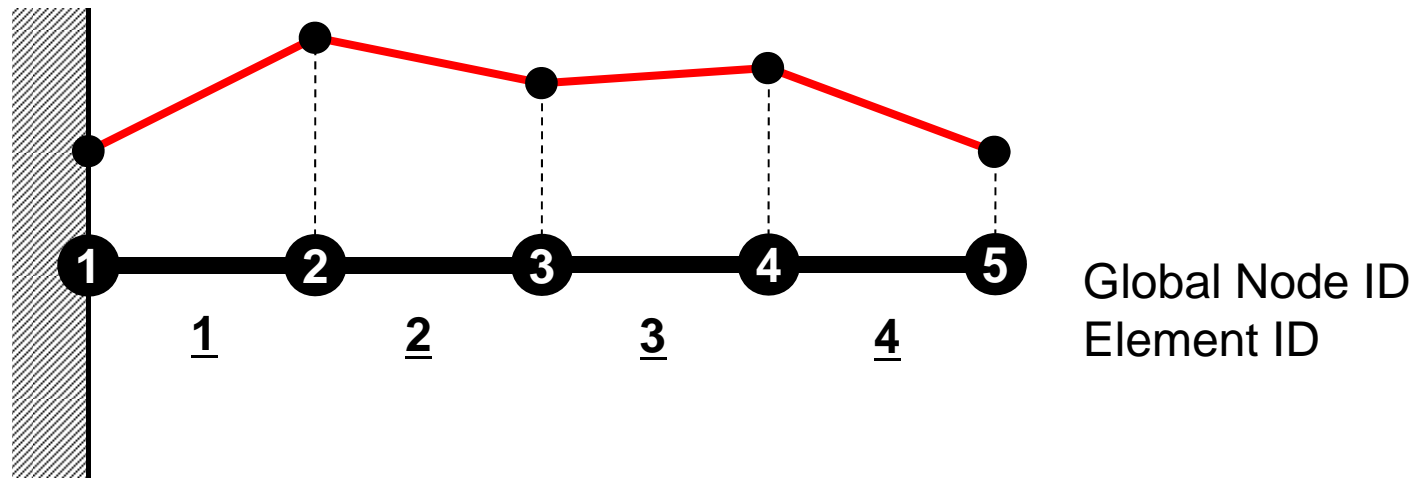
- $T_i$  Temperature at  $i$
- $T_j$  Temperature at  $j$
- Temperature  $T$  on each element is linear function of  $x$  (Piecewise Linear):

$$T = \alpha_1 + \alpha_2 x$$





# Piecewise Linear



Gradient of temperature is constant in each element (might be discontinuous at each “node”)

# 1D Linear Elem.: Shape Function (2/4)

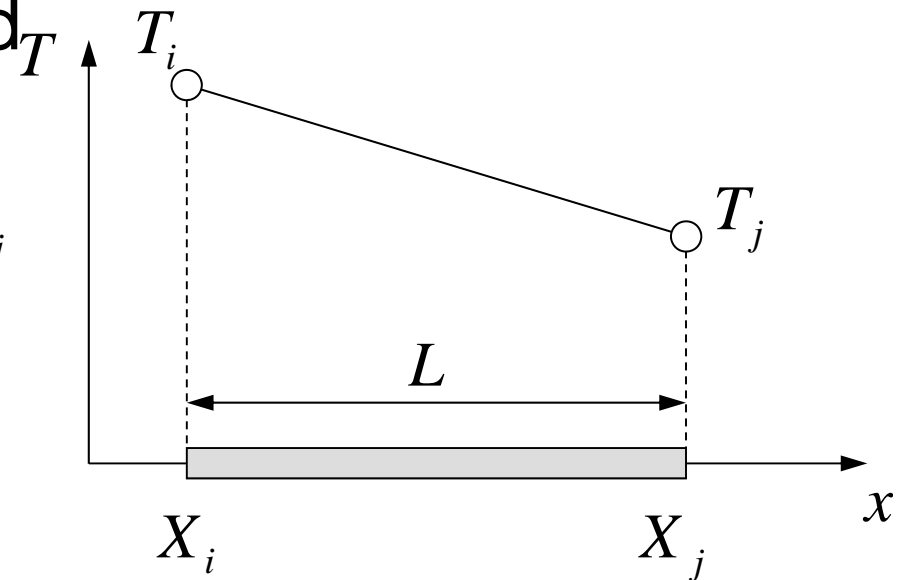
- Coef's are calculated based on info. at each node

$$T = T_i @ x = X_i, \quad T = T_j @ x = X_j$$

$$T_i = \alpha_1 + \alpha_2 X_i, \quad T_j = \alpha_1 + \alpha_2 X_j$$

- Coefficients:

$$\alpha_1 = \frac{T_i X_j - T_j X_i}{L}, \quad \alpha_2 = \frac{T_j - T_i}{L}$$



- $T$  can be written as follows, according to  $T_i$  and  $T_j$ :

$$T = \underbrace{\left( \frac{X_j - x}{L} \right)}_{N_i} T_i + \underbrace{\left( \frac{x - X_i}{L} \right)}_{N_j} T_j$$

$N_i, N_j$

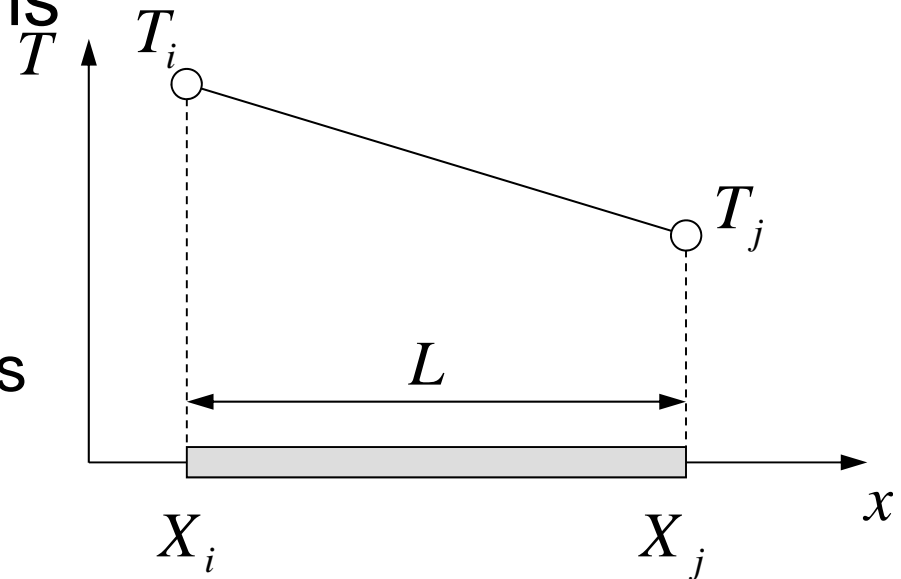
Shape Function or  
Interpolation Function  
function of  $x$  (only)

# 1D Linear Elem.: Shape Function (3/4)

- Number of Shape Functions = Number of Vertices of Each Element

- $N_i$ : Function of Position
- A kind of Test/Trial Functions

$$N_i = \left( \frac{X_j - x}{L} \right), \quad N_j = \left( \frac{x - X_i}{L} \right)$$



- Linear combination of shape functions provides displacement “in” each element
  - Coef’s (unknowns): Temperature at each node

$$T = N_i T_i + N_j T_j \longleftrightarrow$$

$$T_M = \sum_{i=1}^M a_i \Psi_i$$

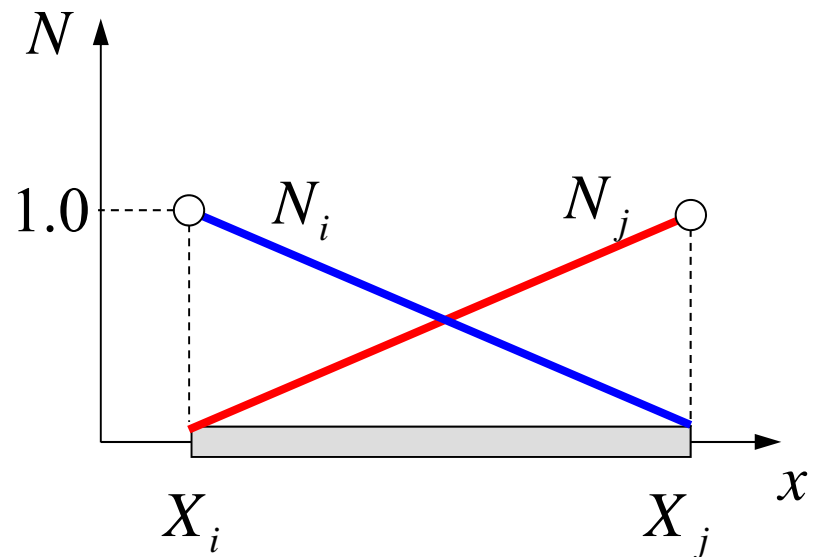
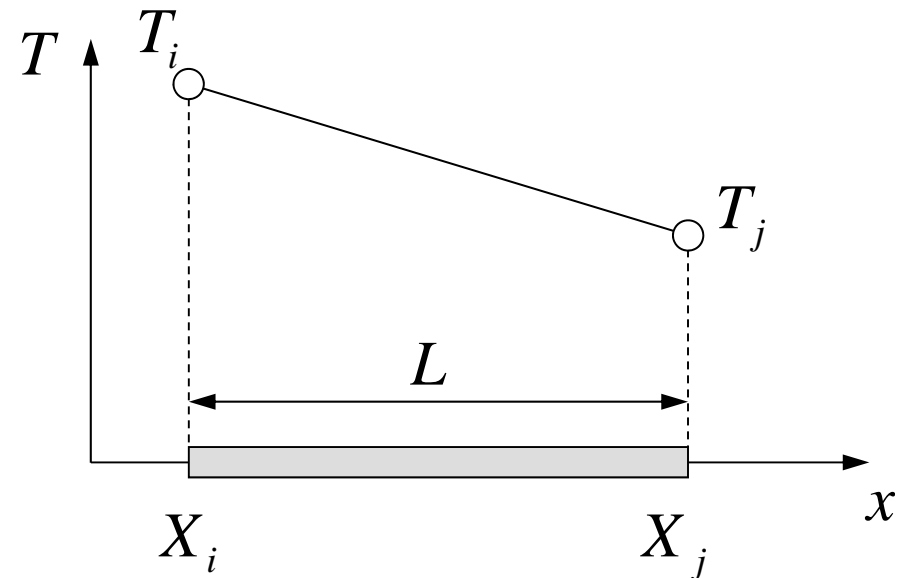
$\Psi_i$  Trial/Test Function (known function of position, defined in domain and at boundary. “Basis” in linear algebra.

$a_i$  Coefficients (unknown)

# 1D Linear Elem.: Shape Function (4/4)

- Value of  $N_i$ 
  - =1 at one of the nodes in element
  - =0 on other nodes

$$N_i = \left( \frac{X_j - x}{L} \right), \quad N_j = \left( \frac{x - X_i}{L} \right)$$



# Galerkin Method (1/4)

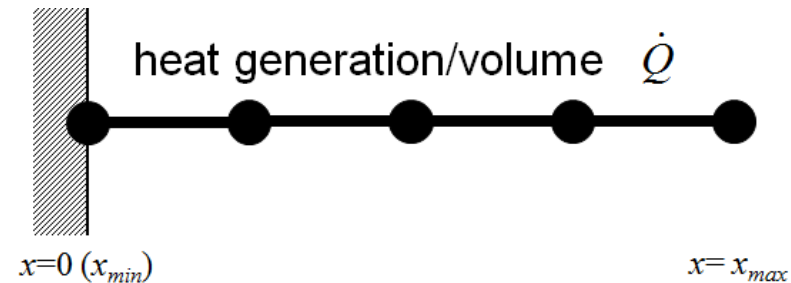
- Governing Equation for 1D Steady State Heat Conduction Problems (Uniform  $\lambda$ ):

$$\lambda \left( \frac{d^2 T}{dx^2} \right) + \dot{Q} = 0$$

$T = [N]\{\phi\}$  Distribution of temperature in each element (matrix form),  $\phi$ : Temperature at each node

- Following integral equation is obtained at each element by Galerkin method, where  $[N]$ 's are also weighting functions:

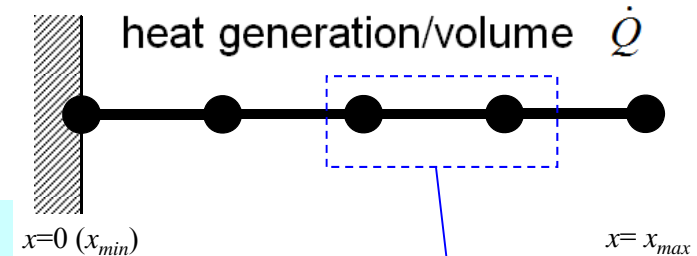
$$\int_V [N]^T \left\{ \lambda \left( \frac{d^2 T}{dx^2} \right) + \dot{Q} \right\} dV = 0$$



# Galerkin Method (2/4)

- Green's Theorem (1D)

$$\int_V A \left( \frac{d^2 B}{dx^2} \right) dV = \int_S A \frac{dB}{dx} dS - \int_V \left( \frac{dA}{dx} \frac{dB}{dx} \right) dV$$

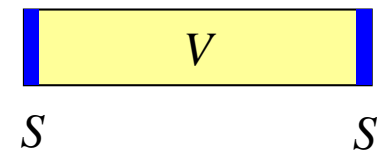


- Apply this to the 1st part of eqn with 2<sup>nd</sup>-order diff.:

$$\int_V \lambda [N]^T \left( \frac{d^2 T}{dx^2} \right) dV = - \int_V \lambda \left( \frac{d[N]^T}{dx} \frac{dT}{dx} \right) dV + \int_S \lambda [N]^T \frac{dT}{dx} dS$$

- Consider the following terms:

$$T = [N] \{ \phi \}, \quad \frac{dT}{dx} = \frac{d[N]}{dx} \{ \phi \} \quad \bar{q} = -\lambda \frac{dT}{dx}$$



