

# Introduction to Applied Scientific Computing using MATLAB

Mohsen Jenadeleh

In this lecture, slides from MIT, Rutgers and Waterloo University are used to form the lecture slides

# Matrix Algebra

- dot product
- matrix-vector multiplication
- matrix-matrix multiplication
- matrix inverse
- solving linear systems
- least-squares solutions
- determinant, rank, condition number
- vector & matrix norms
- iterative solutions of linear systems
- examples
- electric circuits
- temperature distributions

# Iterative solutions of linear systems $Ax=b$

the only practical way to solve very large linear systems is iteratively

## Methods:

- 
1. Jacobi method
  2. Gauss-Seidel method
  3. Relaxation methods
  4. Conjugate Gradient method
  5. Others

- G. H. Golub and C. F. Van Loan, *Matrix Computations*, 3/e, JHU Press, 1996.  
D. S. Watkins, *Fundamentals of Matrix Computations*, 2/e, Wiley, 2002.  
L. N. Trefethen and D. Bau, *Numerical Linear Algebra*, SIAM, 1997.  
A. Bjork, *Numerical Methods for Least Squares Problems*, SIAM, 1996.

rearrange

$$3x = 12$$

$$2x + x = 12$$

rearrange

$$2x = -x + 12$$

$$x = -0.5x + 6$$

scalar example  
illustrating the  
Jacobi method

turn it into a recursion

for  $k = 1, 2, 3, \dots$

$$x(k+1) = -0.5x(k) + 6$$

start with any  $x(1)$ ,

$$x(2) = -0.5x(1) + 6$$

$$x(3) = -0.5x(2) + 6$$

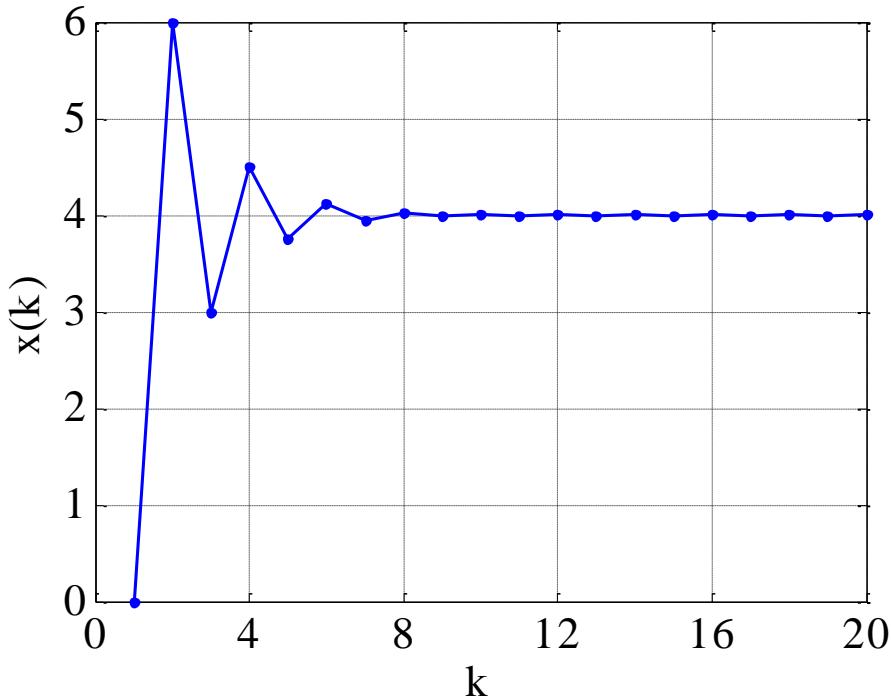
$$x(4) = -0.5x(3) + 6, \text{ etc.}$$

```

x(1)=0; % arbitrary
for k=1:19
    x(k+1) = -0.5*x(k) + 6;
end

k=1:20; plot(k,x,'b.-');

```



```

>> [k; x] '
1 0.0000
2 6.0000
3 3.0000
4 4.5000
5 3.7500
6 4.1250
7 3.9375
8 4.0313
9 3.9844
10 4.0078
11 3.9961
12 4.0020
13 3.9990
14 4.0005
15 3.9998
16 4.0001
17 3.9999
18 4.0000
19 4.0000
20 4.0000

```

```

tol=1e-10; x0=0;

x=x0; k=1;

while 1
    xnew = -0.5*x + 6;
    if abs(xnew-x)<=tol
        break;
    end
    x = xnew;
    k = k+1;
end

```

k, abs(x-4)

