

**CEE432/CEE532/MAE541**

**Developing Software for  
Engineering Applications**

**Lecture 11: Matrix Algebra  
(Chapter 10)**

# Terminology

- Types of arrays,  $\mathbf{A}_{m \times n}$ 
  - One-dimensional, vector,  $\mathbf{g}_{m \times 1}$
  - Square matrix,  $\mathbf{A}_{n \times n}$
  - Symmetric matrix,  $A_{ij} = A_{ji}$
  - Diagonal matrix
  - Identity matrix,  $\mathbf{I}_{n \times n}$
  - Upper Triangular Matrix,  $A_{ij} = 0, i > j$
  - Lower Triangular Matrix,  $A_{ij} = 0, i < j$

# Terminology

- Matrix Operations
  - Addition and subtraction,  $\mathbf{A}_{m \times n} = \mathbf{B}_{m \times n} \pm \mathbf{C}_{m \times n}$
  - Multiplication,  $\mathbf{A}_{m \times n} = \mathbf{B}_{m \times q} \mathbf{C}_{q \times n}$
  - Inverse (Avoid!!),  $\mathbf{A}^{-1}$
  - Transpose,  $\mathbf{A}^T$
  - Determinant (Avoid!!),  $\det(\mathbf{A})$

# Linear Algebraic Equations

$$\mathbf{A}_{m \times n} \mathbf{x}_{n \times 1} = \mathbf{b}_{m \times 1}$$

- Unique, non-trivial solution iff
  - $m=n$
  - $\det(\mathbf{A}) \neq 0$
  - $\mathbf{b}$  is not a null vector

# Solution Strategies

- Direct Solver
  - Gaussian Elimination
  - LU Factorization
  - $\text{LDL}^T$  (Cholesky) Factorization
- Iterative Solver
  - Preconditioned Conjugate Gradient Method

# Important Issues

- How much storage space will be used?
- How can numerically accurate solution can be generated?
- How much time will be taken to obtain the solution?
- How much of additional effort is needed if a solution is to be generated for a new right-hand side vector?

# Storage Scheme: Full

$$\begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{14} \\ A_{21} & A_{22} & A_{23} & A_{24} \\ A_{31} & A_{32} & A_{33} & A_{34} \\ A_{41} & A_{42} & A_{43} & A_{44} \end{bmatrix}$$

Storage requirement:  $4^2=16$  locations. In general, storage requirement =  $n^2$  locations