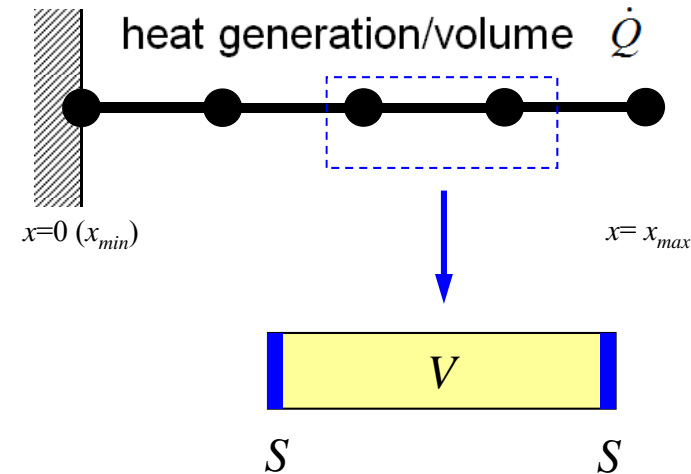
: Heat flux at element surface [QL<sup>-2</sup>T<sup>-1</sup>]

# Galerkin Method (3/4)

- Finally, following eqn is obtained by considering heat generation term  $\dot{Q}$  :

$$-\int_V \lambda \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV \cdot \{\phi\}$$

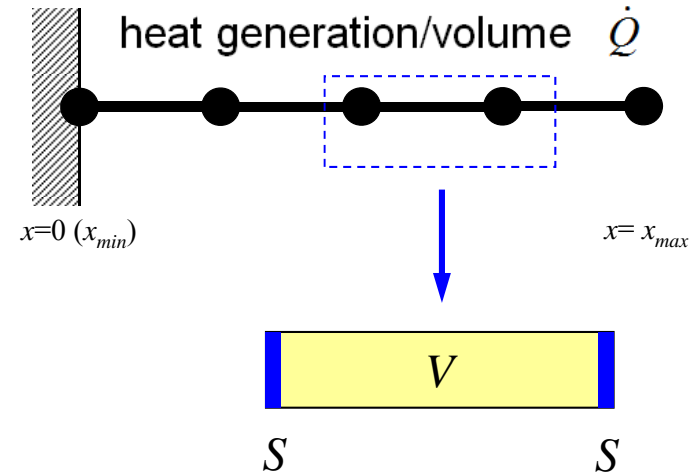
$$-\int_S \bar{q}[N]^T dS + \int_V \dot{Q}[N]^T dV = 0$$



- This is called “weak form (弱形式)”. Original PDE consists of terms with 2<sup>nd</sup>-order diff., but this “weak form” only includes 1<sup>st</sup>-order diff by Green’s theorem.
  - Requirements for shape functions are “weaker” in “weak form”. Linear functions can describe effects of 2<sup>nd</sup>-order differentiation.

# Galerkin Method (4/4)

$$\begin{aligned}
 & - \int_V \lambda \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV \cdot \{\phi\} \\
 & \boxed{- \int_S \bar{q} [N]^T dS} + \int_V \dot{Q} [N]^T dV = 0
 \end{aligned}$$

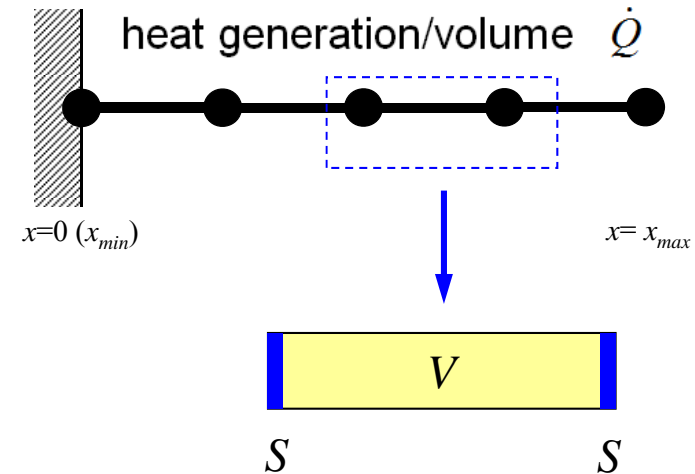


- These terms coincide at element boundaries and disappear. Finally, only terms on the domain boundaries remain.



# Weak Form and Boundary Conditions

- Value of dependent variable is defined (Dirichlet)
  - Weighting Function = 0
  - Principal B.C. (Boundary Condition) (第一種境界条件)
  - Essential B.C. (基本境界条件)
- Derivatives of Unknowns (Neumann)
  - Naturally satisfied in weak form
  - Secondary B.C. (第二種境界条件)
  - Natural B.C (自然境界条件)



$$-\int_V \lambda \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV \cdot \{\phi\}$$

$$-\int_S \bar{q} [N]^T dS + \int_V \dot{Q} [N]^T dV = 0$$

$$\text{where } \bar{q} = -\lambda \frac{dT}{dx}$$

Weak Form with B.C.: on each elem.

$$[k]^{(e)} \{\phi\}^{(e)} = \{f\}^{(e)}$$

$$[k]^{(e)} = \int_V \lambda \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV$$

$$\{f\}^{(e)} = \int_V \dot{Q} [N]^T dV - \int_S \bar{q} [N]^T dS$$

# Integration over Each Element: $[k]$

$$N_i = \left( \frac{X_j - x}{L} \right), \quad N_j = \left( \frac{x - X_i}{L} \right)$$

$$\frac{dN_i}{dx} = \left( \frac{-1}{L} \right), \quad \frac{dN_j}{dx} = \left( \frac{1}{L} \right)$$

$$\int_V \lambda \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV$$

$$= \lambda \int_0^L \begin{bmatrix} -1/L \\ 1/L \end{bmatrix} [-1/L, 1/L] A dx$$

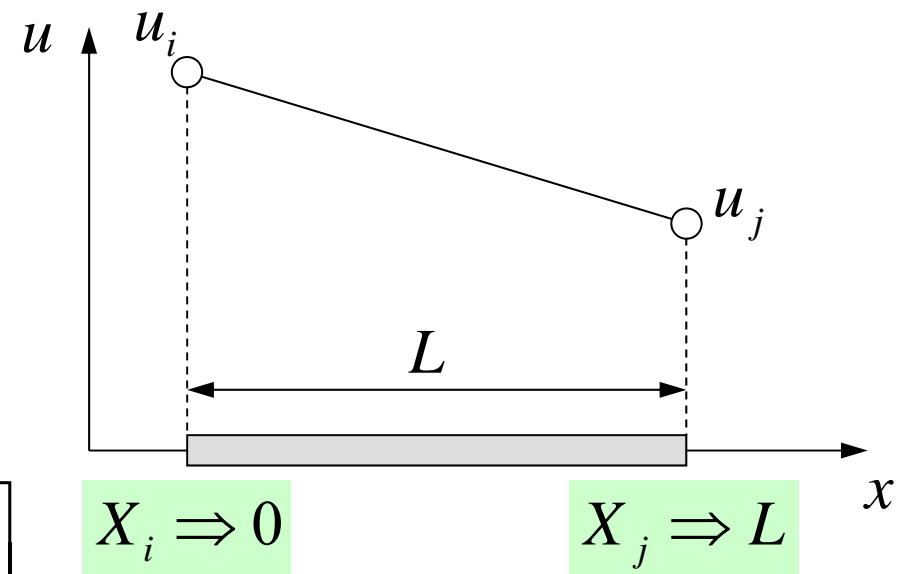
2x1 matrix

1x2 matrix

$$= \frac{\lambda A}{L^2} \int_0^L \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix} dx = \frac{\lambda A}{L} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix}$$

$A$ : Sectional Area

$L$ : Length



$$N_i = \left( 1 - \frac{x}{L} \right), \quad N_j = \left( \frac{x}{L} \right)$$

# Integration over Each Element: $\{f\}$ (1/2)

$$N_i = \left( \frac{X_j - x}{L} \right), \quad N_j = \left( \frac{x - X_i}{L} \right) \quad \frac{dN_i}{dx} = \left( \frac{-1}{L} \right), \quad \frac{dN_j}{dx} = \left( \frac{1}{L} \right)$$

$$N_i = \left( 1 - \frac{x}{L} \right), \quad N_j = \left( \frac{x}{L} \right)$$

$$\int_V \dot{Q} [N]^T dV = \dot{Q} A \int_0^L \begin{bmatrix} 1 - x/L \\ x/L \end{bmatrix} dx = \frac{\dot{Q} A L}{2} \begin{Bmatrix} 1 \\ 1 \end{Bmatrix}$$

Heat Generation  
(Volume)



A: Sectional Area

L: Length

# Integration over Each Element: $\{f\}$ (2/2)

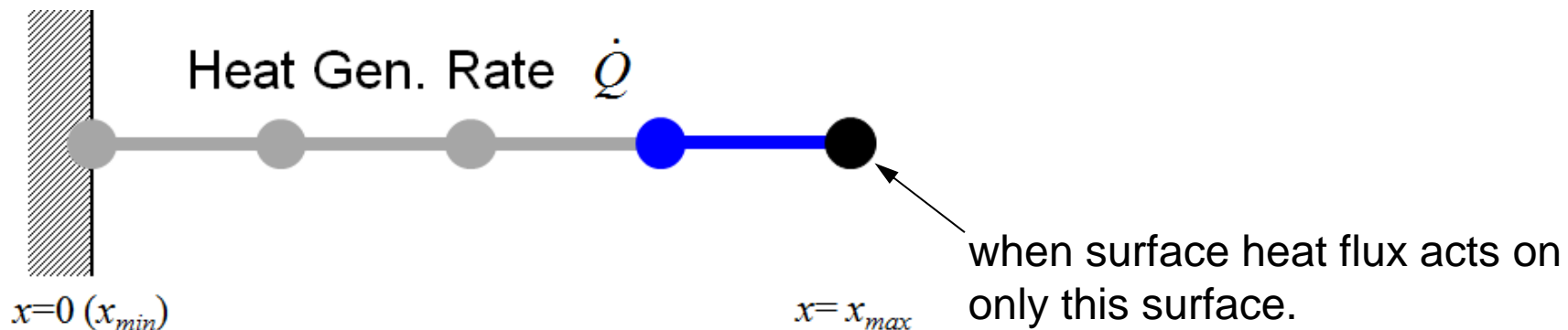
$$N_i = \left( \frac{X_j - x}{L} \right), \quad N_j = \left( \frac{x - X_i}{L} \right) \quad \frac{dN_i}{dx} = \left( \frac{-1}{L} \right), \quad \frac{dN_j}{dx} = \left( \frac{1}{L} \right)$$

$$\int_V \dot{Q} [N]^T dV = \dot{Q} A \int_0^L \begin{bmatrix} 1 - x/L \\ x/L \end{bmatrix} dx = \frac{\dot{Q} A L}{2} \begin{Bmatrix} 1 \\ 1 \end{Bmatrix}$$

Heat Generation  
(Volume)

$$\int_S \bar{q} [N]^T dS = \bar{q} A \Big|_{x=L} = \bar{q} A \begin{Bmatrix} 0 \\ 1 \end{Bmatrix}, \quad \bar{q} = -\lambda \frac{dT}{dx}$$

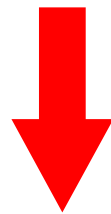
Surface Heat Flux



# Global Equations

- Accumulate Element Equations:

$$[k]^{(e)} \{\phi\}^{(e)} = \{f\}^{(e)} \quad \text{Element Matrix, Element Equations}$$



$$[K] \cdot \{\Phi\} = \{F\} \quad \text{Global Matrix, Global Equations}$$

$$[K] = \sum [k], \quad \{F\} = \sum \{f\}$$

$$\{\Phi\}: \text{global vector of } \{\phi\}$$

This is the final linear equations  
(global equations) to be solved.

# ECCS2012 System

## Creating Directory

```
>$ cd Documents  
>$ mkdir 2013summer your favorite name  
>$ cd 2013summer
```

This is your “top” directory, and is called **<\$E-TOP>** in this class.

## 1D Code for Steady-State Heat Conduction Problems

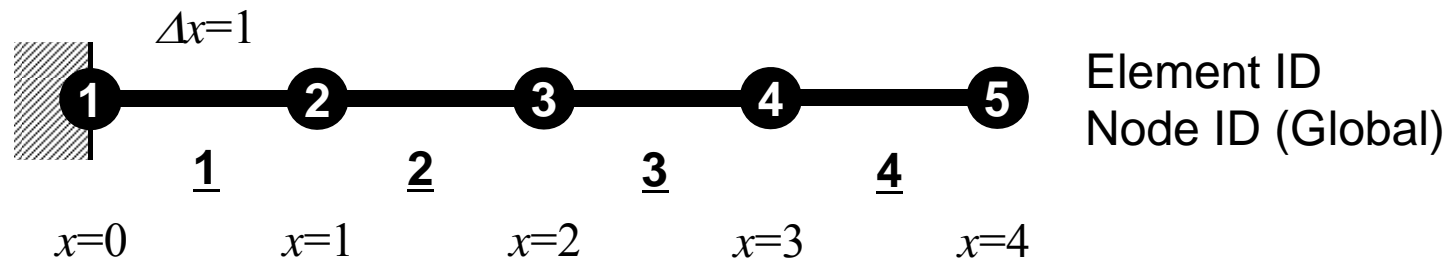
```
>$ cd <$E-TOP>  
>$ cp /home03/skengon/Documents/class_eps/F/1d.tar .  
>$ cp /home03/skengon/Documents/class_eps/C/1d.tar .  
>$ tar xvf 1d.tar  
>$ cd 1d
```

# Compile & GO !

```
>$ cd <${E-TOP}>/1d
>$ cc -O 1d.c          (or g95 -O 1d.f)
>$ ./a.out
```

## Control Data input.dat

4	NE (Number of Elements)
1.0 1.0 1.0 1.0	$\Delta x$ (Length of Each Elem.: $L$ ), $Q$ , $A$ , $\lambda$
100	Number of MAX. Iterations for CG Solver
1.e-8	Convergence Criteria for CG Solver





# Results

```
>$ ./a.out
```

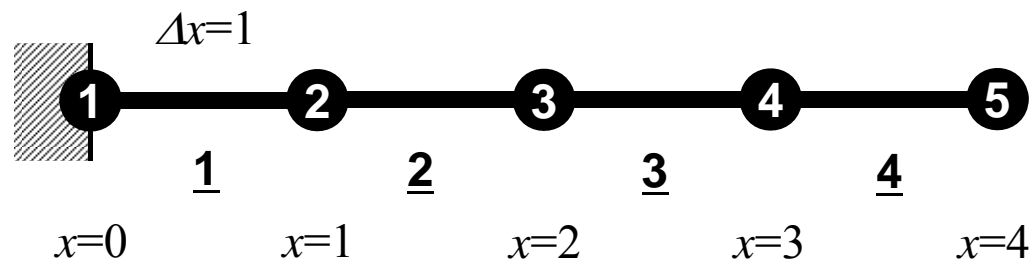
```
4 iters, RESID= 4.154074e-17
```

```
### TEMPERATURE
```