**k** =  
37  
**ans** =  
5.8208e-011

forever  
while loop

```

tol=1e-10; x0=0;

x=x0; k=1;
xnew = -0.5*x+6;

while abs(xnew-x)>tol
    x = xnew;
    k = k+1;
    xnew = -0.5*x + 6;
end

```

k, abs(x-4)

**k** =  
37  
**ans** =  
5.8208e-011

conventional  
while loop

More general version of  $ax=b$ , where  $a, x, b$  are scalars

choose a splitting  $a = d - r$ , such that  $|r/d| < 1$

$$ax = (d - r)x = b$$

$$dx = rx + b$$

$$\rightarrow x = \frac{r}{d}x + \frac{b}{d}$$

key assumption that  
guarantees convergence

this is the scalar version of the  
Jacobi and Gauss-Seidel methods

iterative algorithm:

$$x(k+1) = \frac{r}{d}x(k) + \frac{b}{d}, \quad k = 0, 1, 2, \dots$$

it has solution that converges to  $b/a$ :

$$x(k) = \frac{b}{a} + \left(\frac{r}{d}\right)^k \left(x_0 - \frac{b}{a}\right), \quad k = 0, 1, 2, \dots$$

## Jacobi's method for $A \mathbf{x} = \mathbf{b}$

Define the following ‘Jacobi iteration parameters’

$D = \text{diag}(\text{diag}(A))$  = diagonal part of  $A$

$B = I - D^{-1}A$ ,     $I$  = identity matrix

$\mathbf{c} = D^{-1} \mathbf{b}$

$$A = D - DB$$

□  $A\mathbf{x} = \mathbf{b} \quad \Rightarrow \quad D\mathbf{x} - DB\mathbf{x} = \mathbf{b} \quad \Rightarrow \quad D\mathbf{x} = DB\mathbf{x} + \mathbf{b}$

→  $\mathbf{x} = B\mathbf{x} + \mathbf{c}$



Jacobi iteration:

$$\mathbf{x}(k+1) = B\mathbf{x}(k) + \mathbf{c}, \quad k = 0, 1, 2, \dots$$

Convergence requires the condition that the so-called ‘spectral radius’ of  $B$  be strictly less than unity:

$$\rho(B) = \max(\text{abs}(\text{eig}(B))) < 1$$

This is hard to calculate for large matrices, but a sufficient condition for convergence is that a matrix norm of  $B$ , such as the  $L_1$ ,  $L_2$ ,  $L_\infty$ , or Frobenius norms, be less than unity because of the inequality:

$$\rho(B) \leq \|B\|$$

so that

$$\|B\| < 1 \Rightarrow \rho(B) < 1$$

Note: if the method fails for a system  $\mathbf{A}, \mathbf{b}$ , then, try it on the rearranged system

$\mathbf{A} \rightarrow \text{flipud}(\mathbf{A})$

$\mathbf{b} \rightarrow \text{flipud}(\mathbf{b})$

which sometimes works

Jacobi iteration:

$$\mathbf{x}(k+1) = B\mathbf{x}(k) + \mathbf{c}, \quad k = 0, 1, 2, \dots$$

Exact solution demonstrates the convergence:

$$\mathbf{x}(k) = A^{-1}\mathbf{b} + B^k(\mathbf{x}_0 - A^{-1}\mathbf{b}), \quad k = 0, 1, 2, \dots$$

Jacobi iteration:

$$\mathbf{x}(k+1) = B\mathbf{x}(k) + \mathbf{c}, \quad k = 0, 1, 2, \dots$$

Jacobi iteration – alternative form:

$$\mathbf{x}(k+1) = \mathbf{x}(k) + D^{-1}[\mathbf{b} - A\mathbf{x}(k)], \quad k = 0, 1, 2, \dots$$

Jacobi iteration – relaxation form:

$$\mathbf{x}(k+1) = \mathbf{x}(k) + \omega D^{-1}[\mathbf{b} - A\mathbf{x}(k)], \quad k = 0, 1, 2, \dots$$



relaxation parameter

## MATLAB implementation

```
% given A = NxN matrix, b = Nx1 vector  
% construct the Jacobi parameters  
  
I = eye(size(A)) ; % identity matrix  
D = diag(diag(A)) ; % diagonal part of A  
  
B = I - D\A; % iteration matrix  
c = D\b; % Nx1 vector  
  
rho = max(abs(eig(B))) ; % check if rho < 1
```

Next, implement the Jacobi iteration using a forever while-loop with a stopping condition,

$$\mathbf{x}(k+1) = B\mathbf{x}(k) + \mathbf{c}, \quad k = 0, 1, 2, \dots$$

## MATLAB implementation – v.1

```
tol = 1e-10;          % error tolerance, our choice
x = x0; k = 1;         % x0 = arbitrary Nx1 vector

while 1                % forever loop
    xnew = B*x + c;
    if norm(xnew-x) < tol
        break;
    end
    x = xnew;
    k = k+1;
end

k, x, norm(A*x-b)    % print & verify solution
```

norm() measures the difference between the vectors **xnew** and **x**,  
could also use  $L_\infty$  norm  
**max (abs (xnew-x) )**

Note that it breaks out of the loop just before the tolerance is reached,  
one more iteration after the loop would meet or exceed the tolerance.