## Double Precision

$n=10^4 \rightarrow 800,000,000$  bytes

1 GB = 1,073,741,824 bytes

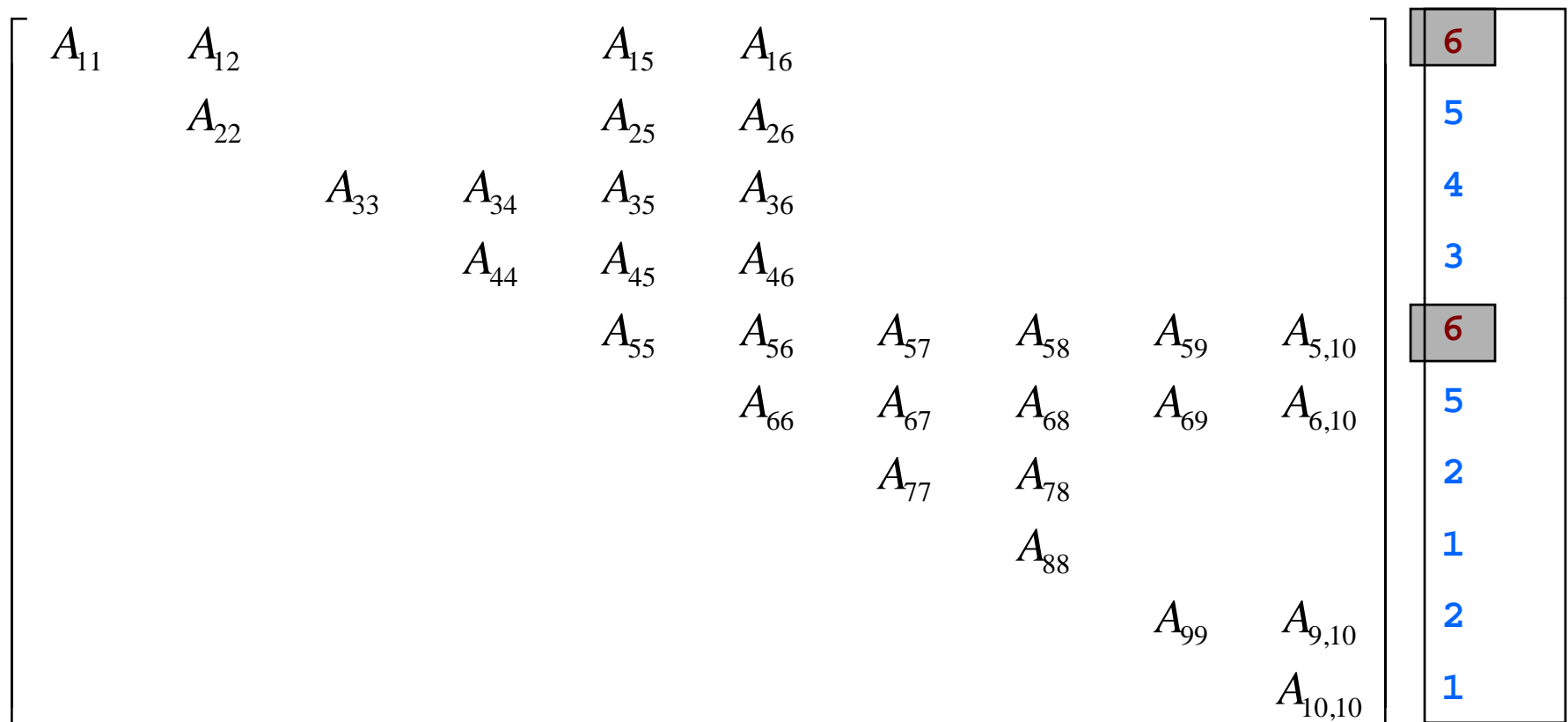
## Rowwise storage

$$\{ A_{11}, A_{12}, \dots, A_{1n}, A_{21}, A_{22}, \dots, A_{2n}, \dots, A_{n1}, A_{n2}, \dots, A_{nn} \}$$

## Columnwise storage

$$\{ A_{11}, A_{21}, \dots, A_{n1}, A_{12}, A_{22}, \dots, A_{n2}, \dots, A_{1n}, A_{2n}, \dots, A_{nn} \}$$