1	0.000000E+00	0.000000E+00
2	3.500000E+00	3.500000E+00
3	6.000000E+00	6.000000E+00
4	7.500000E+00	7.500000E+00
5	8.000000E+00	8.000000E+00

Computational

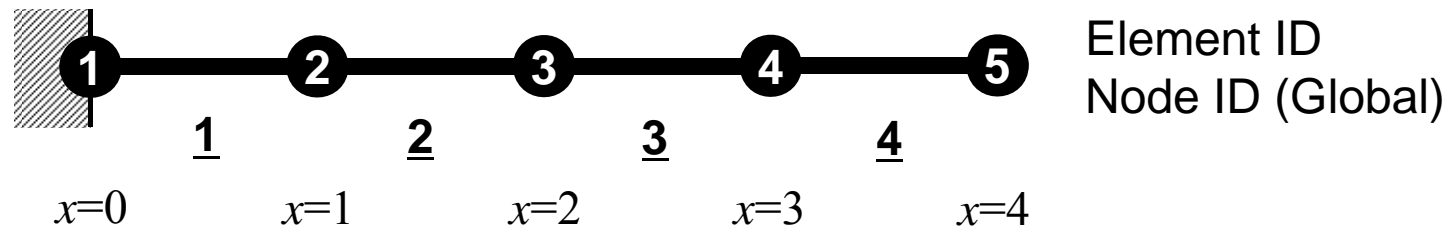
Analytical



Element ID  
Node ID (Global)

# Element Eqn's/Accumulation (1/3)

- 4 elements, 5 nodes



- $[k]$  and  $\{f\}$  of Element-1:

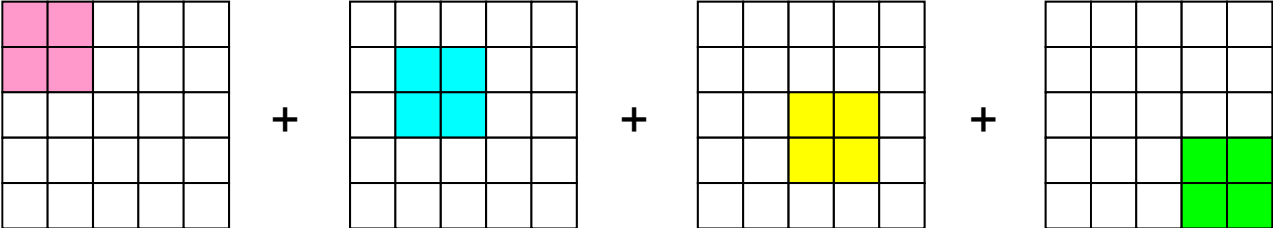
$$[k]^{(1)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix} \quad \{f\}^{(1)} = \frac{\dot{Q}AL}{2} \begin{Bmatrix} 1 \\ 1 \end{Bmatrix}$$

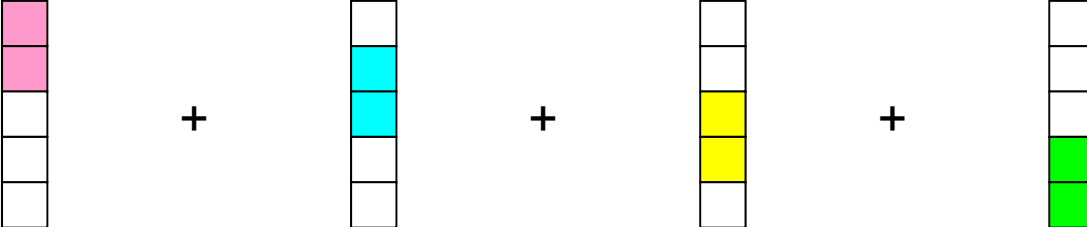
- As for Element-4:

$$[k]^{(4)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix} \quad \{f\}^{(4)} = \frac{\dot{Q}AL}{2} \begin{Bmatrix} 1 \\ 1 \end{Bmatrix}$$

# Element Eqn's/Accumulation (2/3)

- Element-by-Element Accumulation:

$$[K] = \sum_{e=1}^4 [k]^{(e)} =$$


$$\{F\} = \sum_{e=1}^4 \{f\}^{(e)} =$$


# Element Eqn's/Accumulation (3/3)

- Relations to FDM

$$[k]^{(e)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix}$$

$$[K] = \sum_{e=1}^4 [k]^{(e)} = \left[ \begin{array}{cccc} \begin{matrix} +1 & -1 \\ -1 & +1 \end{matrix} & & & \\ & \begin{matrix} +1 & -1 \\ -1 & +1 \end{matrix} & & \\ & & \begin{matrix} +1 & -1 \\ -1 & +1 \end{matrix} & \\ & & & \begin{matrix} +1 & -1 \\ -1 & +1 \end{matrix} \end{array} \right] \times \frac{\lambda A}{L}$$

$$= \begin{bmatrix} +1 & -1 & & & \\ -1 & +2 & -1 & & \\ & -1 & +2 & -1 & \\ & & -1 & +2 & -1 \\ & & & -1 & +1 \end{bmatrix} \times \frac{\lambda A}{L}$$

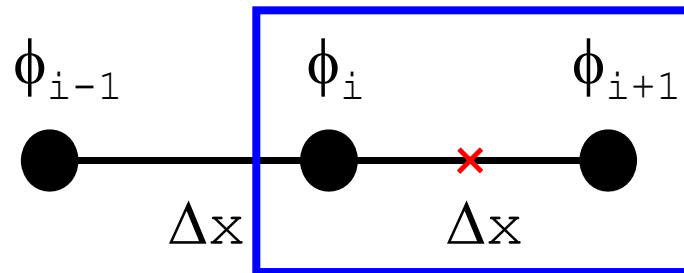
$$\begin{aligned} -\int_V \left( \frac{d^2 T}{dx^2} \right) dV &= -\int_V \left( \frac{T_{i+1} - 2T_i + T_{i-1}}{L^2} \right) dV \\ &= -\left( \frac{T_{i+1} - 2T_i + T_{i-1}}{L^2} \right) \cdot AL = -(T_{i+1} - 2T_i + T_{i-1}) \cdot \frac{A}{L} \end{aligned}$$

Something familiar ...

FEM: Coefficient Matrices are generally sparse  
(many ZERO's)

# 2<sup>nd</sup> –Order Differentiation in FDM

- Approximate Derivative at  $x$  (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  $i$

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

# Element-by-Element Operation

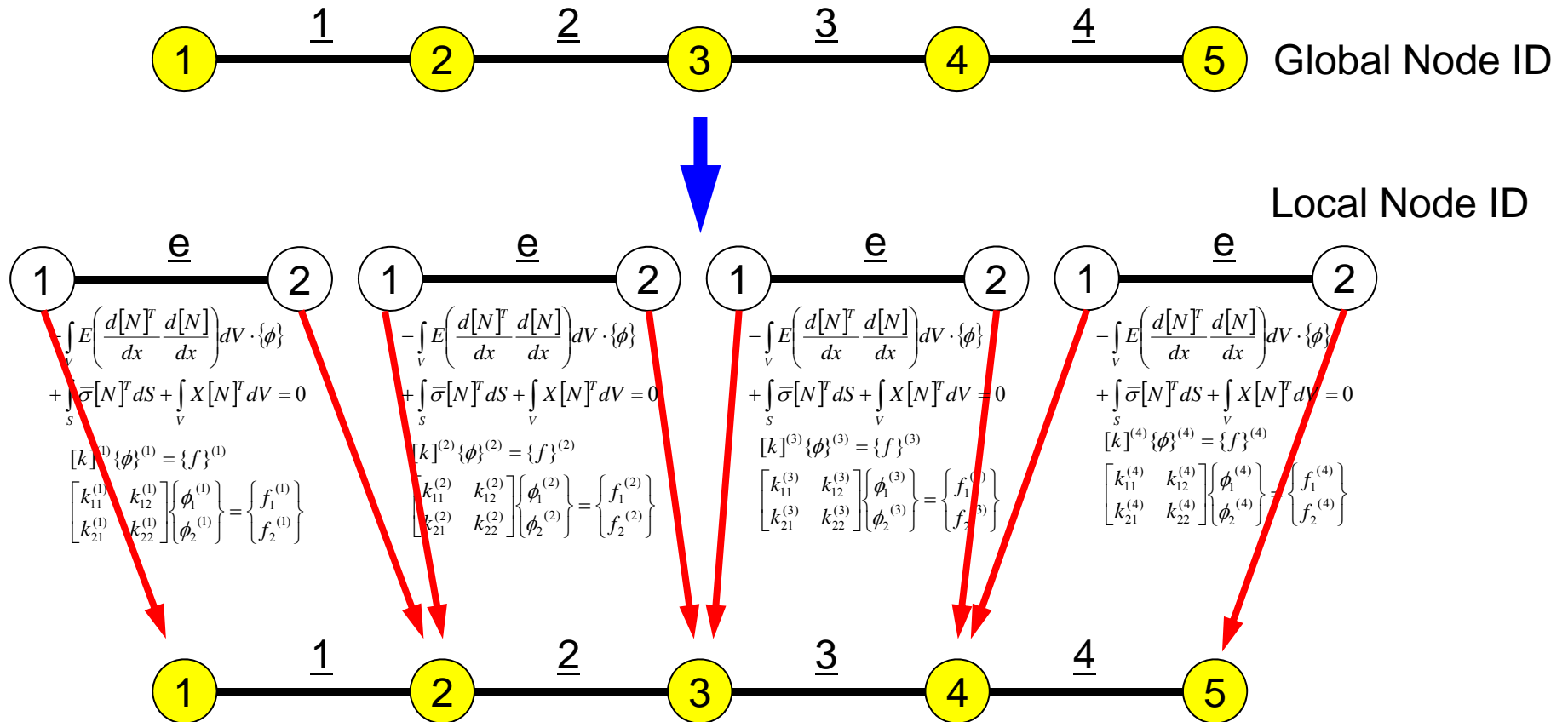
very flexible if each element has different material property, size, etc.

$$[k]^{(e)} = \frac{\lambda^{(e)} A^{(e)}}{L^{(e)}} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix}$$

$$[K] = \sum_{e=1}^4 [k^{(e)}] =$$

$$\begin{array}{c} \begin{array}{ccccc} +1 & -1 & & & \\ -1 & +1 & & & \\ & & & & \\ & & & & \\ & & & & \end{array} \times \frac{\lambda^{(1)} A^{(1)}}{L^{(1)}} + \begin{array}{ccccc} & & & & \\ & & +1 & -1 & \\ & & -1 & +1 & \\ & & & & \\ & & & & \end{array} \times \frac{\lambda^{(2)} A^{(2)}}{L^{(2)}} \\ \\ \begin{array}{ccccc} & & & & \\ & & & & \\ & & +1 & -1 & \\ & & -1 & +1 & \\ & & & & \\ & & & & \end{array} \times \frac{\lambda^{(3)} A^{(3)}}{L^{(3)}} + \begin{array}{ccccc} & & & & \\ & & & & \\ & & & & \\ & & & +1 & -1 \\ & & & -1 & +1 \end{array} \times \frac{\lambda^{(4)} A^{(4)}}{L^{(4)}} \end{array}$$

# Element/Global Operations



$$\begin{aligned}
 & - \int_V E \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV \cdot \{\phi\} \\
 & + \int_S \bar{\sigma} [N]^T dS + \int_V X [N]^T dV = 0 \\
 & [k]^{(1)} \{\phi\}^{(1)} = \{f\}^{(1)} \\
 & \begin{bmatrix} k_{11}^{(1)} & k_{12}^{(1)} \\ k_{21}^{(1)} & k_{22}^{(1)} \end{bmatrix} \begin{bmatrix} \phi_1^{(1)} \\ \phi_2^{(1)} \end{bmatrix} = \begin{bmatrix} f_1^{(1)} \\ f_2^{(1)} \end{bmatrix}
 \end{aligned}$$

$$\begin{aligned}
 & - \int_V E \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV \cdot \{\phi\} \\
 & + \int_S \bar{\sigma} [N]^T dS + \int_V X [N]^T dV = 0 \\
 & [k]^{(2)} \{\phi\}^{(2)} = \{f\}^{(2)} \\
 & \begin{bmatrix} k_{11}^{(2)} & k_{12}^{(2)} \\ k_{21}^{(2)} & k_{22}^{(2)} \end{bmatrix} \begin{bmatrix} \phi_1^{(2)} \\ \phi_2^{(2)} \end{bmatrix} = \begin{bmatrix} f_1^{(2)} \\ f_2^{(2)} \end{bmatrix}
 \end{aligned}$$

$$\begin{aligned}
 & - \int_V E \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV \cdot \{\phi\} \\
 & + \int_S \bar{\sigma} [N]^T dS + \int_V X [N]^T dV = 0 \\
 & [k]^{(3)} \{\phi\}^{(3)} = \{f\}^{(3)} \\
 & \begin{bmatrix} k_{11}^{(3)} & k_{12}^{(3)} \\ k_{21}^{(3)} & k_{22}^{(3)} \end{bmatrix} \begin{bmatrix} \phi_1^{(3)} \\ \phi_2^{(3)} \end{bmatrix} = \begin{bmatrix} f_1^{(3)} \\ f_2^{(3)} \end{bmatrix}
 \end{aligned}$$

$$\begin{aligned}
 & - \int_V E \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV \cdot \{\phi\} \\
 & + \int_S \bar{\sigma} [N]^T dS + \int_V X [N]^T dV = 0 \\
 & [k]^{(4)} \{\phi\}^{(4)} = \{f\}^{(4)} \\
 & \begin{bmatrix} k_{11}^{(4)} & k_{12}^{(4)} \\ k_{21}^{(4)} & k_{22}^{(4)} \end{bmatrix} \begin{bmatrix} \phi_1^{(4)} \\ \phi_2^{(4)} \end{bmatrix} = \begin{bmatrix} f_1^{(4)} \\ f_2^{(4)} \end{bmatrix}
 \end{aligned}$$

$$[K] \{\Phi\} = \{F\}$$

$$\begin{bmatrix} D_1 & AU_{11} & & & & \\ AL_{21} & D_2 & AU_{21} & & & \\ & & AL_{31} & D_3 & AU_{31} & \\ & & & & AL_{41} & D_4 & AU_{41} \\ & & & & & & AL_{51} & D_5 \end{bmatrix} \begin{bmatrix} \Phi_1 \\ \Phi_2 \\ \Phi_3 \\ \Phi_4 \\ \Phi_5 \end{bmatrix} = \begin{bmatrix} B_1 \\ B_2 \\ B_3 \\ B_4 \\ B_5 \end{bmatrix}$$

Mapping Information needed, from element-matrix to global-matrix.