## MATLAB implementation – v.2

```
% in order to save individual iterates x(k) ,  
% use a matrix X whose columns are x(k)  
  
tol = 1e-10;           % error tolerance  
X(:,1)=x0; k=1;       % x0 = arbitrary Nx1 vector  
  
while 1  
    X(:,k+1) = B*X(:,k) + c;  
    if norm(X(:,k+1)-X(:,k)) < tol  
        break;  
    end  
    k = k+1;  
end  
  
k, x = X(:,k)          % x = converged solution  
norm(A*x-b)            % verify solution
```

## MATLAB implementation – v.3

```
% another version that saves the iterates

tol=1e-10;      % error tolerance
x=x0; k=1;       % x0 = arbitrary Nx1 vector
X=x;             % X columns are the iterates x(k)

while 1
    xnew = B*x + c;
    if norm(xnew-x) < tol
        break;
    end
    x = xnew;
    k = k+1;
    X = [X,x];          % append new column to X
end

k, x, norm(A*x-b)    % print & verify solution
```

```
A = [2 1 0; 1 5 -1; 1 -2 4];
```

```
b = [4 8 9]';
```

```
xsol = A\b;
```

```
>> A, b, xsol
```

Example

$$\begin{bmatrix} 2 & 1 & 0 \\ 1 & 5 & -1 \\ 1 & -2 & 4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 4 \\ 8 \\ 9 \end{bmatrix}$$

<b>A =</b>	<b>b =</b>	<b>xsol =</b>
2 1 0	4	1
1 5 -1	8	2
1 -2 4	9	3

```
I = eye(size(A)); % 3x3 identity matrix
```

```
D = diag(diag(A)); % diagonal part of A
```

```
B = I - D\A; % 3x3 iteration matrix
```

```
c = D\b; % 3x1 vector
```

```
rho = max(abs(eig(B)));
```

```

>> D, B                               >> c, rho, norm(B,inf)

D =
  2  0  0
  0  5  0
  0  0  4

B =
  0.00 -0.50  0.00
 -0.20  0.00  0.20
 -0.25  0.50  0.00

c =
  2.00
  1.60
  2.25

rho =
  0.5000
ans =
  0.7500

```

$$\mathbf{x}(k+1) = B\mathbf{x}(k) + \mathbf{c}, \quad k = 0, 1, 2, \dots$$

$$\begin{bmatrix} x_1(k+1) \\ x_2(k+1) \\ x_3(k+1) \end{bmatrix} = \begin{bmatrix} 0.00 & -0.50 & 0.00 \\ -0.20 & 0.00 & 0.20 \\ -0.25 & 0.50 & 0.00 \end{bmatrix} \begin{bmatrix} x_1(k) \\ x_2(k) \\ x_3(k) \end{bmatrix} + \begin{bmatrix} 2.00 \\ 1.60 \\ 2.25 \end{bmatrix}$$

```

tol=1e-12;                      % error tolerance
x=[0 1 2]'; k=1;                % x0 = [0 1 2]' = arbitrary
X=x;                            % X = save all iterates

while 1
    xnew = B*x + c;
    if norm(xnew-x)<tol        % norm() measures
        break;                  % the distance
    end                         % between x,xnew
    x = xnew;
    k = k+1;
    X = [X,x];                 % append new column
end

k, x, norm(A*x-b), norm(x-xsol), size(X)

