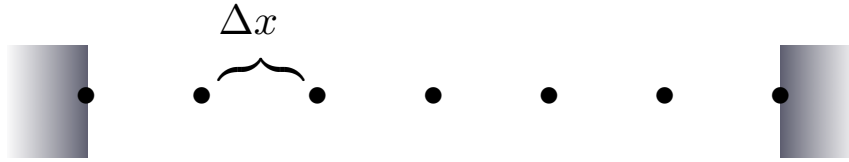


2. Matrix approach to B.V.P. : Implicit methods

How does this become a linear algebra (matrix · vector = vector) problem?

Consider $d^2_x T - T = 0$, $T(0) = T_0$, $T(1) = T_1$.



- Set up grid -- here, constant Δx .
- Store solution in array $[T(1), T(2), T(3), \dots] = T(I)$.
- Linear algebra ($\mathbf{A} \cdot \mathbf{T} = \mathbf{B}$).

Setting up the \mathbf{A} matrix:

Say $d^2_x T - T = 0$ with $T(0) = T_0$; $T(1) = T_1$. Recall second order central

$$\frac{\delta^2 T_i}{\delta x^2} = \frac{T_{i+1} - 2T_i + T_{i-1}}{\Delta x^2}$$

($T(i), i=1 \dots I$) i.e. $T(i) = \mathbf{T}$

As a vector product: column solution vector

$$\frac{\delta^2 T_i}{\delta x^2} = \left(\underbrace{0, 0, \dots, 0}_{i-2}, \frac{1}{\Delta x^2}, \frac{-2}{\Delta x^2}, \frac{1}{\Delta x^2}, \underbrace{0, \dots, 0}_{I-i-1} \right) \cdot \mathbf{T}$$

i^{th} row of matrix

Stack this for all $i=2, I-1$. B.c.s on first and last row (add -1 to diag for diff eq.)

- Solve simultaneous equations for $T(i)$ including boundary conditions

Store only the 3 diagonal elements, not the whole matrix

$$\begin{pmatrix} \bullet & 0 & 0 & 0, 0, \dots, 0, 0 \\ \frac{1}{\Delta x^2}, \frac{-2}{\Delta x^2} - 1, \frac{1}{\Delta x^2}, 0, 0, \dots, 0, 0 \\ 0, 0, \dots, 0, \frac{1}{\Delta x^2}, \frac{-2}{\Delta x^2} - 1, \frac{1}{\Delta x^2}, 0, \dots, 0, 0 \\ 0, 0, \dots, 0, \frac{1}{\Delta x^2}, \frac{-2}{\Delta x^2} - 1, \frac{1}{\Delta x^2}, 0, \dots, 0, 0 \\ 0, 0, \dots, 0, \frac{1}{\Delta x^2}, \frac{-2}{\Delta x^2} - 1, \frac{1}{\Delta x^2}, 0, \dots, 0, 0 \\ \vdots \\ 0, 0, 0, \dots, 0, 0, \bullet \end{pmatrix} \begin{pmatrix} T_0 \\ 0 \\ 0 \\ \vdots \\ \vdots \\ \vdots \\ T_1 \end{pmatrix}$$

Boundary conditions first and last rows

$A_{i,i-1}=1/\Delta x^2$, $A_{i,i}=-2/\Delta x^2 - 1$, $A_{i,i+1}=1/\Delta x^2$; all other $A_{i,j}$'s are 0: matrix of size $A(N, 3)$

Note first and last row have only two elements inside tridiag-structure. (Another element could be handled by modifying Gauss elimination.)

Matrix for $d^2_x T - T = 0$

$$B(2:I-1) = 0$$

$$A(2:I-1, 1) = 1/\Delta x^2$$

$$A(2:I-1, 2) = -2/\Delta x^2 - 1$$

$$A(2:I-1, 3) = 1/\Delta x^2$$

$$\text{b.c.'s: } A(1, 2) = ?$$

$$A(1, 3) = ?$$

$$A(N, 1) = ?$$

$$A(N, 2) = ?$$

$$B(1) = ?$$

$$B(N) = ?$$

What goes into the B.C. rows?