# Accumulation to Global/overall Matrix

$[K]\{\Phi\} = \{F\}$

$D$	$X$			$X$	$X$										
$X$	$D$	$X$		$X$	$X$	$X$									
	$X$	$D$	$X$		$X$	$X$	$X$								
		$X$	$D$			$X$	$X$								
$X$	$X$			$D$	$X$			$X$	$X$						
$X$	$X$	$X$		$X$	$D$	$X$		$X$	$X$	$X$					
$X$	$X$	$X$		$X$	$D$	$X$		$X$	$X$	$X$					
								$X$	$X$		$D$	$X$			
				$X$	$X$			$X$	$D$	$X$					
				$X$	$X$	$X$		$X$	$D$	$X$					
						$X$	$X$			$X$	$D$				
								$X$	$X$		$D$	$X$			
										$X$	$X$				
												$D$	$X$		

$\left[ \begin{matrix} \Phi_1 \\ \Phi_2 \\ \Phi_3 \\ \Phi_4 \\ \Phi_5 \\ \Phi_6 \\ \Phi_7 \\ \Phi_8 \\ \Phi_9 \\ \Phi_{10} \\ \Phi_{11} \\ \Phi_{12} \\ \Phi_{13} \\ \Phi_{14} \\ \Phi_{15} \\ \Phi_{16} \end{matrix} \right] = \left[ \begin{matrix} F_1 \\ F_2 \\ F_3 \\ F_4 \\ F_5 \\ F_6 \\ F_7 \\ F_8 \\ F_9 \\ F_{10} \\ F_{11} \\ F_{12} \\ F_{13} \\ F_{14} \\ F_{15} \\ F_{16} \end{matrix} \right]$



- 1D-code for Static Linear-Elastic Problems by Galerkin FEM
- Sparse Linear Solver
  - Conjugate Gradient Method
  - Preconditioning
- Storage of Sparse Matrices
- Program

# 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**, FDM, DEM, MD (solid), BEM w/FMM

# Direct Method

## 直接法

- Gaussian Elimination/LU Factorization.
  - compute  $\mathbf{A}^{-1}$  directly.

### Good

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

### Bad

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

# Iterative Method

## 反復法

- Stationary Method
  - SOR, Gauss-Seidel, Jacobi
  - **Generally slow, impractical**
- 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 Parallel FEM

# 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: 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,  $\alpha_i$  : 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  $i = 1, 2, \dots$ 
   $z^{(i-1)} = r^{(i-1)}$ 
   $\rho_{i-1} = r^{(i-1)} \cdot z^{(i-1)}$ 
  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 (Double Precision:  $a\{X\} + \{Y\}$ )

$x^{(i)}$  : Vector

$\alpha_i$  : Scalar

# Procedures of Conjugate Gradient

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for i= 1, 2, ...
   $z^{(i-1)} = r^{(i-1)}$ 
   $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
  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

$\alpha_i$  : 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

$\alpha_i$  : 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
  - Double
  - $\{y\} = a\{x\} + \{y\}$

$x^{(i)}$  : Vector

$\alpha_i$  : 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

```

$x^{(i)}$  : Vector

$\alpha_i$  : Scalar

# Algorithm of CG Method (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  $f(x)$ :

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

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

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  $h$

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

# Algorithm of CG Method (2/5)

CG method minimizes  $f(x)$  at each iteration. Assume that approximate solution:  $x^{(0)}$ , and search direction vector  $p^{(k)}$  is defined at  $k$ -th iteration.

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

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

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

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

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

# Algorithm of CG Method (3/5)

Residual vector at  $(k+1)$ -th iteration:  $r^{(k+1)} = b - Ax^{(k+1)}$ ,  $r^{(k)} = b - Ax^{(k)}$

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

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

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

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

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

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

# Algorithm of CG Method (4/5)

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

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

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

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

Thus, following relation is obtained:

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



# Algorithm of CG Method (5/5)

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

$\left( p^{(k+1)}, Ap^{(k)} \right) = 0$   $p^{(k)}$  is “conjugate” for matrix A

Following “conjugate” 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 (1/2)

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

$$x^{(k+1)} = x^{(i+1)} + \sum_{j=i+1}^k \alpha_j p^{(j)}$$

$$r^{(k+1)} = b - Ax^{(k+1)} = b - A \left[ x^{(i+1)} + \sum_{j=i+1}^k \alpha_j p^{(j)} \right]$$

$$= [b - Ax^{(i+1)}] - \sum_{j=i+1}^k \alpha_j Ap^{(j)} = r^{(i+1)} - \sum_{j=i+1}^k \alpha_j Ap^{(j)}$$

$$(p^{(i)}, r^{(k+1)}) = \left( p^{(i)}, r^{(i+1)} - \sum_{j=i+1}^k \alpha_j Ap^{(j)} \right)$$

$$= \underbrace{(p^{(i)}, r^{(i+1)})}_{=0} - \underbrace{\left( p^{(i)}, \sum_{j=i+1}^k \alpha_j Ap^{(j)} \right)}_{=0} = 0$$

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

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

# Proof (2/2)

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

$$\begin{aligned} 0 &= \left( p^{(i)}, r^{(k+1)} \right) = \left( r^{(i)} + \beta_{i-1} p^{(i-1)}, r^{(k+1)} \right) \\ &= \left( \beta_{i-1} p^{(i-1)}, r^{(k+1)} \right) + \left( r^{(i)}, r^{(k+1)} \right) = \left( r^{(i)}, r^{(k+1)} \right) \end{aligned}$$

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

$$\begin{aligned} \left( p^{(k)}, r^{(k)} \right) &= \left( r^{(k)} + \beta_{k-1} p^{(k-1)}, r^{(k)} \right) \\ &= \left( \beta_{k-1} p^{(k-1)}, r^{(k)} \right) + \left( r^{(k)}, r^{(k)} \right) = \left( r^{(k)}, r^{(k)} \right) \end{aligned}$$

$$\alpha_k, \beta_k$$

Usually, we use simpler definitions of  $\alpha_k, \beta_k$  as follows:

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

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

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

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

# 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

```

$x^{(i)}$  : Vector

$\alpha_i$  : Scalar

# Preconditioning for Iterative Solvers

- Convergence rate of iterative solvers strongly depends on the spectral properties (eigenvalue distribution) of the coefficient matrix  $\mathbf{A}$ .
  - Eigenvalue distribution is small, eigenvalues are close to 1
  - In "ill-conditioned" problems, "condition number" (ratio of max/min eigenvalue if  $\mathbf{A}$  is symmetric) is large.
- A preconditioner  $\mathbf{M}$  (whose properties are similar to those of  $\mathbf{A}$ ) transforms the linear system into one with more favorable spectral properties
  - In "ill-conditioned" problems, "condition number" (ratio of max/min eigenvalue if  $\mathbf{A}$  is symmetric) is large.
  - $\mathbf{M}$  transforms original equation  $\mathbf{Ax}=\mathbf{b}$  into  $\mathbf{A}'\mathbf{x}=\mathbf{b}'$  where  $\mathbf{A}'=\mathbf{M}^{-1}\mathbf{A}$ ,  $\mathbf{b}'=\mathbf{M}^{-1}\mathbf{b}$
  - If  $\mathbf{M}\sim\mathbf{A}$ ,  $\mathbf{M}^{-1}\mathbf{A}$  is close to identity matrix.
  - If  $\mathbf{M}^{-1}=\mathbf{A}^{-1}$ , this is the best preconditioner (a.k.a. Gaussian Elimination)

# Preconditioned CG Solver

```

Compute  $r^{(0)} = b - [A]x^{(0)}$ 
for  $i = 1, 2, \dots$ 
    solve  $[M]z^{(i-1)} = r^{(i-1)}$ 
     $\rho_{i-1} = r^{(i-1)} z^{(i-1)}$ 
    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

```

# 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



# 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  $[M]z^{(i-1)} = r^{(i-1)}$**  is very easy.
- Provides fast convergence for simple problems.
- 1d.f, 1d.c

- More detailed discussions on preconditioning will be provided in “Multicore Programming”.

- 1D-code for Static Linear-Elastic Problems by Galerkin FEM
- Sparse Linear Solver
  - Conjugate Gradient Method
  - Preconditioning
- **Storage of Sparse Matrices**
- Program



# Variables/Arrays in 1d.f, 1d.c related to coefficient matrix

name	type	size	description
<b>N</b>	I	-	# Unknowns
<b>NPLU</b>	I	-	# Non-Zero Off-Diagonal Components
<b>Diag(:)</b>	R	N	Diagonal Components
<b>U(:)</b>	R	N	Unknown Vector
<b>Rhs(:)</b>	R	N	RHS Vector
<b>Index(:)</b>	I	0:N N+1	Off-Diagonal Components (Number of Non-Zero Off-Diagonals at Each ROW)
<b>Item(:)</b>	I	NPLU	Off-Diagonal Components (Corresponding Column ID)
<b>AMat(:)</b>	R	NPLU	Off-Diagonal Components (Value)

Only non-zero components are stored according to “Compressed Row Storage”.

# Mat-Vec. Multiplication for Sparse Matrix

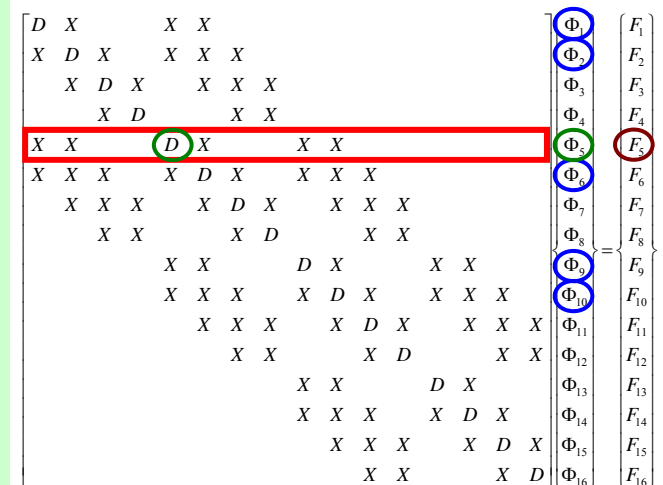
## Compressed Row Storage (CRS)

**Diag (i)** Diagonal Components (REAL, i=1~N)  
**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=1, index(N))  
**AMat(k)** Off-Diagonal Components (Value)  
 ( REAL, k=1, index(N) )

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

```

do i= 1, N
  Y(i)= Diag(i)*X(i)
  do k= Index(i-1)+1, Index(i)
    Y(i)= Y(i) + Amat(k)*X(Item(k))
  enddo
enddo
  
```



# CRS or CSR ? for Compressed Row Storage

- In Japan and USA, “CRS” is very general for abbreviation of “Compressed Row Storage”, but they usually use “CSR” in Europe (especially in France).
- “CRS” in France
  - Compagnie Républicaine de Sécurité
    - Republic Security Company of France
- French scientists may feel uncomfortable when we use “CRS” in technical papers and/or presentations.



# 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

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

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

```

do j= 1, N
  Y(j) = 0. d0
  do i= 1, N
    Y(j) = Y(j) + A(i, j)*X(i)
  enddo
enddo

```

# 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): Fortran

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

N= 8

对角成分

Diag(1) = 1.1  
 Diag(2) = 3.6  
 Diag(3) = 5.7  
 Diag(4) = 9.8  
 Diag(5) = 11.5  
 Diag(6) = 12.4  
 Diag(7) = 23.1  
 Diag(8) = 51.3

# Compressed Row Storage (CRS)

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

# Compressed Row Storage (CRS)

		# Non-Zero Off-Diag.						
1	<table border="1"> <tr> <td>1.1 ①</td> <td>2.4 ②</td> <td>3.2 ⑤</td> <td></td> <td></td> </tr> </table>	1.1 ①	2.4 ②	3.2 ⑤			2	$index(0) = 0$
1.1 ①	2.4 ②	3.2 ⑤						
2	<table border="1"> <tr> <td>3.6 ②</td> <td>4.3 ①</td> <td>2.5 ④</td> <td>3.7 ⑥</td> <td>9.1 ⑧</td> </tr> </table>	3.6 ②	4.3 ①	2.5 ④	3.7 ⑥	9.1 ⑧	4	$index(1) = 2$
3.6 ②	4.3 ①	2.5 ④	3.7 ⑥	9.1 ⑧				
3	<table border="1"> <tr> <td>5.7 ③</td> <td>1.5 ⑤</td> <td>3.1 ⑦</td> <td></td> <td></td> </tr> </table>	5.7 ③	1.5 ⑤	3.1 ⑦			2	$index(2) = 6$
5.7 ③	1.5 ⑤	3.1 ⑦						
4	<table border="1"> <tr> <td>9.8 ④</td> <td>4.1 ②</td> <td>2.5 ⑤</td> <td>2.7 ⑥</td> <td></td> </tr> </table>	9.8 ④	4.1 ②	2.5 ⑤	2.7 ⑥		3	$index(3) = 8$
9.8 ④	4.1 ②	2.5 ⑤	2.7 ⑥					
5	<table border="1"> <tr> <td>11.5 ⑤</td> <td>3.1 ①</td> <td>9.5 ②</td> <td>10.4 ③</td> <td>4.3 ⑦</td> </tr> </table>	11.5 ⑤	3.1 ①	9.5 ②	10.4 ③	4.3 ⑦	4	$index(4) = 11$
11.5 ⑤	3.1 ①	9.5 ②	10.4 ③	4.3 ⑦				
6	<table border="1"> <tr> <td>12.4 ⑥</td> <td>6.5 ③</td> <td>9.5 ⑦</td> <td></td> <td></td> </tr> </table>	12.4 ⑥	6.5 ③	9.5 ⑦			2	$index(5) = 15$
12.4 ⑥	6.5 ③	9.5 ⑦						
7	<table border="1"> <tr> <td>23.1 ⑦</td> <td>6.4 ②</td> <td>2.5 ③</td> <td>1.4 ⑥</td> <td>13.1 ⑧</td> </tr> </table>	23.1 ⑦	6.4 ②	2.5 ③	1.4 ⑥	13.1 ⑧	4	$index(6) = 17$
23.1 ⑦	6.4 ②	2.5 ③	1.4 ⑥	13.1 ⑧				
8	<table border="1"> <tr> <td>51.3 ⑧</td> <td>9.5 ②</td> <td>1.3 ③</td> <td>9.6 ④</td> <td>3.1 ⑥</td> </tr> </table>	51.3 ⑧	9.5 ②	1.3 ③	9.6 ④	3.1 ⑥	4	$index(7) = 21$
51.3 ⑧	9.5 ②	1.3 ③	9.6 ④	3.1 ⑥				
					4	$index(8) = 25$		