```

```
>> k, x, norm(A*x-b), norm(x-xsol), size(X)

k =
40

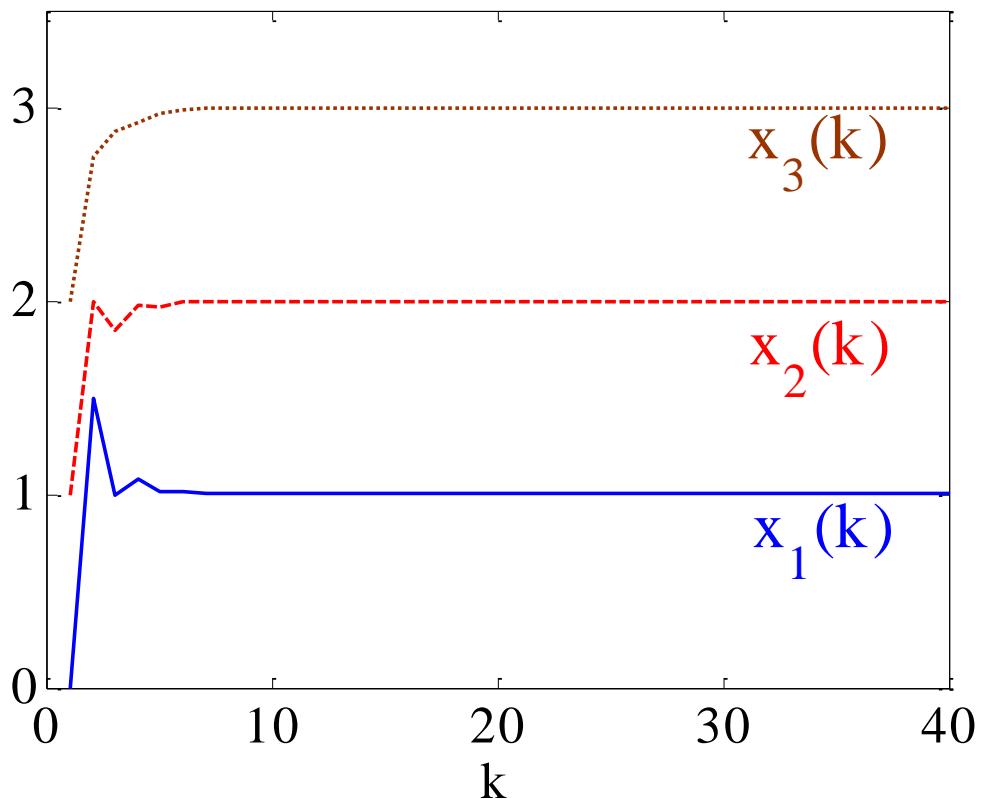
x =
1.0000
2.0000
3.0000

ans =
2.6657e-012

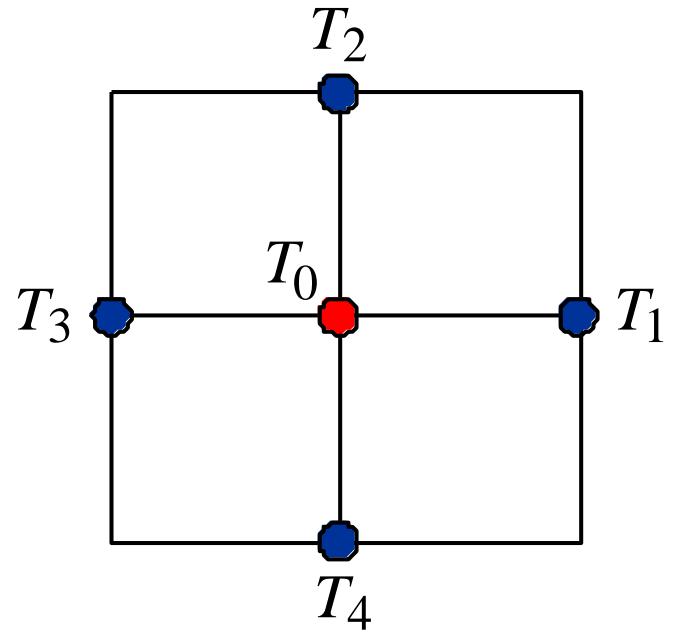
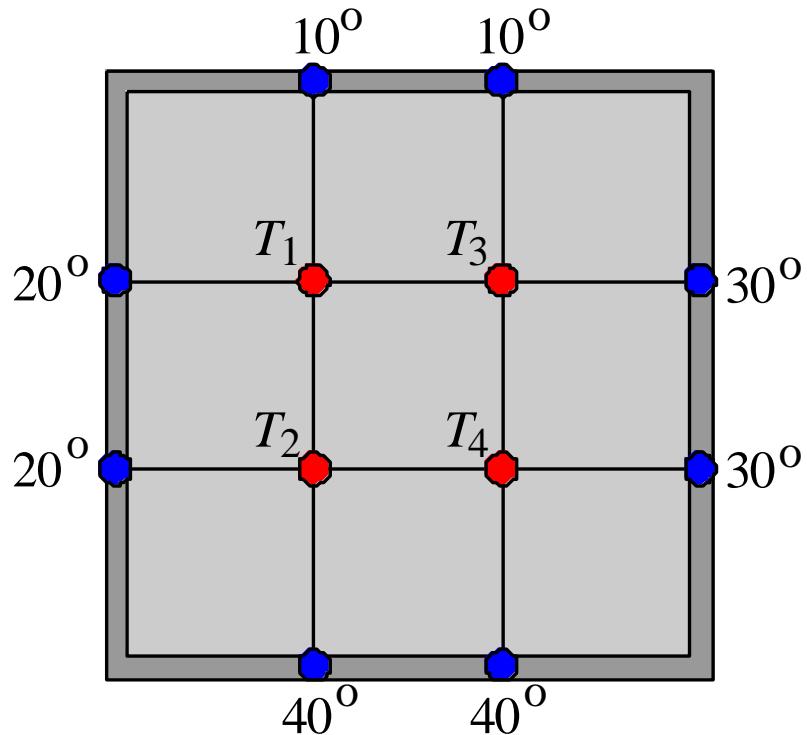
ans =
1.3634e-012

ans =
3      40

>> K=1:k;
>> plot(K,X');
```