If $T(1) = T_0$ then first row is

$$(1, 0, 0 \dots 0) \text{ or } A(1, 2)=1, A(1, 3)=0$$

because $(1, 0, 0 \dots 0) \cdot (T(1), T(2), \dots) = T(1)$ so

$$B(1)=T_0$$

Or, if heat flux is given, using one-sided, first order difference

If $\delta T / \delta x(1) = Q$ then first row is

$$(-1/\Delta x, 1/\Delta x, 0 \dots 0) \text{ or } A(1, 2)=-1/\Delta x, A(1, 3)=1/\Delta x \text{ and}$$

$$B(1)=Q$$

because $(-1/\Delta x, 1/\Delta x, 0 \dots 0) \cdot (T(1), T(2), \dots) = (T(2) - T(1))/\Delta x$

Similarly for last row $A(l, 2)=1, A(l, 3)=0, B(l)=T_1$

Formally $A \cdot T = B$ is solved: $T = A^{-1} \cdot B$.

However, *almost never invert a matrix explicitly. Will review Gauss elimination*

3. Gauss elimination; Thomas algorithm for tri-diagonal matrices

- a. Reminder on linear algebra. System of equations $\mathbf{A} \cdot \mathbf{x} = \mathbf{b}$ <- includes b.c.
E.g.

$$\mathbf{A} = \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \quad \mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \quad \mathbf{b} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$$

$$\begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} a_{11}x_1 + a_{12}x_2 \\ a_{21}x_1 + a_{22}x_2 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$$

Formal solution $\mathbf{x} = \mathbf{A}^{-1} \cdot \mathbf{b}$. In CFD \mathbf{A} might be $10^6 \times 10^6 = \#$ grid points. \mathbf{A}^{-1} very expensive to compute; rarely done. But, don't have to find \mathbf{A}^{-1} , just solve equations.

Will see: for large systems even this is done approximately. Gauss elimination is an exact method.

- b. Reminder on Gauss elimination

e.g. $x_1 + x_2 + x_3 = 0$

$$2x_1 + 4x_2 + 3x_3 = 1$$

$$2x_1 + 2x_2 + x_3 = 2$$

$$\begin{pmatrix} 1 & 1 & 1 \\ 2 & 4 & 3 \\ 2 & 2 & 1 \end{pmatrix} \cdot \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}$$

Convert to lower triangular form \mathbf{A} \mathbf{b}

$$\begin{pmatrix} 1 & 1 & 1 \\ 2 & 4 & 3 \\ 2 & 2 & 1 \end{pmatrix} \quad \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}$$

$$\begin{pmatrix} 1 & 1 & 0 \\ 4 & 2 & 0 \\ 2 & 2 & 1 \end{pmatrix} \quad \begin{pmatrix} 2 \\ 5 \\ 2 \end{pmatrix}$$

$$\begin{pmatrix} 2 & 0 & 0 \\ 4 & 2 & 0 \\ 2 & 2 & 1 \end{pmatrix} \quad \begin{pmatrix} 1 \\ 5 \\ 2 \end{pmatrix}$$

Back substitute

$$\begin{array}{rcl}
 2x_1 & = & 1 \\
 4x_1 + 2x_2 & = & 5 \\
 2x_1 + 2x_2 + x_3 & = & 2
 \end{array}
 \qquad
 \begin{array}{rcl}
 \rightarrow x_1 & = & 1/2 \\
 \rightarrow x_2 & = & 3/2 \\
 \rightarrow x_3 & = & -2
 \end{array}$$

Schematic of elimination step

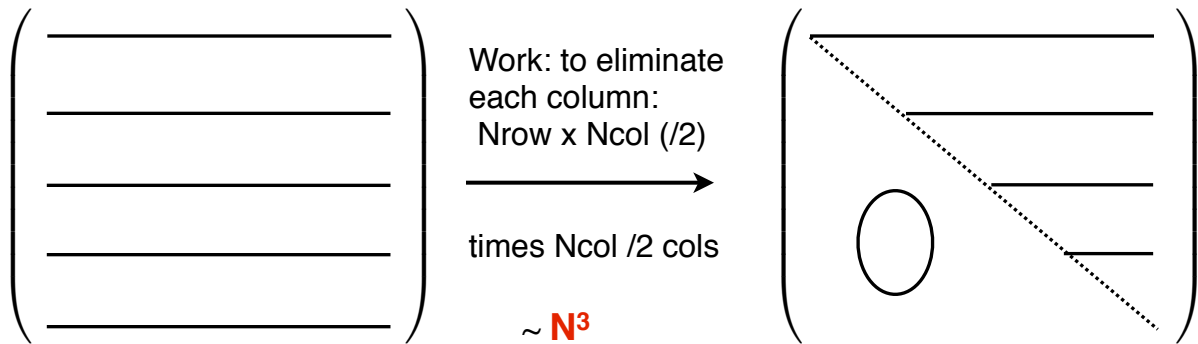
$$\begin{pmatrix} \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \end{pmatrix} \rightarrow \begin{pmatrix} \bullet & \bullet & 0 \\ \bullet & \bullet & 0 \\ \bullet & \bullet & \bullet \end{pmatrix} \rightarrow \begin{pmatrix} \bullet & 0 & 0 \\ \bullet & \bullet & 0 \\ \bullet & \bullet & \bullet \end{pmatrix}$$