# Storage Scheme: Symmetric, Banded

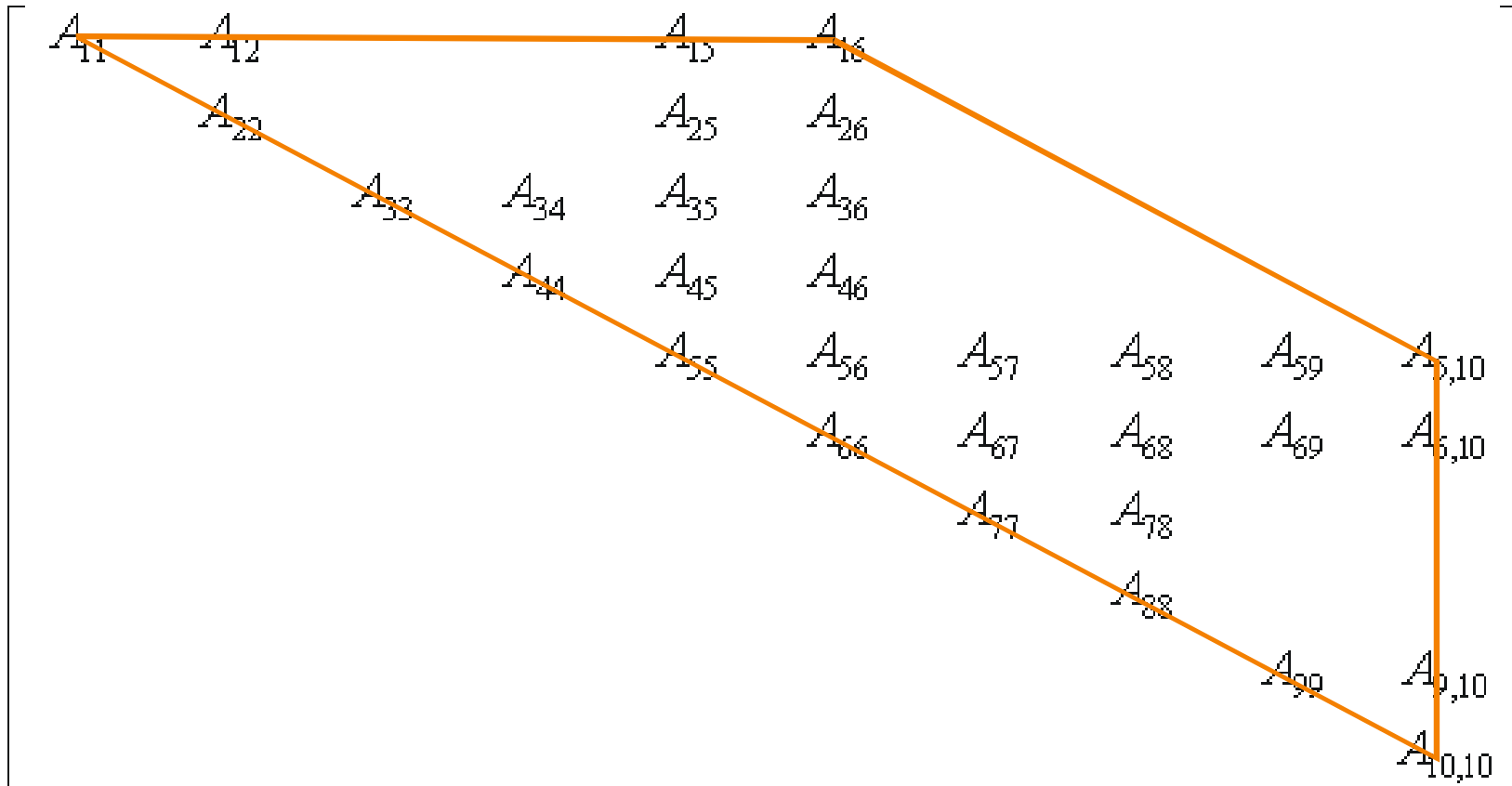


$$(HBW)_i = c - i + 1$$

$$HBW = \max_i (HBW)_i$$



# Storage Scheme: Symmetric, Banded



# Storage Scheme: Symmetric, Banded

$$\mathbf{A}_{10 \times 6}^{banded} = \begin{bmatrix} A_{11} & A_{12} & 0 & 0 & A_{15} & A_{16} \\ A_{22} & 0 & 0 & 0 & A_{25} & A_{26} \\ A_{33} & A_{34} & A_{35} & A_{36} & 0 & 0 \\ A_{44} & A_{45} & A_{46} & 0 & 0 & 0 \\ A_{55} & A_{56} & 0 & 0 & 0 & 0 \\ A_{66} & A_{67} & A_{68} & A_{69} & A_{6,10} & 0 \\ A_{77} & A_{78} & 0 & 0 & 0 & 0 \\ A_{88} & 0 & 0 & 0 & 0 & 0 \\ A_{99} & K_{9,10} & 0 & 0 & 0 & 0 \\ A_{10,10} & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