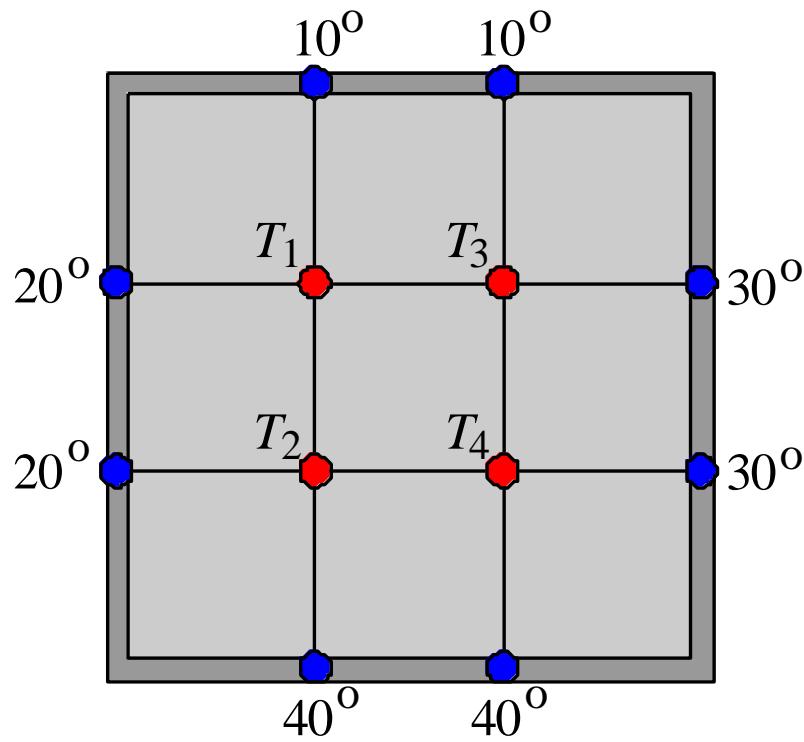
# Temperature Distribution



$$T_0 = \frac{1}{4} (T_1 + T_2 + T_3 + T_4)$$

follows from discretizing  
the Laplace equation

$$\nabla^2 T = \frac{\partial^2 T}{\partial x^2} + \frac{\partial^2 T}{\partial y^2} = 0$$