**NPLU= 25**  
**(=index(N))**

$index(i-1)+1^{th} \sim index(i)^{th}$   
Non-Zero Off-Diag. Components corresponding to  $i$ -th row

# Compressed Row Storage (CRS)

		# Non-Zero Off-Diag.	index (i) =
1	1.1 ①	2 ②,1 ⑤,2	0
2	3.6 ②	4 ①,3 ④,4 ⑥,5 ⑧,6	2
3	5.7 ③	2 ⑤,7 ⑦,8	6
4	9.8 ④	3 ②,9 ⑤,10 ⑥,11	8
5	11.5 ⑤	4 ①,12 ②,13 ③,14 ⑦,15	11
6	12.4 ⑥	2 ③,16 ⑦,17	15
7	23.1 ⑦	4 ②,18 ③,19 ⑥,20 ⑧,21	17
8	51.3 ⑧	4 ②,22 ③,23 ④,24 ⑥,25	21
			25

NPLU= 25  
(=index(N))

$index(i-1)+1^{th} \sim index(i)^{th}$   
Non-Zero Off-Diag. Components corresponding to  $i$ -th row

# Compressed Row Storage (CRS)

1	1.1 ①	2.4 ②,1	3.2 ⑤,2		
2	3.6 ②	4.3 ①,3	2.5 ④,4	3.7 ⑥,5	9.1 ⑧,6
3	5.7 ③	1.5 ⑤,7	3.1 ⑦,8		
4	9.8 ④	4.1 ②,9	2.5 ⑤,10	2.7 ⑥,11	
5	11.5 ⑤	3.1 ①,12	9.5 ②,13	10.4 ③,14	4.3 ⑦,15
6	12.4 ⑥	6.5 ③,16	9.5 ⑦,17		
7	23.1 ⑦	6.4 ②,18	2.5 ③,19	1.4 ⑥,20	13.1 ⑧,21
8	51.3 ⑧	9.5 ②,22	1.3 ③,23	9.6 ④,24	3.1 ⑥,25

Example:

`item( 7) = 5, AMAT( 7) = 1.5`

`item(19) = 3, AMAT(19) = 2.5`

# Compressed Row Storage (CRS)

1	1.1 ①	2.4 ②,1	3.2 ⑤,2		
2	3.6 ②	4.3 ①,3	2.5 ④,4	3.7 ⑥,5	9.1 ⑧,6
3	5.7 ③	1.5 ⑤,7	3.1 ⑦,8		
4	9.8 ④	4.1 ②,9	2.5 ⑤,10	2.7 ⑥,11	
5	11.5 ⑤	3.1 ①,12	9.5 ②,13	10.4 ③,14	4.3 ⑦,15
6	12.4 ⑥	6.5 ③,16	9.5 ⑦,17		
7	23.1 ⑦	6.4 ②,18	2.5 ③,19	1.4 ⑥,20	13.1 ⑧,21
8	51.3 ⑧	9.5 ②,22	1.3 ③,23	9.6 ④,24	3.1 ⑥,25

**Diag (i)** Diagonal Components (REAL, i=1~N)  
**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=1, index(N))  
**AMat(k)** Off-Diagonal Components (Value) (REAL, k=1, index(N))

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

```
do i= 1, N
  Y(i)= D(i)*X(i)
  do k= index(i-1)+1, index(i)
    Y(i)= Y(i) + AMAT(k)*X(item(k))
  enddo
enddo
```



- 1D-code for Static Linear-Elastic Problems by Galerkin FEM
- Sparse Linear Solver
  - Conjugate Gradient Method
  - Preconditioning
- Storage of Sparse Matrices
- Program

# Finite Element Procedures

- Initialization
  - Control Data
  - Node, Connectivity of Elements (N: Node#, NE: Elem#)
  - Initialization of Arrays (Global/Element Matrices)
  - Element-Global Matrix Mapping (Index, Item)
- Generation of Matrix
  - Element-by-Element Operations (do icel= 1, NE)
    - Element matrices
    - Accumulation to global matrix
  - Boundary Conditions
- Linear Solver
  - Conjugate Gradient Method

# Program: 1d.f (1/6)

## variables and arrays

```
!C
!C 1D Steady-State Heat Transfer
!C FEM with Piece-wise Linear Elements
!C CG (Conjugate Gradient) Method
!C
!C  $d/dx(CdT/dx) + Q = 0$ 
!C  $T=0@x=0$ 
!C
program heat1D
implicit REAL*8 (A-H, O-Z)

integer :: N, NPLU, ITERmax
integer :: R, Z, P, Q, DD

real(kind=8) :: dX, RESID, EPS
real(kind=8) :: AREA, QV, COND
real(kind=8), dimension(:), allocatable :: PHI, RHS, X
real(kind=8), dimension(:), allocatable :: DIAG, AMAT
real(kind=8), dimension(:, :), allocatable :: W

real(kind=8), dimension(2, 2) :: KMAT, EMAT

integer, dimension(:), allocatable :: ICELNOD
integer, dimension(:), allocatable :: INDEX, ITEM
```

# Variable/Arrays (1/2)

Name	Type	Size	I/O	Definition
<b>NE</b>	I		I	# Element
<b>N</b>	I		O	# Node
<b>NPLU</b>	I		O	# Non-Zero Off-Diag. Components
<b>IterMax</b>	I		I	MAX Iteration Number for CG
<b>errno</b>	I		O	ERROR flag
<b>R, Z, Q, P, DD</b>	I		O	Name of Vectors in CG
<b>dx</b>	R		I	Length of Each Element
<b>Resid</b>	R		O	Residual for CG
<b>Eps</b>	R		I	Convergence Criteria for CG
<b>Area</b>	R		I	Sectional Area of Element
<b>QV</b>	R		I	Heat Generation Rate/Volume/Time $\dot{Q}$
<b>COND</b>	R		I	Thermal Conductivity

# Variable/Arrays (2/2)

Name	Type	Size	I/O	Definition
<b>X</b>	R	N	○	Location of Each Node
<b>PHI</b>	R	N	○	Temperature of Each Node
<b>Rhs</b>	R	N	○	RHS Vector
<b>Diag</b>	R	N	○	Diagonal Components
<b>W</b>	R	(N, 4)	○	Work Array for CG
<b>Amat</b>	R	NPLU	○	Off-Diagonal Components (Value)
<b>Index</b>	I	0:N	○	Number of Non-Zero Off-Diagonals at Each ROW
<b>Item</b>	I	NPLU	○	Off-Diagonal Components (Corresponding Column ID)
<b>Icelnod</b>	I	2*NE	○	Node ID for Each Element
<b>Kmat</b>	R	(2, 2)	○	Element Matrix [k]
<b>Emat</b>	R	(2, 2)	○	Element Matrix

# Program: 1d.f (2/6)

## Initialization, Allocation of Arrays

```
!C
!C +-----+
!C | INIT. |
!C +-----+
!C===
```

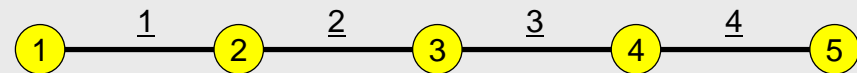
```
open (11, file='input.dat', status='unknown')
read (11,*) NE
read (11,*) dX, QV, AREA, COND
read (11,*) ITERmax
read (11,*) EPS
close (11)
```

Control Data input.dat

4	NE (Number of Elements)
1.0 1.0 1.0 1.0	$\Delta x$ (Length of Each Elem.: $L$ ), $Q$ , $A$ , $\lambda$
100	Number of MAX. Iterations for CG Solver
1.e-8	Convergence Criteria for CG Solver

```
N= NE + 1
allocate (PHI(N), DIAG(N), AMAT(2*N-2), RHS(N))
allocate (ICELNOD(2*NE), X(N))
allocate (INDEX(0:N), ITEM(2*N-2), W(N,4))
```

```
PHI = 0. d0
AMAT= 0. d0
DIAG= 0. d0
RHS= 0. d0
X= 0. d0
```



**NE: # Element**  
**N : # Node (NE+1)**

# Program: 1d.f (2/6)

## Initialization, Allocation of Arrays

```
!C
!C +-----+
!C | INIT. |
!C +-----+
!C===
      open (11, file='input.dat', status='unknown')
      read (11,*) NE
      read (11,*) dX, QV, AREA, COND
      read (11,*) ITERmax
      read (11,*) EPS
      close (11)
```

```
      N= NE + 1
      allocate (PHI(N), DIAG(N), AMAT(2*N-2), RHS(N))
      allocate (ICELNOD(2*NE), X(N))
      allocate (INDEX(0:N), ITEM(2*N-2), W(N,4))
```

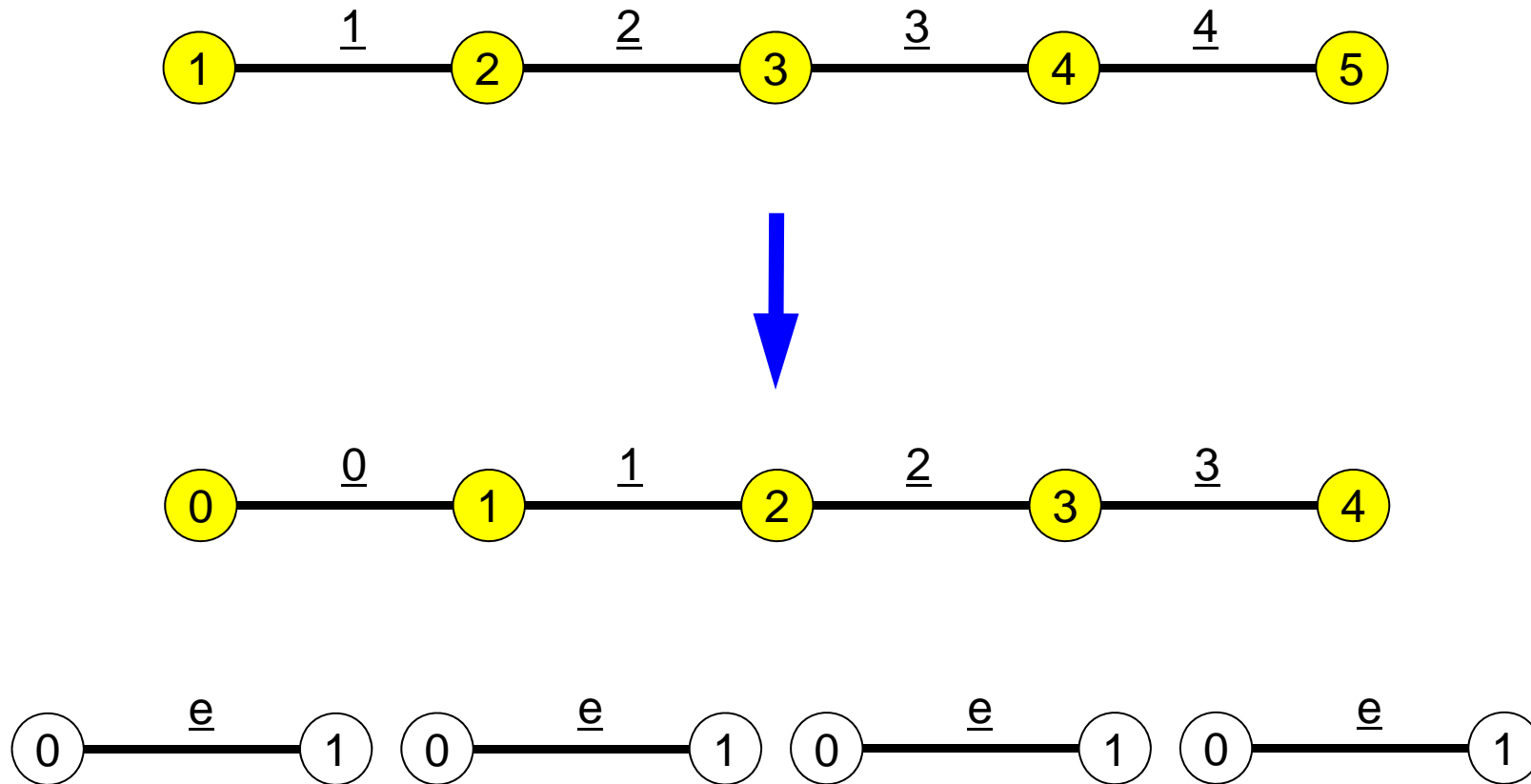
```
      PHI = 0. d0
      AMAT= 0. d0
      DIAG= 0. d0
      RHS= 0. d0
      X= 0. d0
```

**Amat:** Non-Zero Off-Diag. Comp.  
**Item:** Corresponding Column ID





Attention: In C program, node and element ID's start from 0.



# Program: 1d.f (2/6)

## Initialization, Allocation of Array

```

!C
!C +-----+
!C | INIT. |
!C +-----+
!C===
open (11, file='input.dat', status='unknown')
read (11,*) NE
read (11,*) dX, QV, AREA, COND
read (11,*) ITERmax
read (11,*) EPS
close (11)

```

```

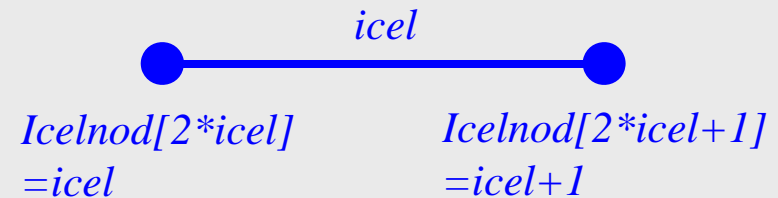
N= NE + 1
allocate (PHI(N), DIAG(N), AMAT(2*N-2), RHS(N))
allocate (ICELNOD(2*NE), X(N))
allocate (INDEX(0:N), ITEM(2*N-2), W(N,4))

```

```

PHI = 0. d0
AMAT= 0. d0
DIAG= 0. d0
RHS= 0. d0
X= 0. d0

```



**Amat:** Non-Zero Off-Diag. Comp.  
**Item:** Corresponding Column ID

Number of non-zero off-diag. components is 2 for each node. This number is 1 at boundary nodes).