# Storage Scheme: Symmetric, Banded

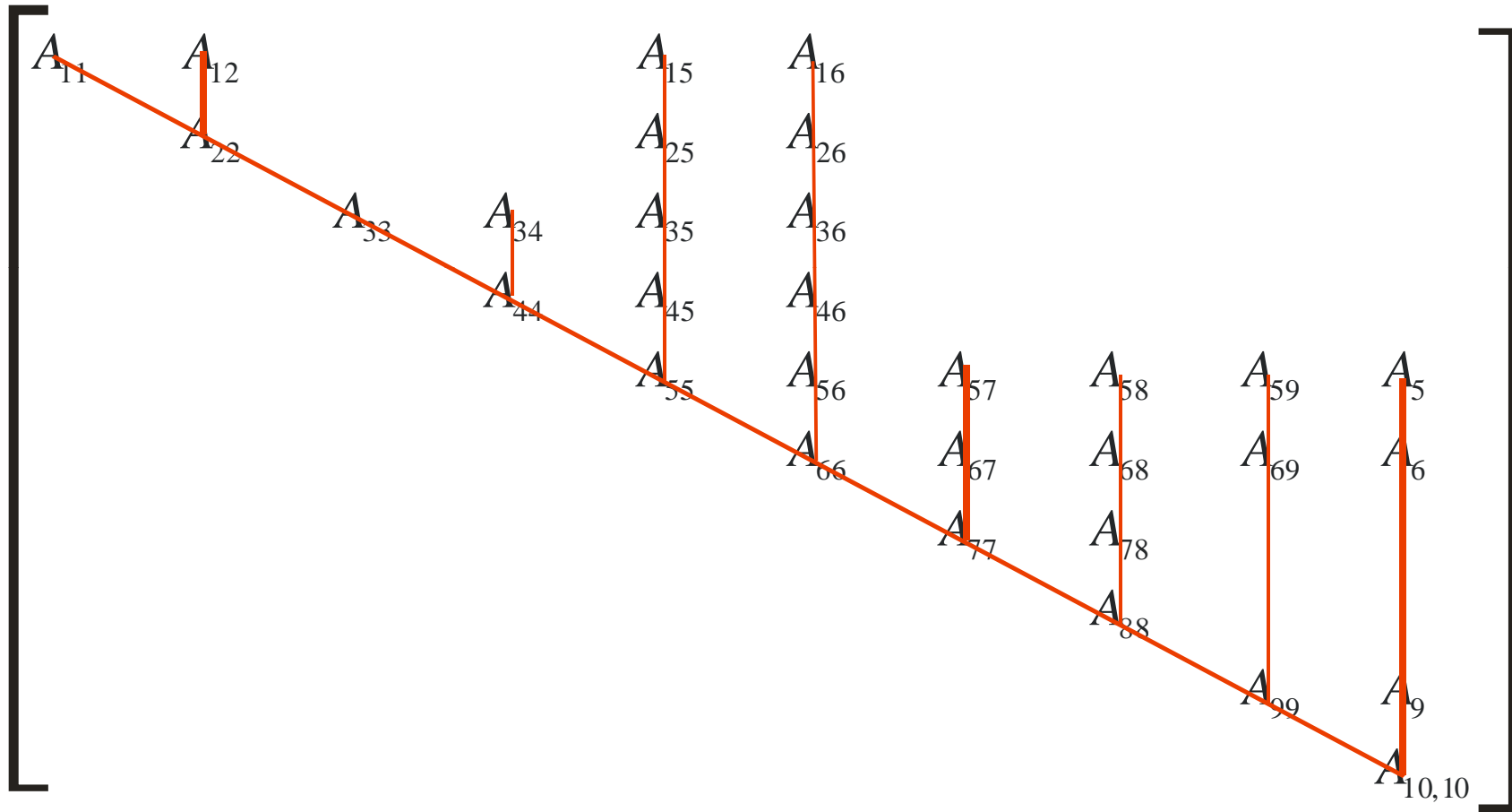
$$\begin{bmatrix} A_{11} & A_{12} & 0 & 0 & A_{15} & A_{16} \\ A_{22} & 0 & 0 & 0 & A_{25} & A_{26} \\ A_{33} & A_{34} & A_{35} & A_{36} & 0 & 0 \\ A_{44} & A_{45} & A_{46} & 0 & 0 & 0 \\ A_{55} & A_{56} & 0 & 0 & 0 & 0 \\ A_{66} & A_{67} & A_{68} & A_{69} & A_{6,10} & 0 \\ A_{77} & A_{78} & 0 & 0 & 0 & 0 \\ A_{88} & 0 & 0 & 0 & 0 & 0 \\ A_{99} & K_{9,10} & 0 & 0 & 0 & 0 \\ A_{10,10} & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$