$$T_1 = \frac{1}{4}(10 + 20 + T_2 + T_3)$$

$$T_2 = \frac{1}{4}(20 + 40 + T_1 + T_4)$$

$$T_3 = \frac{1}{4}(10 + 30 + T_1 + T_4)$$

$$T_4 = \frac{1}{4}(30 + 40 + T_2 + T_3)$$

$$4T_1 - T_2 - T_3 = 30$$

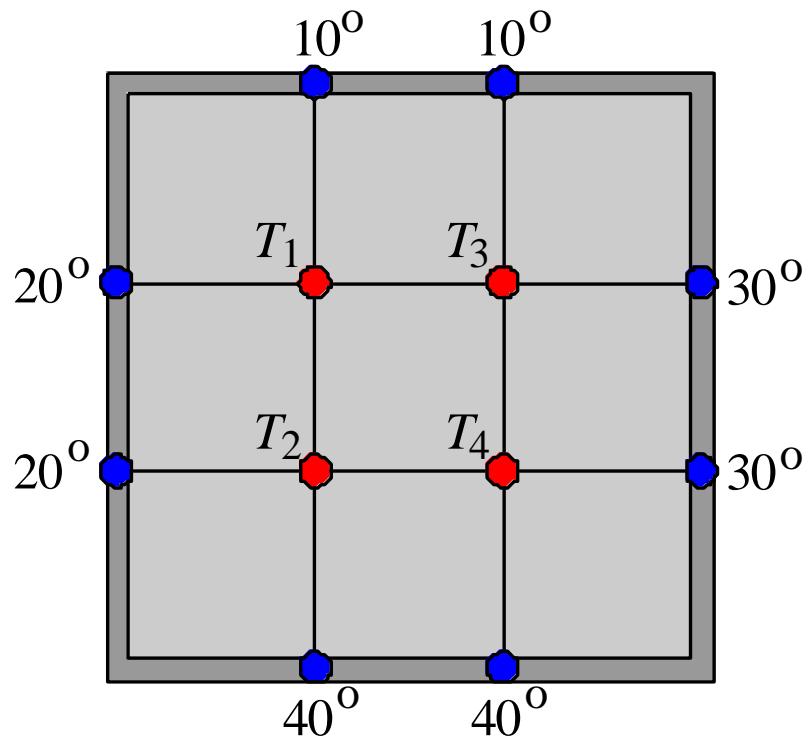
$$4T_2 - T_1 - T_4 = 60$$

$$4T_3 - T_1 - T_4 = 40$$

$$4T_4 - T_2 - T_3 = 70$$

$$\begin{bmatrix} 4 & -1 & -1 & 0 \\ -1 & 4 & 0 & -1 \\ -1 & 0 & 4 & -1 \\ 0 & -1 & -1 & 4 \end{bmatrix} \begin{bmatrix} T_1 \\ T_2 \\ T_3 \\ T_4 \end{bmatrix} = \begin{bmatrix} 30 \\ 60 \\ 40 \\ 70 \end{bmatrix}$$

$$\mathbf{A} \mathbf{x} = \mathbf{b}$$

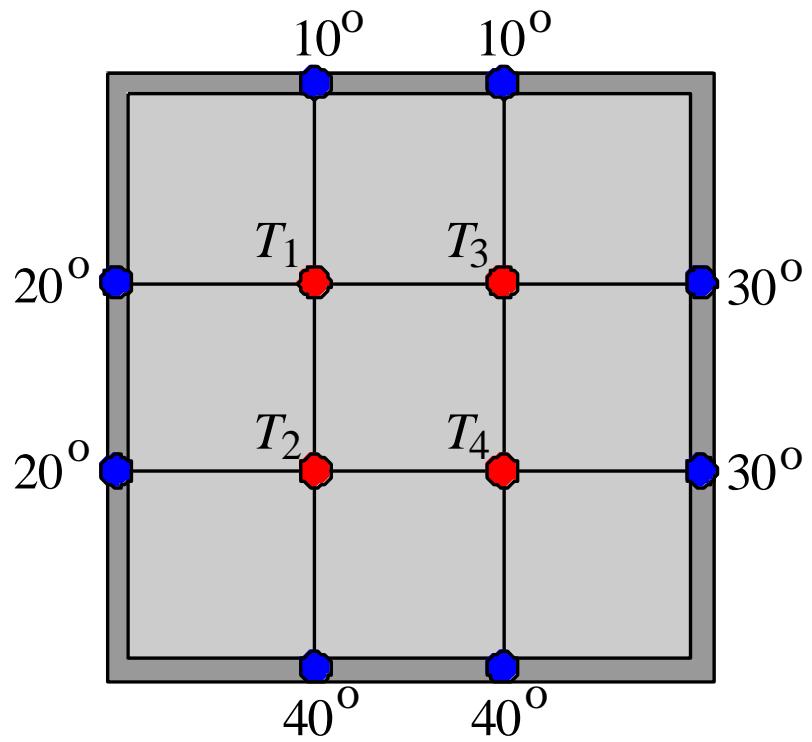


$$\begin{bmatrix} 4 & -1 & -1 & 0 \\ -1 & 4 & 0 & -1 \\ -1 & 0 & 4 & -1 \\ 0 & -1 & -1 & 4 \end{bmatrix} \begin{bmatrix} T_1 \\ T_2 \\ T_3 \\ T_4 \end{bmatrix} = \begin{bmatrix} 30 \\ 60 \\ 40 \\ 70 \end{bmatrix}$$

$\mathbf{A} \mathbf{x} = \mathbf{b}$

```
>> A = [ 4 -1 -1 0
          -1 4 0 -1
          -1 0 4 -1
          0 -1 -1 4];
>> b = [30; 60; 40; 70];
```

```
>> x = A\b
x =
20.0000
27.5000
22.5000
30.0000
```