**Total Number of Non-Zero Off-Diag. Components:**

$$2 * (N - 2) + 1 + 1 = 2 * N - 2$$

# Program: 1d.f (3/6)

## Initialization, Allocation of Arrays (cont.)

```
do i= 1, N
  X(i)= dfloat(i-1)*dX
enddo

do icel= 1, NE
  ICELNOD(2*icel-1)= icel
  ICELNOD(2*icel )= icel + 1
enddo

KMAT (1, 1)= +1. d0
KMAT (1, 2)= -1. d0
KMAT (2, 1)= -1. d0
KMAT (2, 2)= +1. d0
```

**x:** X-coordinate  
component of each node

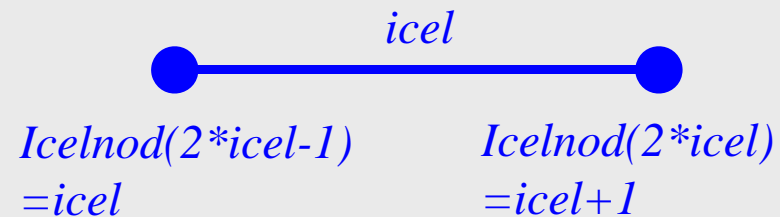
# Program: 1d.f (3/6)

## Initialization, Allocation of Arrays (cont.)

```
do i= 1, N
  X(i)= dfloat (i-1)*dX
enddo
```

```
do icel= 1, NE
  ICELNOD(2*icel-1)= icel
  ICELNOD(2*icel )= icel + 1
enddo
```

```
KMAT (1, 1)= +1. d0
KMAT (1, 2)= -1. d0
KMAT (2, 1)= -1. d0
KMAT (2, 2)= +1. d0
```



# Program: 1d.f (3/6)

## Initialization, Allocation of Arrays (cont.)

```
do i= 1, N
  X(i)= dfloat (i-1)*dX
enddo
```

```
do icel= 1, NE
  ICELNOD (2*icel-1)= icel
  ICELNOD (2*icel )= icel + 1
enddo
```

```
KMAT (1, 1)= +1. d0
KMAT (1, 2)= -1. d0
KMAT (2, 1)= -1. d0
KMAT (2, 2)= +1. d0
```

$$[k]^{(e)} = \int_V \lambda \left( \frac{d[N]^T}{dx} \frac{d[N]}{dx} \right) dV = \frac{\lambda A}{L} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix}$$

[Kmat]

# Program: 1d.f (4/6)

## Global Matrix: Column ID for Non-Zero Off-Diag's

```
!C
!C +-----+
!C | CONNECTIVITY |
!C +-----+
!C===
```

Number of non-zero off-diag. components is 2 for each node. This number is 1 at boundary nodes).

Total Number of Non-Zero Off-Diag. Components:  
 $2*(N-2)+1+1= 2*N-2= NPLU= Index[N]$

**INDEX = 2**

**INDEX(0) = 0**  
**INDEX(1) = 1**  
**INDEX(N) = 1**

```
do i= 1, N
  INDEX(i)= INDEX(i) + INDEX(i-1)
enddo
```

**NPLU= INDEX(N)**

```
do i= 1, N
  jS= INDEX(i-1)
  if (i.eq.1) then
    ITEM(jS+1)= i+1
  else if
& (i.eq.N) then
    ITEM(jS+1)= i-1
  else
    ITEM(jS+1)= i-1
    ITEM(jS+2)= i+1
  endif
enddo
```

```
!C===
```

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

$index(i-1)+1^{th} \sim index(i)^{th}$   
 Non-Zero Off-Diag. Components corresponding to  $i$ -th row

# Program: 1d.f (4/6)

## Global Matrix: Column ID for Non-Zero Off-Diag's

```
!C
!C +-----+
!C | CONNECTIVITY |
!C +-----+
!C===
```

```
INDEX = 2
```

```
INDEX(0) = 0
```

```
INDEX(1) = 1
```

```
INDEX(N) = 1
```

```
do i= 1, N
  INDEX(i) = INDEX(i) + INDEX(i-1)
enddo
```

```
NPLU = INDEX(N)
```

```
do i= 1, N
  jS = INDEX(i-1)
  if (i.eq.1) then
    ITEM(jS+1) = i+1
  else if
& (i.eq.N) then
    ITEM(jS+1) = i-1
  else
    ITEM(jS+1) = i-1
    ITEM(jS+2) = i+1
  endif
enddo
```

```
!C===
```



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

$\text{index}(i-1)+1^{\text{th}} \sim \text{index}(i)^{\text{th}}$   
Non-Zero Off-Diag. Components corresponding to  $i$ -th row

# Program: 1d.f (5/6)

## Element Matrix ~ Global Matrix

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD (2*icel-1)
  in2= ICELNOD (2*icel )
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT (1, 1)= Ck*KMAT (1, 1)
  EMAT (1, 2)= Ck*KMAT (1, 2)
  EMAT (2, 1)= Ck*KMAT (2, 1)
  EMAT (2, 2)= Ck*KMAT (2, 2)

  DIAG(in1)= DIAG(in1) + EMAT (1, 1)
  DIAG(in2)= DIAG(in2) + EMAT (2, 2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

  AMAT (k1)= AMAT (k1) + EMAT (1, 2)
  AMAT (k2)= AMAT (k2) + EMAT (2, 1)

  QN= 0.50d0*QV*AREA*DL
  RHS (in1)= RHS (in1) + QN
  RHS (in2)= RHS (in2) + QN
enddo
!C===

```





# Program: 1d.f (5/6)

## Element Matrix ~ Global Matrix

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD(2*icel-1)
  in2= ICELNOD(2*icel )
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT(1,1)= Ck*KMAT(1,1)
  EMAT(1,2)= Ck*KMAT(1,2)
  EMAT(2,1)= Ck*KMAT(2,1)
  EMAT(2,2)= Ck*KMAT(2,2)

  DIAG(in1)= DIAG(in1) + EMAT(1,1)
  DIAG(in2)= DIAG(in2) + EMAT(2,2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

  AMAT(k1)= AMAT(k1) + EMAT(1,2)
  AMAT(k2)= AMAT(k2) + EMAT(2,1)

  QN= 0.50d0*QV*AREA*DL
  RHS(in1)= RHS(in1) + QN
  RHS(in2)= RHS(in2) + QN
enddo
!C===

```



$$[Emat] = [k]^{(e)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix} = \frac{\lambda A}{L} [Kmat]$$

# Program: 1d.f (5/6)

## Element Matrix ~ Global Matrix

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD(2*icel-1)
  in2= ICELNOD(2*icel )
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT(1,1)= Ck*KMAT(1,1)
  EMAT(1,2)= Ck*KMAT(1,2)
  EMAT(2,1)= Ck*KMAT(2,1)
  EMAT(2,2)= Ck*KMAT(2,2)

  DIAG(in1)= DIAG(in1) + EMAT(1,1)
  DIAG(in2)= DIAG(in2) + EMAT(2,2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

  AMAT(k1)= AMAT(k1) + EMAT(1,2)
  AMAT(k2)= AMAT(k2) + EMAT(2,1)

  QN= 0.50d0*QV*AREA*DL
  RHS(in1)= RHS(in1) + QN
  RHS(in2)= RHS(in2) + QN
enddo
!C===

```



$$[Emat] = [k]^{(e)} = \frac{EA}{L} \begin{bmatrix} +1 & -1 \\ -1 & +1 \end{bmatrix}$$

# Program: 1d.f (5/6)

## Element Matrix ~ Global Matrix

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD(2*icel-1)
  in2= ICELNOD(2*icel)
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT(1,1)= Ck*KMAT(1,1)
  EMAT(1,2)= Ck*KMAT(1,2)
  EMAT(2,1)= Ck*KMAT(2,1)
  EMAT(2,2)= Ck*KMAT(2,2)

  DIAG(in1)= DIAG(in1) + EMAT(1,1)
  DIAG(in2)= DIAG(in2) + EMAT(2,2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

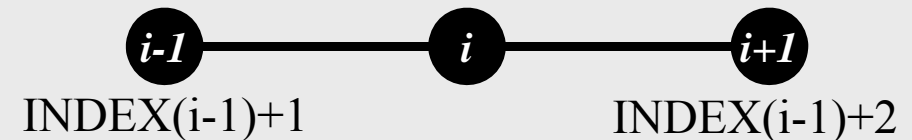
  AMAT(k1)= AMAT(k1) + EMAT(1,2)
  AMAT(k2)= AMAT(k2) + EMAT(2,1)

  QN= 0.5d0*QV*AREA*DL
  RHS(in1)= RHS(in1) + QN
  RHS(in2)= RHS(in2) + QN
enddo
!C===

```



Non-zero Off-Diag. at  $i$ -th row:  
**Index( $i-1$ )+1, Index( $i-1$ )+2**



$$[Emat] = [k]^{(e)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & \ominus 1 \\ \ominus 1 & +1 \end{bmatrix} \begin{matrix} k1 \\ k2 \end{matrix}$$

# General Elements: k1

“in2” as a off-diag. component of “in1”

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD(2*icel-1)
  in2= ICELNOD(2*icel )
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT(1,1)= Ck*KMAT(1,1)
  EMAT(1,2)= Ck*KMAT(1,2)
  EMAT(2,1)= Ck*KMAT(2,1)
  EMAT(2,2)= Ck*KMAT(2,2)

  DIAG(in1)= DIAG(in1) + EMAT(1,1)
  DIAG(in2)= DIAG(in2) + EMAT(2,2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

  AMAT(k1)= AMAT(k1) + EMAT(1,2)
  AMAT(k2)= AMAT(k2) + EMAT(2,1)

  QN= 0.50d0*QV*AREA*DL
  RHS(in1)= RHS(in1) + QN
  RHS(in2)= RHS(in2) + QN
enddo
!C===

```



Non-zero Off-Diag. at  $i$ -th row:  
**INDEX(i-1)+1, INDEX(i-1)+2**



$$[Emat] = [k]^{(e)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & \ominus -1 \\ -1 & +1 \end{bmatrix} \quad k1$$

# General Elements: k2

“in1” as a off-diag. component of “in2”

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD(2*icel-1)
  in2= ICELNOD(2*icel )
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT(1, 1)= Ck*KMAT(1, 1)
  EMAT(1, 2)= Ck*KMAT(1, 2)
  EMAT(2, 1)= Ck*KMAT(2, 1)
  EMAT(2, 2)= Ck*KMAT(2, 2)

  DIAG(in1)= DIAG(in1) + EMAT(1, 1)
  DIAG(in2)= DIAG(in2) + EMAT(2, 2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

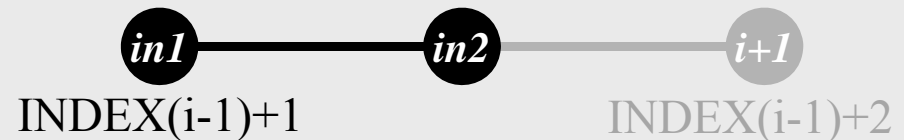
  AMAT(k1)= AMAT(k1) + EMAT(1, 2)
  AMAT(k2)= AMAT(k2) + EMAT(2, 1)

  QN= 0.50d0*QV*AREA*DL
  RHS(in1)= RHS(in1) + QN
  RHS(in2)= RHS(in2) + QN
enddo
!C===

```



Non-zero Off-Diag. at  $i$ -th row:  
**Index(i-1)+1, Index(i-1)+2**



$$[Emat] = [k]^{(e)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & -1 \\ \textcircled{-1} & +1 \end{bmatrix}$$

k2

# 0-th Element: k1

“in2” as a off-diag. component of “in1”

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD(2*icel-1)
  in2= ICELNOD(2*icel )
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT(1, 1)= Ck*KMAT(1, 1)
  EMAT(1, 2)= Ck*KMAT(1, 2)
  EMAT(2, 1)= Ck*KMAT(2, 1)
  EMAT(2, 2)= Ck*KMAT(2, 2)

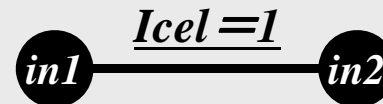
  DIAG(in1)= DIAG(in1) + EMAT(1, 1)
  DIAG(in2)= DIAG(in2) + EMAT(2, 2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

  AMAT(k1)= AMAT(k1) + EMAT(1, 2)
  AMAT(k2)= AMAT(k2) + EMAT(2, 1)

  QN= 0.50d0*QV*AREA*DL
  RHS(in1)= RHS(in1) + QN
  RHS(in2)= RHS(in2) + QN
enddo
!C===

```



Non-zero Off-Diag. at  $i$ -th row:  
**INDEX( $i-1$ )+1 only**



$$[Emat] = [k]^{(e)} = \frac{\lambda A}{L} \begin{bmatrix} +1 & \ominus -1 \\ -1 & +1 \end{bmatrix} \quad k1$$

# Program: 1d.f (5/6)

## RHS: Heat Generation Term

```

!C +-----+
!C | MATRIX ASSEMBLE |
!C +-----+
!C===
do icel= 1, NE
  in1= ICELNOD(2*icel-1)
  in2= ICELNOD(2*icel )
  X1 = X(in1)
  X2 = X(in2)
  DL = dabs(X2-X1)

  cK= AREA*COND/DL
  EMAT(1,1)= Ck*KMAT(1,1)
  EMAT(1,2)= Ck*KMAT(1,2)
  EMAT(2,1)= Ck*KMAT(2,1)
  EMAT(2,2)= Ck*KMAT(2,2)

  DIAG(in1)= DIAG(in1) + EMAT(1,1)
  DIAG(in2)= DIAG(in2) + EMAT(2,2)

  if (icel.eq.1) then
    k1= INDEX(in1-1) + 1
  else
    k1= INDEX(in1-1) + 2
  endif
  k2= INDEX(in2-1) + 1

  AMAT(k1)= AMAT(k1) + EMAT(1,2)
  AMAT(k2)= AMAT(k2) + EMAT(2,1)

  QN= 0.50d0*QV*AREA*DL
  RHS(in1)= RHS(in1) + QN
  RHS(in2)= RHS(in2) + QN
enddo
!C===

```



$$\int_V \dot{Q}[N]^T dV = \dot{Q}A \int_0^L \begin{bmatrix} 1-x/L \\ x/L \end{bmatrix} dx = \frac{\dot{Q}AL}{2} \begin{Bmatrix} 1 \\ 1 \end{Bmatrix}$$

# Program: 1d.f (6/6)

## Dirichlet B.C. @ X=0