$$A_{i,j} = 0 \text{ if } (j - i + 1) > HBW$$

$$A_{i,j} = 0 \text{ if } (j < i)$$

$$A_{i,j} \Rightarrow A_{i,j-i+1}^{banded}$$

# Storage Scheme: Symmetric, Skyline



# Storage Scheme: Symmetric, Skyline

$$\mathbf{A}_{35 \times 1}^{skyline} = \{A_{11}, A_{22}, A_{12}, A_{33}, A_{44}, A_{34}, \dots, A_{10,10}, A_{9,10}, A_{8,10}, A_{7,10}, A_{6,10}, A_{5,10}\}$$

$$\mathbf{D}_{11}^{loc} = \{1, 2, 4, 5, 7, 12, 18, 21, 25, 30, 36\}$$

$$A_{i,j} = 0 \text{ if } (j - i) > (D_{j+1}^{loc} - D_j^{loc})$$

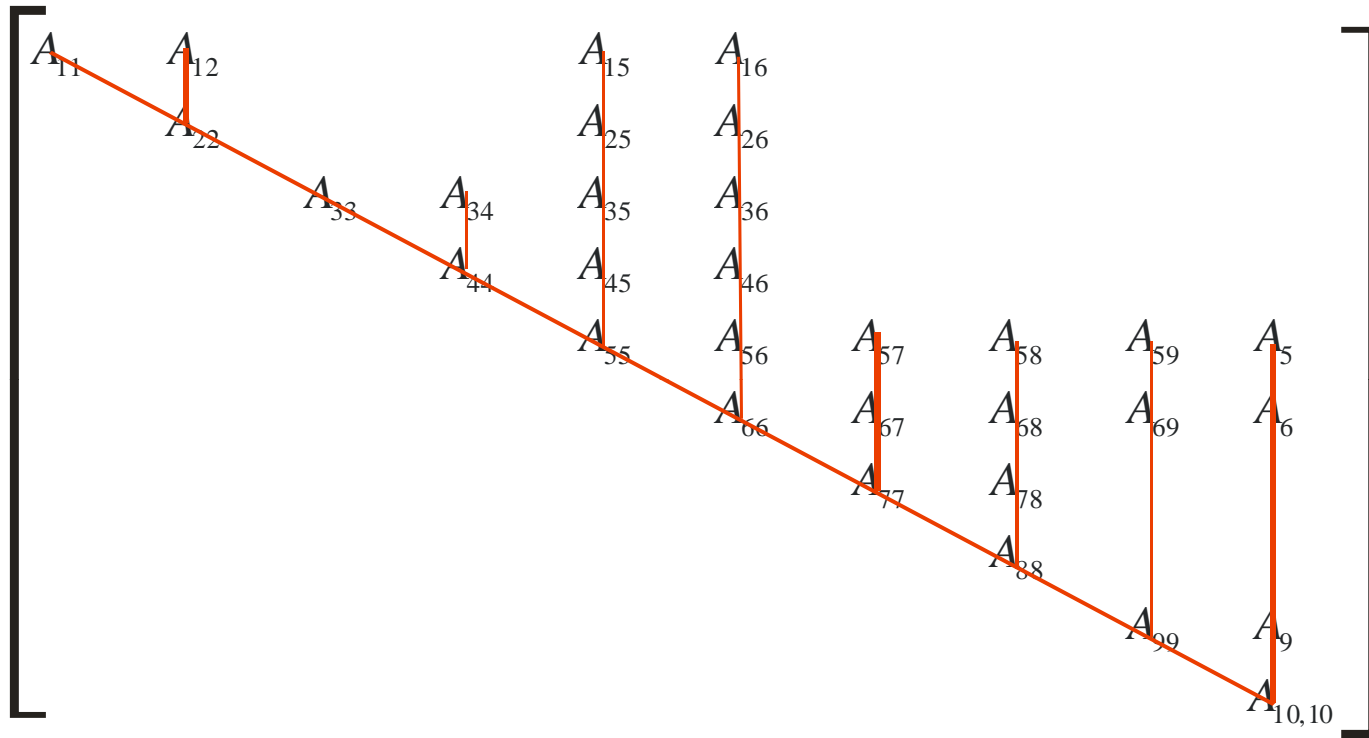
$$A_{i,j} = 0 \text{ if } (j < i)$$

$$A_{i,j} \Rightarrow l = D_j^{loc} + j - i \Rightarrow A_l^{skyline}$$

# Storage Scheme: Comparison

Storage Scheme	Storing	Equivalent Integer Words
Full	$\mathbf{A}_{n \times n}$	$2n^2$
Symmetric Banded	$\mathbf{A}_{n \times hbw}^{banded}$	$2nq$
Symmetric Skyline	$\mathbf{A}_m^{skyline}, \mathbf{D}_{n+1}^{loc}$	$2m + n + 1$

# Storage Scheme: Sparse



$$\mathbf{A}^{sparse}_{31 \times 1} = \{A_{11}, A_{12}, A_{15}, A_{16}, A_{22}, A_{25}, A_{26}, \dots, A_{88}, A_{99}, A_{9,10}, A_{10,10}\}$$

$$\mathbf{C}_{31 \times 1} = \{1, 2, 5, 6, 2, 5, 6, 3, 4, 5, 6, \dots, 8, 9, 10, 10\}$$

$$\mathbf{R}_{11 \times 1} = \{1, 5, 8, 12, 15, 21, 26, 28, 29, 31, 32\}$$

# Comparison w/ Example

Storage Scheme	Equivalent Integer Words*
Full	$10 \times 10 \times 2 = 200$
Banded (HBW=6)	$10 \times 6 \times 2 = 120$
Skyline	$35 \times 2 + 11 = 81$
Sparse	$31 \times 2 + 31 + 10 = 103$

\*Assuming elements of  $\mathbf{A}$  are *double*.



# Direct Solvers: Gaussian Elimination

- To obtain another equivalent  $\mathbf{Ax}=\mathbf{b}$ 
  - We can interchange two rows.
  - Multiply both sides by a constant.
  - Multiply one equation by a constant and add it to another equation.

$$\begin{bmatrix} \times & \times & \times \\ \times & \times & \times \\ \times & \times & \times \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} \times \\ \times \\ \times \end{Bmatrix} \Rightarrow \begin{bmatrix} \times & \times & \times \\ 0 & \times & \times \\ 0 & 0 & \times \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} \times \\ \times \\ \times \end{Bmatrix}$$

# Example

$$\begin{bmatrix} 8 & -1 \\ 4 & 7 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix} = \begin{Bmatrix} 6 \\ 18 \end{Bmatrix} \Rightarrow \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix} = \begin{Bmatrix} 1 \\ 2 \end{Bmatrix}$$



**Multiply first  
equation by (-4/8)  
and add to  
second**

$$\longrightarrow \begin{bmatrix} 8 & -1 \\ 0 & 7.5 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix} = \begin{Bmatrix} 6 \\ 15 \end{Bmatrix} \Rightarrow \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix} = \begin{Bmatrix} 1 \\ 2 \end{Bmatrix}$$

# Gaussian Elimination

## Forward Elimination

$$\begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{1i} & A_{1n} \\ A_{21} & A_{22} & A_{23} & A_{2i} & A_{2n} \\ A_{31} & A_{32} & A_{33} & A_{3i} & A_{3n} \\ A_{i1} & A_{i2} & A_{i3} & A_{ii} & A_{in} \\ A_{n1} & A_{n2} & A_{n3} & A_{ni} & A_{nn} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \\ x_i \\ x_n \end{Bmatrix} = \begin{Bmatrix} b_1 \\ b_2 \\ b_3 \\ b_i \\ b_n \end{Bmatrix}$$

$$\det(\mathbf{A}) = 0$$

$$|A_{ii}^{(i-1)}| \leq \varepsilon$$

$$\begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{1i} & A_{1n} \\ 0 & A_{22}^{(1)} & A_{23}^{(1)} & A_{2i}^{(1)} & A_{2n}^{(1)} \\ 0 & 0 & A_{33}^{(2)} & A_{3i}^{(2)} & A_{3n}^{(2)} \\ 0 & 0 & 0 & A_{ii}^{(i-1)} & A_{in}^{(i-1)} \\ 0 & 0 & 0 & 0 & A_{nn}^{(n-1)} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \\ x_i \\ x_n \end{Bmatrix} = \begin{Bmatrix} b_1 \\ b_2^{(1)} \\ b_3^{(2)} \\ b_i^{(i-1)} \\ b_n^{(n)} \end{Bmatrix}$$