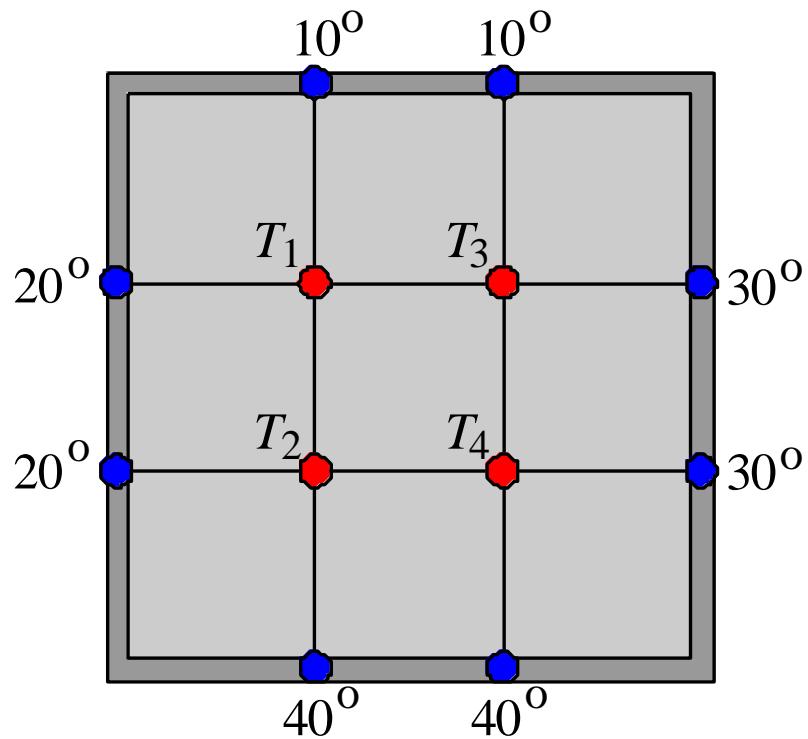


display solution **T** in a  
rectangular pattern

nodes were numbered in  
column order

```
T = zeros(2,2); % shape of T
T(:) = x

T =
    20.0000    22.5000
    27.5000    30.0000
```

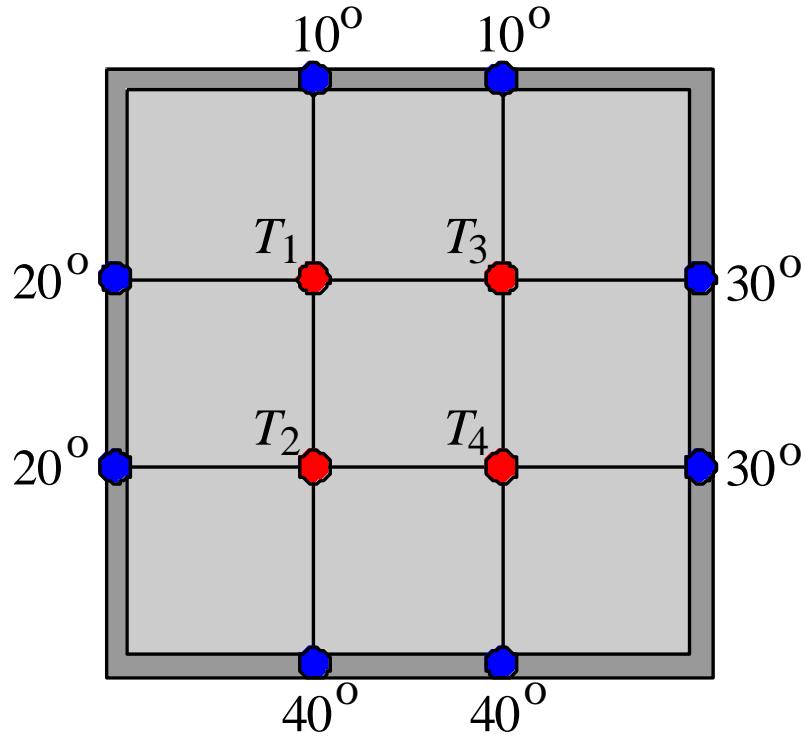


$$\begin{bmatrix} 4 & -1 & -1 & 0 \\ -1 & 4 & 0 & -1 \\ -1 & 0 & 4 & -1 \\ 0 & -1 & -1 & 4 \end{bmatrix} \begin{bmatrix} T_1 \\ T_2 \\ T_3 \\ T_4 \end{bmatrix} = \begin{bmatrix} 30 \\ 60 \\ 40 \\ 70 \end{bmatrix}$$

$\mathbf{A} \mathbf{x} = \mathbf{b}$

Rules for constructing  $\mathbf{A}$  and  $\mathbf{b}$ :