```
!C
!C +-----+
!C | BOUNDARY CONDITIONS |
!C +-----+
!C===

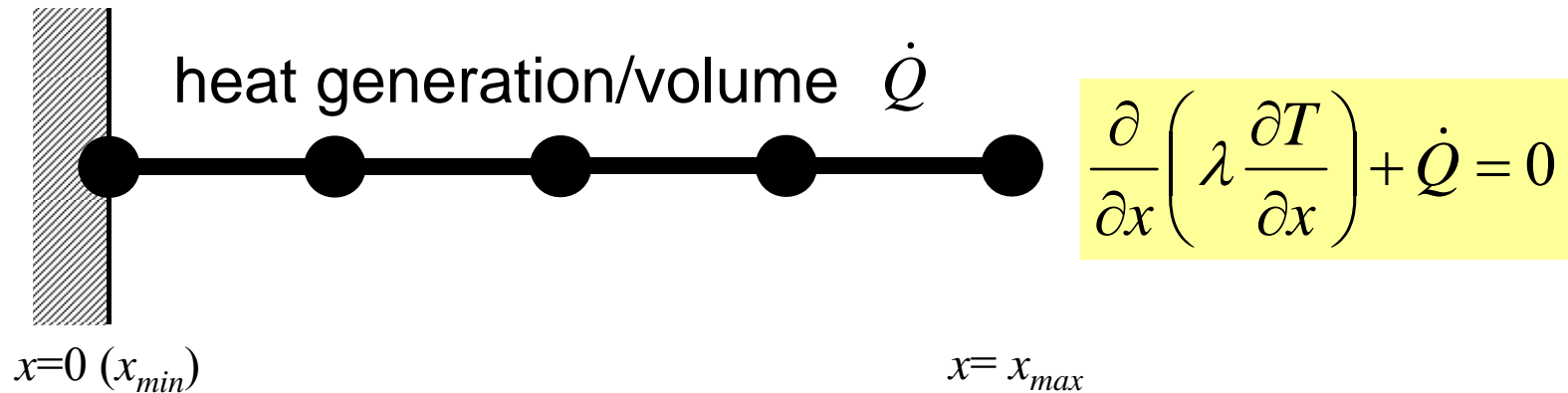
!C
!C-- X=Xmin
      i= 1
      jS= INDEX(i-1)

      AMAT(jS+1)= 0. d0
      DIAG(i)= 1. d0
      RHS (i)= 0. d0

      do k= 1, NPLU
        if (ITEM(k).eq. 1) AMAT(k)= 0. d0
      enddo
!C===
```



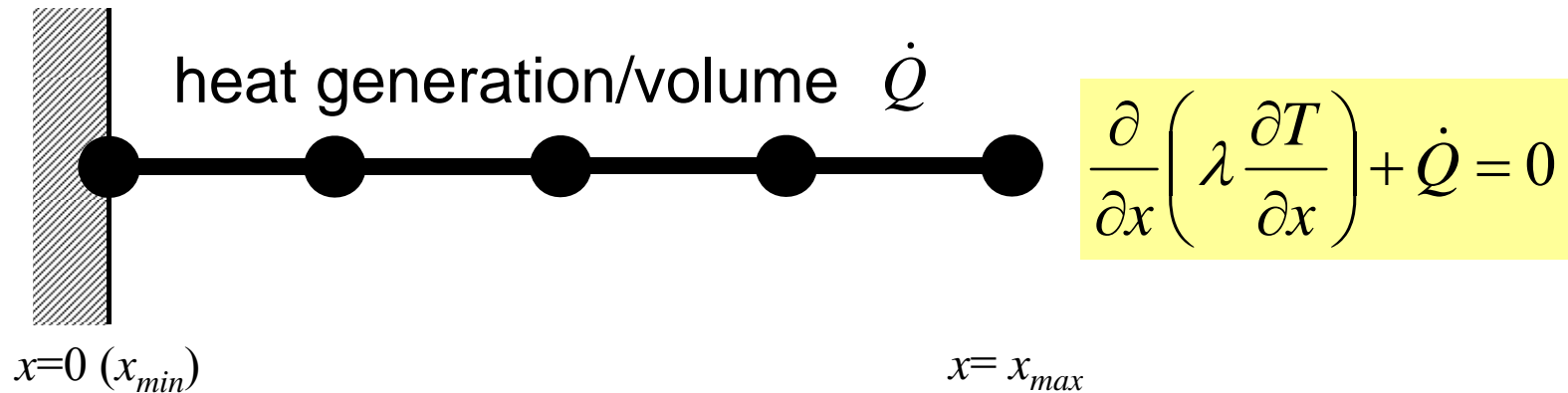
# 1D Steady State Heat Conduction



- **Uniform: Sectional Area:  $A$ , Thermal Conductivity:  $\lambda$**
- Heat Generation Rate/Volume/Time [ $\text{QL}^{-3}\text{T}^{-1}$ ]  $\dot{Q}$
- Boundary Conditions
  - $x=0$  :  $T=0$  (Fixed Temperature)
  - $x=x_{max}$  :  $\frac{\partial T}{\partial x} = 0$  (Insulated)

# (Linear) Equation at $x=0$

$$T_1 = 0 \text{ (or } T_0 = 0)$$



- **Uniform: Sectional Area:  $A$ , Thermal Conductivity:  $\lambda$**
- Heat Generation Rate/Volume/Time [ $\text{QL}^{-3}\text{T}^{-1}$ ]  $\dot{Q}$
- Boundary Conditions
  - $x=0$  :  $T=0$  (Fixed Temperature)
  - $x=x_{max}$  :  $\frac{\partial T}{\partial x} = 0$  (Insulated)

# Program: 1d.f (6/6)

## Dirichlet B.C. @ X=0

```

!C
!C +-----+
!C | BOUNDARY CONDITIONS |
!C +-----+
!C===

!C
!C-- X=Xmin
  i= 1
  js= INDEX(i-1)

  AMAT (js+1)= 0. d0
  DIAG (i)= 1. d0
  RHS (i)= 0. d0

  do k= 1, NPLU
    if (ITEM(k).eq. 1) AMAT (k)= 0. d0
  enddo
!C===

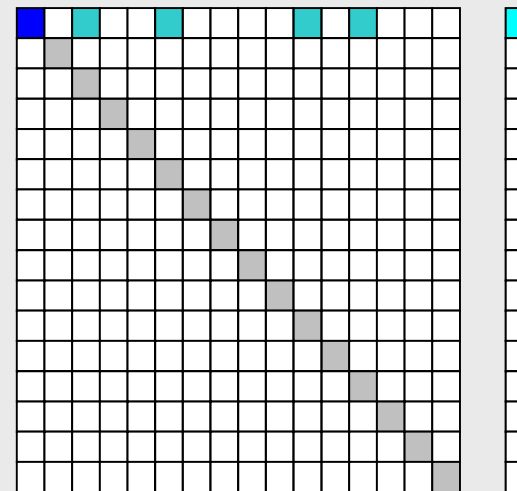
```

$$T_1=0$$

Diagonal Component=1

RHS=0

Off-Diagonal Components= 0.



# Program: 1d.f (6/6)

## Dirichlet B.C. @ X=0

```

!C
!C +-----+
!C | BOUNDARY CONDITIONS |
!C +-----+
!C===

!C
!C-- X=Xmin
  i= 1
  js= INDEX(i-1)

  AMAT (js+1)= 0. d0
  DIAG (i)= 1. d0
  RHS (i)= 0. d0

  do k= 1, NPLU
    if (ITEM(k).eq. 1) AMAT (k)= 0. d0
  enddo
!C===

```

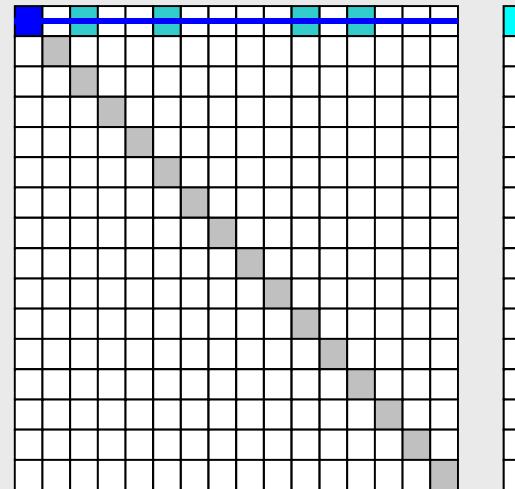
$$T_1=0$$

Diagonal Component=1

RHS=0

Off-Diagonal Components= 0.

Erase !



# Program: 1d.f (6/6)

## Dirichlet B.C. @ X=0

```

!C
!C +-----+
!C | BOUNDARY CONDITIONS |
!C +-----+
!C===

!C
!C-- X=Xmin
      i= 1
      js= INDEX(i-1)

      AMAT (js+1)= 0. d0
      DIAG (i)= 1. d0
      RHS (i)= 0. d0

      do k= 1, NPLU
        if (ITEM(k).eq. 1) AMAT (k)= 0. d0
      enddo
!C===

```

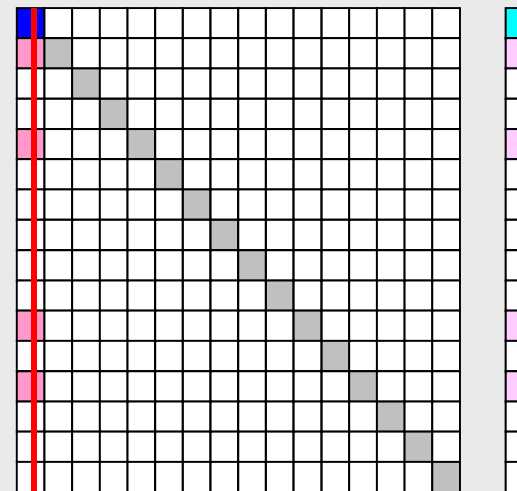
$$T_1=0$$

Diagonal Component=1

RHS=0

Off-Diagonal Components= 0.

### Elimination and Erase



Column components of boundary nodes (Dirichlet B.C.) are moved to RHS and eliminated for keeping symmetrical feature of the matrix (in this case just erase off-diagonal components)

# if $T_1 \neq 0$

```
!C
!C +-----+
!C | BOUNDARY CONDITIONS |
!C +-----+
!C===
```

Column components of boundary nodes (Dirichlet B.C.) are moved to RHS and eliminated for keeping symmetrical feature of the matrix.

```
!C
!C-- X=Xmin
      i= 1
      jS= INDEX(i-1)

      AMAT(jS+1)= 0. d0
      DIAG(i)= 1. d0
      RHS (i)= PHImin
```

$$Diag_j \phi_j + \sum_{k=Index[j]}^{Index[j+1]-1} Amat_k \phi_{Item[k]} = Rhs_j$$

```
do i= 1, N
  do k= INDEX(i-1)+1, INDEX(i)
    if (ITEM(k).eq.1) then
      RHS (i)= RHS(i) - AMAT(k)*PHImin
      AMAT(k)= 0. d0
    endif
  enddo
enddo
```

```
!C===
```

# if $T_1 \neq 0$

```
!C
!C +-----+
!C | BOUNDARY CONDITIONS |
!C +-----+
!C===
```

```
!C
!C-- X=Xmin
      i= 1
      jS= INDEX(i-1)
```

```
      AMAT(jS+1)= 0. d0
      DIAG(i)= 1. d0
      RHS (i)= PHImin
```

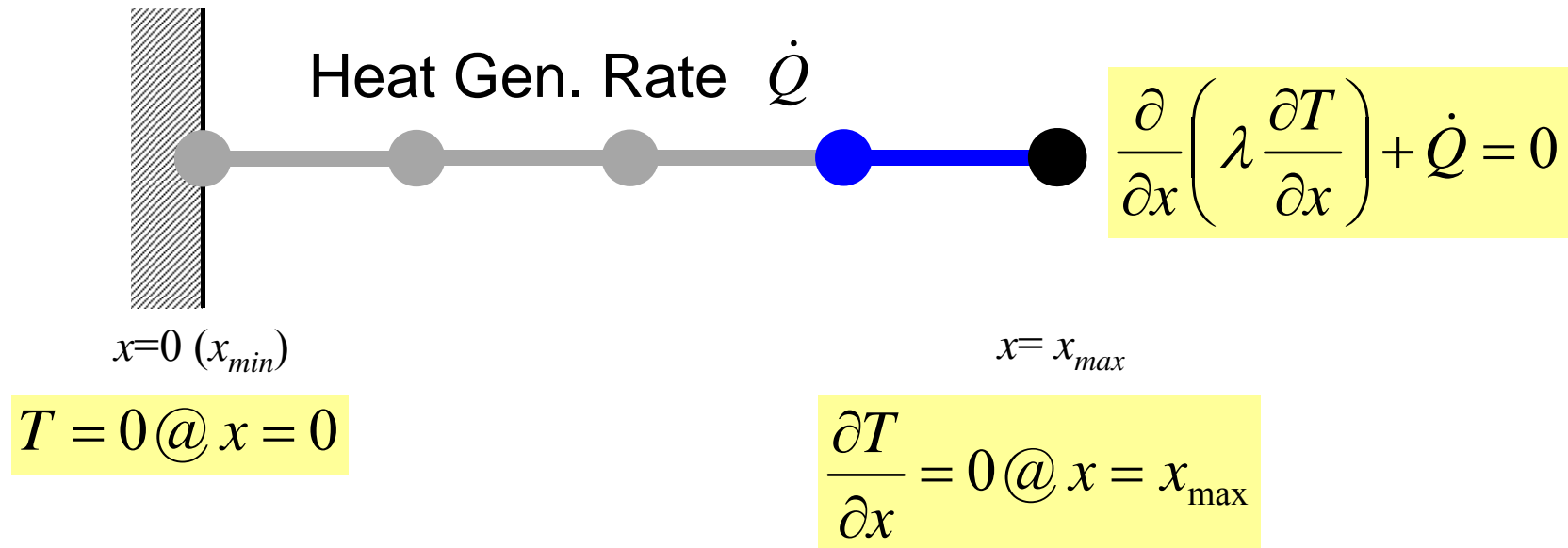
```
      do i= 1, N
        do k= INDEX(i-1)+1, INDEX(i)
          if (ITEM(k).eq.1) then
            RHS (i)= RHS(i) - AMAT(k)*PHImin
            AMAT(k)= 0. d0
          endif
        enddo
      enddo
```