# Gaussian Elimination

## Backward Substitution

$$\begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{1i} & A_{1n} \\ 0 & A_{22}^{(1)} & A_{23}^{(1)} & A_{2i}^{(1)} & A_{2n}^{(1)} \\ 0 & 0 & A_{33}^{(2)} & A_{3i}^{(2)} & A_{3n}^{(2)} \\ & & & & \\ 0 & 0 & 0 & A_{ii}^{(i-1)} & A_{in}^{(i-1)} \\ & & & & \\ 0 & 0 & 0 & 0 & A_{nn}^{(n-1)} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \\ \\ x_i \\ \\ x_n \end{Bmatrix} = \begin{Bmatrix} b_1 \\ b_2^{(1)} \\ b_3^{(2)} \\ \\ b_i^{(i-1)} \\ \\ b_n^{(n)} \end{Bmatrix}$$

$$x_n = \frac{b_n}{A_{nn}}$$

$$x_i = \frac{b_i - \sum_{j=i+1}^n A_{ij} x_j}{A_{ii}} \quad i = n-1, n-2, \dots, 1$$

# Matrix Toolbox

- Example Program 10.3.1
  - CMatrixToolBox class
  - Several matrix functionalities

## Gaussian Elimination Method

```
template <class T>
bool CMatToolBox<T>::AxEqb (CMatrix<T>& A,
                             CVector<T>& x,
                             CVector<T>& b,
                             T TOL)
```

# Client Code

```
CMatToolBox<double> MTBDP;    // double precision version
...
const int NUMEQNS=3;
const double TOL = 1.0e-6;
CMatrix<double> dMA(NUMEQNS,NUMEQNS);
CVector<double> dVx(NUMEQNS), dVb(NUMEQNS);
dMA(1,1) = 10.0;    dMA(1,2) = -5.0;    dMA(1,3) = 2.0;
dMA(2,1) = 3.0;    dMA(2,2) = 20.0;    dMA(2,3) = 5.0;
dMA(3,1) = -2.0;   dMA(3,2) = 7.0;    dMA(3,3) = 15.0;

dVb(1) = 6.0;      dVb(2) = 58.0;      dVb(3) = 57.0;
if (MTBDP.AxEqb (dMA, dVx, dVb, TOL))
{
    MTBDP.Display ("Vector x in Ax = b", dVx);
}
else
    std::cout << "Error in AxEqb.\n";
```

# Error Analysis

- How good is the “numerical” solution?

## Residual Vector

$$\mathbf{r} = \mathbf{Ax} - \mathbf{b} \neq \mathbf{0}$$

## Absolute Error

$$\mathcal{E}_{abs} = \|\mathbf{r}\|$$

## Residual Vector

$$\mathcal{E}_{rel} = \frac{\|\mathbf{r}\|}{\|\mathbf{b}\|}$$

Error magnitude is a function of the condition number of  $\mathbf{A}$ .

$$cond(\mathbf{A}) = \frac{\lambda_{\max}}{\lambda_{\min}}$$

# Improving the Solution

- Partial Pivoting
  - Locate the largest remaining element (in the current column) and switch rows.
- Full Pivoting
  - Locate the largest remaining element and switch rows and columns.
- Scaling
  - Multiply the equations by a constant so that the ratio of the largest to the smallest element in the matrix is reduced.



# LU Factorization

For a general  $\mathbf{A}$

$$\mathbf{A} = \mathbf{L}\mathbf{U}$$

**L:** lower triangular matrix

**U:** upper triangular matrix

$$\begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{14} \\ A_{21} & A_{22} & A_{23} & A_{24} \\ A_{31} & A_{32} & A_{33} & A_{34} \\ A_{41} & A_{42} & A_{43} & A_{44} \end{bmatrix} = \begin{bmatrix} 1 & & & \\ L_{21} & 1 & & \\ L_{31} & L_{32} & 1 & \\ L_{41} & L_{42} & L_{43} & 1 \end{bmatrix} \begin{bmatrix} U_{11} & U_{12} & U_{13} & U_{14} \\ & U_{22} & U_{23} & U_{24} \\ & & U_{33} & U_{34} \\ & & & U_{44} \end{bmatrix}$$