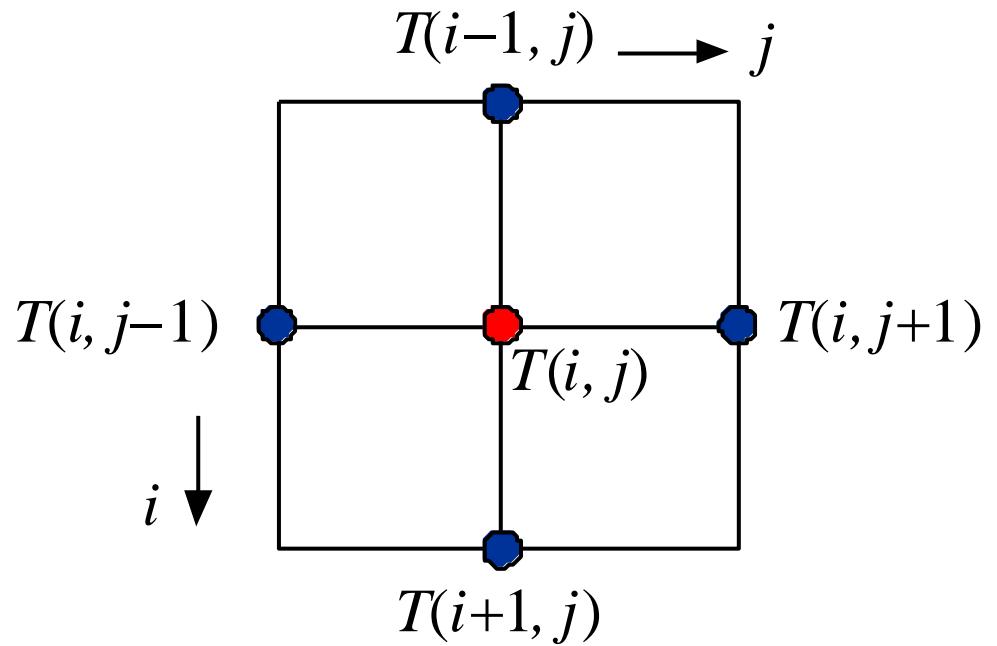
- 1) main diagonal is 4
- 2) if nodes  $i, j$  are connected, then, -1, otherwise, 0
- 3)  $\mathbf{b}(i)$  is sum of boundary values connected to node  $i$



used also to solve 2D electrostatics problems

## Iterative Solution

convenient for large number of subdivisions



$$T(i, j) = \frac{1}{4} [T(i + 1, j) + T(i - 1, j) + T(i, j + 1) + T(i, j - 1)]$$

```

N=4; M=4;
left=20; right=30; up=10; dn=40;                                boundary values
T(1,:) = repmat(up,1,M);    T(N,:) = repmat(dn,1,M);
T(:,1) = repmat(left,N,1); T(:,M) = repmat(right,N,1);

Tnew = T;

tol = 1e-4; K = 100;
for k=1:K,
    for i=2:N-1,           ← iterate over internal nodes only
        for j=2:M-1,
            Tnew(i,j) = (T(i-1,j) + T(i+1,j) +...
                            T(i,j-1) + T(i,j+1))/4;
        end
    end
    if norm(Tnew-T) < tol, break; end
    T = Tnew;
end

T(1,[1,end]) = nan; T(end,[1,end]) = nan;

```

T = % start-up

NaN	10	10	NaN
20	0	0	30
20	0	0	30
NaN	40	40	NaN

% converged after k = 19 iterations

% to within the specified tol = 1e-4

T =

NaN	10.0000	10.0000	NaN
20.0000	19.9999	22.4999	30.0000
20.0000	27.4999	29.9999	30.0000
NaN	40.0000	40.0000	NaN

T =	% after k=1 iteration			
	NaN	10.0000	10.0000	NaN
	20.0000	7.5000	10.0000	30.0000
	20.0000	15.0000	17.5000	30.0000
	NaN	40.0000	40.0000	NaN
T =	% after k=2 iterations			
	NaN	10.0000	10.0000	NaN
	20.0000	13.7500	16.2500	30.0000
	20.0000	21.2500	23.7500	30.0000
	NaN	40.0000	40.0000	NaN
T =	% after k=3 iterations			
	NaN	10.0000	10.0000	NaN
	20.0000	16.8750	19.3750	30.0000
	20.0000	24.3750	26.8750	30.0000
	NaN	40.0000	40.0000	NaN

```

N=30; M=30;
left=0; right=0; up=0; dn=60;
tol = 1e-6; K = 5000;

% breaks out at k = 2475

[X,Y] = meshgrid(2:M-1, 2:N-1);
Z = T(2:M-1, 2:N-1);
surf(X,Y,Z);

```