Or, sometimes think of it like this:

$$\begin{array}{c}
 \text{LD} \quad \begin{matrix} \text{D} & \text{UD} \end{matrix} \\
 \begin{pmatrix} \diagdown & \diagdown & \bullet \\ \bullet & \diagdown & \diagdown \\ \bullet & \bullet & \diagdown \end{pmatrix} \rightarrow \begin{pmatrix} \diagdown & \bullet & \\ \bullet & \diagdown & 0 \\ \bullet & \bullet & \diagdown \end{pmatrix} \rightarrow \begin{pmatrix} \diagdown & & 0 \\ \bullet & \diagdown & \\ \bullet & \bullet & \diagdown \end{pmatrix}
 \end{array}$$

Main, upper and lower diagonals Lower triangular (L)
(Upper triangular, U; c.f. LU)

c. Schematic of complexity (explain representation)



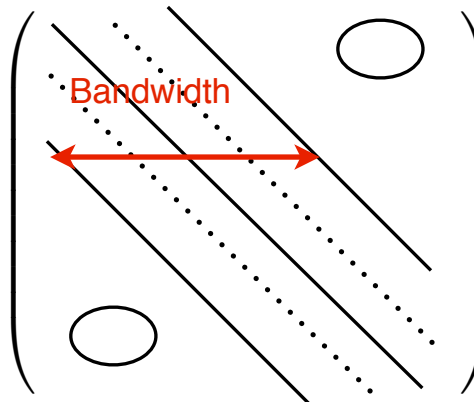
N^3 operations means 3 nested loops from 1 to N ; i.e.,

```
DO nrow=1,N      ! Use the elements of current row
  DO i=nrow,N rows ! To eliminate whole column
    DO j=nrow,N cols
      operate on A(i,j)
```

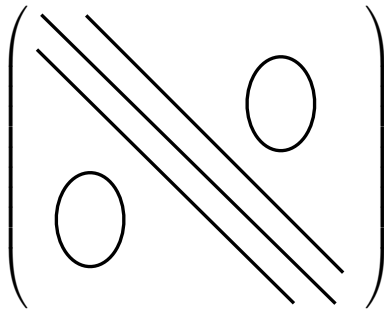
$N \times N/2 \times N/2 \sim N^3$ Hence work increases as size of matrix cubed: *computational complexity* in CFD, N may be $\sim 10^6$ So work $\sim 10^{18}$ ops

d. Simplifications: special form of matrix

Sparse - mostly 0's $< N^3$
 Banded - 0's outside diagonal band $N \times BW^2$
 Tridiagonal N (BW=3)



a. Thomas Algorithm for tridiagonal (handout posted on BlackBoard)


 $A_{ij}=0 \text{ for } j \neq i \pm 1$
 $A(i,3)$ stores matrix

Solve

$$\begin{pmatrix} A_{12} & A_{13} & 0 & 0 & 0 & \dots \\ A_{21} & A_{22} & A_{23} & 0 & 0 & \dots \\ 0 & A_{31} & A_{32} & A_{33} & 0 & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & A_{N-1,1} & A_{N-1,2} & A_{N-1,3} \\ 0 & 0 & 0 & 0 & A_{N,1} & A_{N,2} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \dots \\ x_N \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \\ \dots \\ b_N \end{pmatrix}$$

by eliminating upper diagonal and back substituting.

```

!*****
SUBROUTINE TDAG(a,b,N)
!*****
  REAL :: A(N,3),b(N),x(n)
!----
! eliminate A(*,3)
!----
  DO i=N-1,1,-1
    fac = A(i,3)/A(i+1,2)
    A(i,2) = A(i,2)-fac*A(i+1,1)
    b(i) = b(i)-fac*b(i+1)
  ENDDO
!----
! Now A is lower triangular. Back substitution
!----
  x(1) = b(1)/A(1,2)
  DO j=2,N
    x(j) = (b(j)-A(j,1)*x(j-1))/A(j,2)
  ENDDO

  RETURN
END

```