**Crout's Method**

# LU Factorization

$$\mathbf{A}_{n \times n} \mathbf{x}_{n \times 1} = \mathbf{b}_{n \times 1}$$

$$\mathbf{L}_{n \times n} \mathbf{U}_{n \times n} \mathbf{x}_{n \times 1} = \mathbf{b}_{n \times 1}$$

**Factorization**

$$\mathbf{A} = \mathbf{L}\mathbf{U}$$

**Forward Substitution**

$$\mathbf{L}_{n \times n} \mathbf{y}_{n \times 1} = \mathbf{b}_{n \times 1} \Rightarrow \text{Solve for } \mathbf{y}$$

**Backward Substitution**

$$\mathbf{U}_{n \times n} \mathbf{x}_{n \times 1} = \mathbf{y}_{n \times 1} \Rightarrow \text{Solve for } \mathbf{x}$$

# LU Factorization

## Storage Scheme

$$\begin{bmatrix} A_{11} & A_{12} & A_{13} & A_{14} \\ A_{21} & A_{22} & A_{23} & A_{24} \\ A_{31} & A_{32} & A_{33} & A_{34} \\ A_{41} & A_{42} & A_{43} & A_{44} \end{bmatrix} \Rightarrow \begin{bmatrix} U_{11} & U_{12} & U_{13} & U_{14} \\ L_{21} & U_{22} & U_{23} & U_{24} \\ L_{31} & L_{32} & U_{33} & U_{34} \\ L_{41} & L_{42} & L_{43} & U_{44} \end{bmatrix}$$

**Note**

$$j = 1, 2, 3, \dots, n$$

$$i = 1, 2, \dots, j \Rightarrow U_{ij} = A_{ij} - \sum_{k=1}^{i-1} L_{ik} U_{kj}$$

$$i = j+1, j+2, \dots, n \Rightarrow L_{ij} = \frac{A_{ij} - \sum_{k=1}^{j-1} L_{ik} U_{kj}}{U_{jj}}$$

# LU Factorization

## Storage Scheme

$$\left\{ \begin{array}{c} y_1 \\ y_2 \\ \dots \\ y_n \end{array} \right\} \Rightarrow \left\{ \begin{array}{c} x_1 \\ x_2 \\ \dots \\ x_n \end{array} \right\} : x_i = \frac{y_i - \sum_{j=i+1}^n U_{ij} x_j}{U_{ii}} \quad i = n-1, n-2, \dots, 1$$

# LU Factorization: Advantages

- Clearly separates factorization from forward and backward substitutions (helps solve multiple RHS vectors).
- No additional storage required.
- Easy to compute determinant

$$\det(\mathbf{A}) = \det(\mathbf{U}) = U_{11}U_{22} \cdots U_{nn}$$

# Example

$$\begin{bmatrix} 10 & -5 & 2 \\ 3 & 20 & 5 \\ -2 & 7 & 15 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} 6 \\ 58 \\ 57 \end{Bmatrix}$$

## Factorization

$$U_{11} = A_{11} = 10$$

$$\mathbf{j=1} \quad L_{21} = \frac{A_{21}}{U_{11}} = \frac{3}{10} = 0.3$$

$$L_{31} = \frac{A_{31}}{U_{11}} = \frac{-2}{10} = -0.2$$

**j=2**

$$U_{12} = A_{12} = -5$$

$$U_{22} = A_{22} - L_{21}U_{12} = 21.5$$

$$L_{32} = \frac{A_{32} - L_{31}U_{12}}{U_{22}} = 0.27907$$

**j=3**

$$U_{13} = A_{13} = 2$$

$$U_{23} = A_{23} - L_{21}U_{13} = 4.4$$

$$U_{33} = A_{33} - L_{31}U_{13} - L_{32}U_{23} = 14.1721$$

$$\mathbf{LU} = \begin{bmatrix} 10 & -5 & 2 \\ 0.3 & 21.5 & 4.4 \\ -0.2 & 0.27907 & 14.1721 \end{bmatrix}$$

# Example

## Forward Substitution

$$y_1 = \frac{b_1}{L_{11}} = 6$$

$$y_2 = \frac{b_2 - L_{21}y_1}{L_{22}} = 56.2$$

$$y_3 = \frac{b_3 - L_{31}y_1 - L_{32}y_2}{L_{33}} = 42.5163$$

## Backward Substitution

$$x_3 = \frac{y_3}{U_{33}} = 3$$

$$x_2 = \frac{y_2 - U_{23}x_3}{U_{22}} = 2$$

$$x_1 = \frac{y_1 - U_{12}x_2 - U_{13}x_3}{U_{11}} = 1$$

$$\det(A) = U_{11}U_{22}U_{33} = (10)(21.5)(14.1721) = 3047$$

# Cholesky Decomposition

For a symmetric, positive definite  $\mathbf{A}$

$$\mathbf{A} = \mathbf{LDL}^T \quad \mathbf{L}: \text{lower triangular matrix (1's on the diagonal)}$$