```
!C===
```

$$\begin{aligned}
 & \text{Diag}_j \phi_j + \sum_{k=\text{Index}[j], k \neq k_s}^{\text{Index}[j+1]-1} \text{Amat}_k \phi_{\text{Item}[k]} \\
 &= \text{Rhs}_j - \text{Amat}_{k_s} \phi_{\text{Item}[k_s]} \\
 &= \text{Rhs}_j - \text{Amat}_{k_s} \phi_{\min} \quad \text{where } \text{Item}[k_s] = 0
 \end{aligned}$$

Column components of boundary nodes (Dirichlet B.C.) are moved to RHS and eliminated for keeping symmetrical feature of the matrix.

# Secondary B.C. (Insulated)



$$\int_S \bar{q} [N]^T dS = \bar{q} A|_{x=L} = \bar{q} A \begin{Bmatrix} 0 \\ 1 \end{Bmatrix}, \quad \bar{q} = -\lambda \frac{dT}{dx} \quad \text{Surface Flux}$$



$$\frac{\partial T}{\partial x} = 0 @ x = x_{max}$$

According to insulated B.C.,  $\bar{q} = 0$  is satisfied. No contribution by this term. Insulated B.C. is automatically satisfied without explicit operations  
 -> Natural B.C.



# Preconditioned CG Solver

```

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

```

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

# 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  $[M]z^{(i-1)} = r^{(i-1)}$**  is very easy.
- Provides fast convergence for simple problems.
- 1d.f, 1d.c

# CG Solver (1/6)

```
!C
!C +-----+
!C | CG iterations |
!C +-----+
!C===
      R = 1
      Z = 2
      Q = 2
      P = 3
      DD= 4

      do i= 1, N
        W(i,DD)= 1.0D0 / DIAG(i)
      enddo
```

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

# CG Solver (1/6)

```
!C
!C +-----+
!C | CG iterations |
!C +-----+
!C===
      R = 1
      Z = 2
      Q = 2
      P = 3
      DD= 4

      do i= 1, N
        W(i,DD)= 1.0D0 / DIAG(i)
      enddo
```

```
W(i, 1) = W(i, R)   => {r}
W(i, 2) = W(i, Z)   => {z}
W(i, 2) = W(i, Q)   => {q}
W(i, 3) = W(i, P)   => {p}
W(i, 4) = W(i, DD) => 1/{D}
```

```
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
```

# CG Solver (2/6)

```
!C
!C-- {r0} = {b} - [A]{xini} |
```

**!C 初期残差**

```
do i = 1, N
  W(i,R) = DIAG(i)*PHI(i)
  do j = INDEX(i-1)+1, INDEX(i)
    W(i,R) = W(i,R) + AMAT(j)*PHI(ITEM(j))
  enddo
enddo
```

```
BNRM2 = 0.0D0
do i = 1, N
  BNRM2 = BNRM2 + RHS(i) **2
  W(i,R) = RHS(i) - W(i,R)
enddo
```

**BNRM2 = |b|<sup>2</sup>**  
**for convergence criteria**  
**of CG solvers**

**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)
  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
```

# CG Solver (3/6)

```

do iter= 1, ITERmax

!C
!C-- {z}= [Minv]{r}

do i= 1, N
  W(i,Z)= W(i,DD) * W(i,R)
enddo

!C
!C-- RHO= {r}{z}

RHO= 0.d0
do i= 1, N
  RHO= RHO + W(i,R)*W(i,Z)
enddo

```

```

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

```

# CG Solver (4/6)

```

!C
!C-- {p} = {z} if      ITER=1
!C  BETA= RHO / RH01  otherwise

      if ( iter.eq.1 ) then
        do i= 1, N
          W(i,P)= W(i,Z)
        enddo
      else
        BETA= RHO / RH01
        do i= 1, N
          W(i,P)= W(i,Z) + BETA*W(i,P)
        enddo
      endif

!C
!C-- {q} = [A] {p}

      do i= 1, N
        W(i,Q) = DIAG(i)*W(i,P)
        do j= INDEX(i-1)+1, INDEX(i)
          W(i,Q) = W(i,Q) + AMAT(j)*W(ITEM(j),P)
        enddo
      enddo

```

```

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

```

# CG Solver (5/6)

```

!C
!C-- ALPHA= RHO / {p} {q}

      C1= 0.d0
      do i= 1, N
        C1= C1 + W(i, P)*W(i, Q)
      enddo
      ALPHA= RHO / C1

!C
!C-- {x}= {x} + ALPHA*{p}
!C  {r}= {r} - ALPHA*{q}

      do i= 1, N
        PHI (i)= PHI (i) + ALPHA * W(i, P)
        W(i, R)= W(i, R) - ALPHA * W(i, Q)
      enddo

```

```

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

```



# CG Solver (6/6)

```

DNRM2 = 0.0
do i= 1, N
  DNRM2= DNRM2 + W(i,R)**2
enddo

RESID= dsqrt(DNRM2/BNRM2)

if ( RESID. le. EPS) goto 900
RHO1 = RHO  ρi-2

enddo
900 continue

```

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

**Control Data** input.dat

```

4          NE (Number of Elements)
1.0 1.0 1.0 1.0 Δx (Length of Each Elem.: L), Q, A, λ
100       Number of MAX. Iterations for CG Solver
1.e-8     Convergence Criteria for CG Solver

```

```

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)
  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

```

# Finite Element Procedures

- Initialization
  - Control Data
  - Node, Connectivity of Elements (N: Node#, NE: Elem#)
  - Initialization of Arrays (Global/Element Matrices)
  - Element-Global Matrix Mapping (Index, Item)
- Generation of Matrix
  - Element-by-Element Operations (do icel= 1, NE)
    - Element matrices
    - Accumulation to global matrix
  - Boundary Conditions
- Linear Solver
  - Conjugate Gradient Method

# Remedies for Higher Accuracy

- Finer Meshes

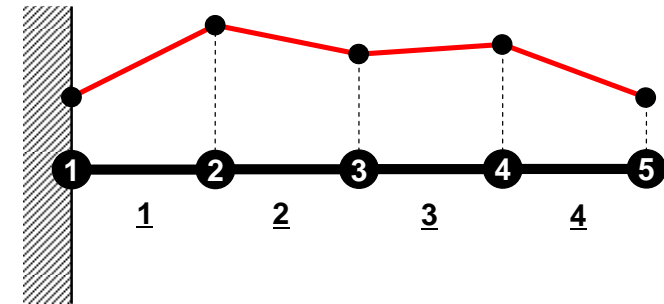
NE=8, dx=12.5

8 iters, RESID= 2.822910E-16 U(N)= 1.953586E-01

### DISPLACEMENT

1	0.000000E+00	-0.000000E+00
2	1.101928E-02	1.103160E-02
3	2.348034E-02	2.351048E-02
4	3.781726E-02	3.787457E-02
5	5.469490E-02	5.479659E-02
6	7.520772E-02	7.538926E-02
7	1.013515E-01	1.016991E-01
8	1.373875E-01	1.381746E-01
9	1.953586E-01	1.980421E-01

$$\therefore u = \frac{F}{EA_1} [\log(A_1 x + A_2) - \log(A_2)]$$



NE=20, dx=5

20 iters, RESID= 5.707508E-15 U(N)= 1.975734E-01

### DISPLACEMENT

1	0.000000E+00	-0.000000E+00
2	4.259851E-03	4.260561E-03
3	8.719160E-03	8.720685E-03
4	1.339752E-02	1.339999E-02
.....		
17	1.145876E-01	1.146641E-01
18	1.295689E-01	1.296764E-01
19	1.473466E-01	1.475060E-01
20	1.692046E-01	1.694607E-01
21	1.975734E-01	1.980421E-01

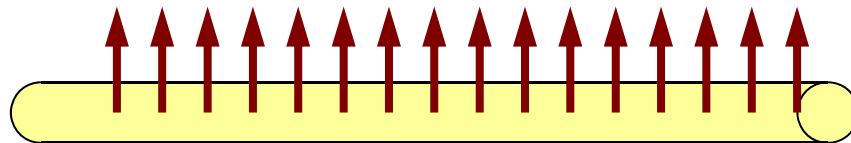
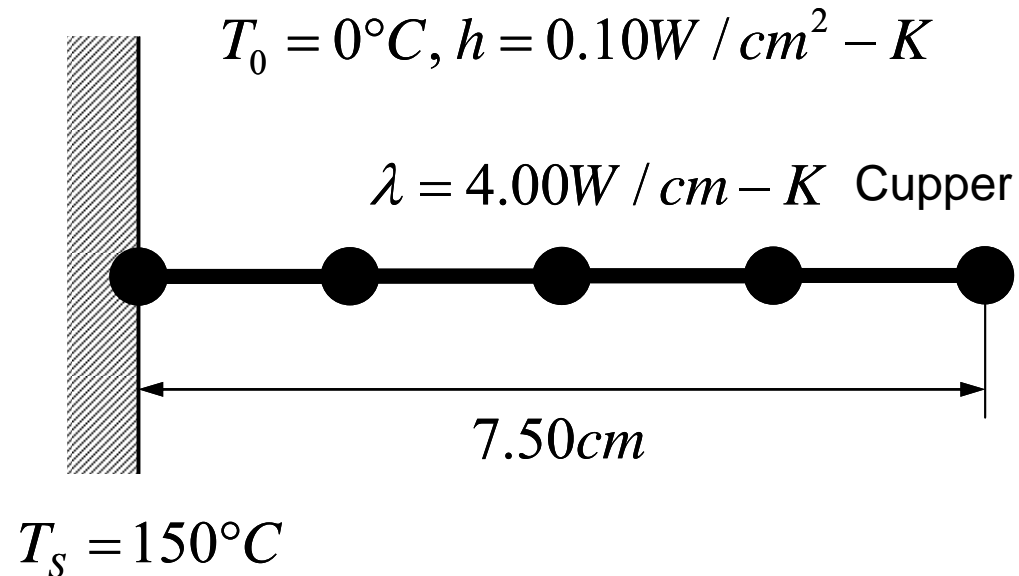
# Remedies for Higher Accuracy

- Finer Meshes
- Higher Order Shape/Interpolation Function (高次補間関数・形状関数)
  - Higher-Order Element (高次要素)
  - Linear-Element, 1<sup>st</sup>-Order Element: Lower Order (低次要素)
- Formulation which assures continuity of n-th order derivatives
  - $C^n$  Continuity ( $C^n$ 連続性)

# Remedies for Higher Accuracy

- Finer Meshes
- Higher Order Shape/Interpolation Function (高次補間関数・形状関数)
  - Higher-Order Element (高次要素)
  - Linear-Element, 1<sup>st</sup>-Order Element: Lower Order (低次要素)
- Formulation which assures continuity of n-th order derivatives
  - $C^n$  Continuity ( $C^n$ 連続性)
- **Linear Elements**
  - Piecewise Linear
  - $C^0$  Continuity
    - Only dependent variables are continuous at element boundary

# Example: 1D Heat Transfer (1/2)



Convective Heat Transfer on  
Cylindrical Surface

- Temp. Thermal Fins
- Circular Sectional Area,  $r=1\text{cm}$
- Boundary Condition
  - $x=0$  : Fixed Temperature
  - $x=7.5$  : Insulated
- Convective Heat Transfer on Cylindrical Surface
  - $q = h(T - T_0)$
  - $q$  : Heat Flux
    - Heat Flow/Unit Surface Area/sec.

# Example: 1D Heat Transfer (2/2)

## ### RESULTS (linear interpolation)

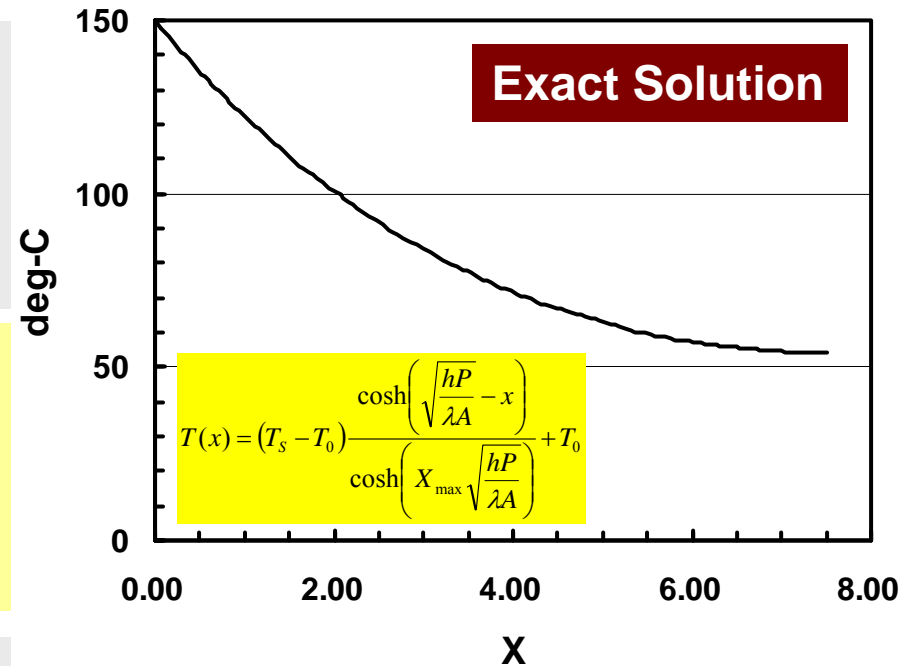
ID	X	FEM.	ANALYTICAL	ERR (%)
1	0.00000	150.00000	150.00000	0.00000
2	1.87500	102.62226	103.00165	0.25292
3	3.75000	73.82803	74.37583	0.36520
4	5.62500	58.40306	59.01653	0.40898
5	7.50000	53.55410	54.18409	0.41999

## ### RESULTS (quadratic interpolation)

ID	X	FEM.	ANALYTICAL	ERR (%)
1	0.00000	150.00000	150.00000	0.00000
2	1.87500	102.98743	103.00165	0.00948
3	3.75000	74.40203	74.37583	0.01747
4	5.62500	59.02737	59.01653	0.00722
5	7.50000	54.21426	54.18409	0.02011

## ### RESULTS (linear interpolation)

ID	X	FEM.	ANALYTICAL	ERR (%)
1	0.00000	150.00000	150.00000	0.00000
2	0.93750	123.71561	123.77127	0.03711
3	1.87500	102.90805	103.00165	0.06240
4	2.81250	86.65618	86.77507	0.07926
5	3.75000	74.24055	74.37583	0.09019
6	4.68750	65.11151	65.25705	0.09703
7	5.62500	58.86492	59.01653	0.10107
8	6.56250	55.22426	55.37903	0.10317
9	7.50000	54.02836	54.18409	0.10382



Quadratic interpolation provides more accurate solution, especially if X is close to 7.50cm.