$\mathbf{D}$ : diagonal matrix

$$\begin{bmatrix} K_{11} & K_{12} & \cdot & K_{1n} \\ K_{12} & K_{22} & \cdot & K_{2n} \\ \cdot & \cdot & \cdot & \cdot \\ K_{1n} & K_{2n} & \cdot & K_{nn} \end{bmatrix} = \begin{bmatrix} 1 & 0 & \cdot & 0 \\ L_{21} & 1 & \cdot & 0 \\ \cdot & \cdot & \cdot & \cdot \\ L_{n1} & L_{n2} & \cdot & 1 \end{bmatrix} \begin{bmatrix} D_1 & 0 & \cdot & 0 \\ 0 & D_2 & \cdot & 0 \\ \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \cdot & D_n \end{bmatrix} \begin{bmatrix} 1 & L_{21} & \cdot & L_{n1} \\ 0 & 1 & \cdot & L_{n2} \\ \cdot & \cdot & \cdot & \cdot \\ 0 & 0 & \cdot & 1 \end{bmatrix}$$

$$= \begin{bmatrix} D_1 & D_1 L_{21} & D_1 L_{31} & \cdot & D_1 L_{n1} \\ & D_2 L_{21}^2 & D_1 L_{21} L_{31} + D_2 L_{32} & \cdot & D_1 L_{21} L_{n1} + D_2 L_{n2} \\ & & D_1 L_{31}^2 + D_2 L_{32}^2 + D_3 & \cdot & D_1 L_{31} L_{n1} + D_2 L_{32} L_{n2} + D_3 L_{n3} \\ & & & \cdot & \cdot \\ \text{sym} & & & & D_1 L_{n1}^2 + D_2 L_{n2}^2 + \dots + D_n \end{bmatrix}$$



# Cholesky Decomposition

$$\mathbf{A}_{n \times n} \mathbf{x}_{n \times 1} = \mathbf{b}_{n \times 1}$$

$$\mathbf{L}_{n \times n} \mathbf{D}_{n \times 1} \mathbf{L}_{n \times n}^T \mathbf{x}_{n \times 1} = \mathbf{b}_{n \times 1}$$

**Decomposition**

$$\mathbf{A} = \mathbf{L} \mathbf{D} \mathbf{L}^T$$

**Forward Substitution**

$$\mathbf{L}_{n \times n} \mathbf{y}_{n \times 1} = \mathbf{b}_{n \times 1} \Rightarrow \text{Solve for } \mathbf{y}$$

**Backward Substitution**

$$\mathbf{D}_{n \times n} \mathbf{L}_{n \times n}^T \mathbf{x}_{n \times 1} = \mathbf{y}_{n \times 1} \Rightarrow \text{Solve for } \mathbf{x}$$

**Note**

$$\mathbf{D} \mathbf{L}^T = \begin{bmatrix} D_1 & D_1 L_{12} & \cdot & D_1 L_{1n} \\ & D_2 & \cdot & D_2 L_{2n} \\ & & \cdot & \cdot \\ \mathbf{0} & & & D_n \end{bmatrix}$$

# Cholesky Decomposition: Advantages

- Clearly separates factorization from forward and backward substitutions (helps solve multiple RHS vectors).
- Can store all required elements (**L** and **D**) in the upper (or lower) triangular portion of **A**. No additional storage required.

# Example

$$\begin{bmatrix} 3.5120 & 0.7679 & 0 & 0 & 0 \\ 0.7679 & 3.1520 & 0 & -2 & 0 \\ 0 & 0 & 3.5120 & -0.7679 & 0.7679 \\ 0 & -2 & -0.7679 & 3.1520 & -1.1520 \\ 0 & 0 & 0.7679 & -1.1520 & 3.1520 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{Bmatrix} = \begin{Bmatrix} 0 \\ 0 \\ 0 \\ -0.04 \\ 0 \end{Bmatrix}$$

## Decomposition

**i=1**

$$D_1 = A_{11} = 3.5120$$

$$L_{21} = \frac{A_{21}}{D_1} = 0.21865$$

$$L_{31} = L_{41} = L_{51} = 0$$

**i=2**

$$D_2 = A_{22} - L_{21}^2 D_1 = 2.9841$$

$$L_{32} = \frac{A_{32} - L_{31} D_1 L_{21}}{D_2} = 0$$

$$L_{42} = \frac{A_{42} - L_{41} D_1 L_{21}}{D_2} = -0.670219$$

$$L_{52} = 0$$

# Example

**i=3**

$$D_3 = A_{33} - L_{31}^2 D_1 - L_{32}^2 D_2 = 3.5120$$

$$L_{43} = \frac{A_{43} - L_{41} D_1 L_{31} - L_{42} D_2 L_{32}}{D_3} = -0.21865$$

$$L_{53} = \frac{A_{53} - L_{51} D_1 L_{31} - L_{52} D_2 L_{32}}{D_3} = 0.21865$$

**i=4**

$$D_4 = A_{44} - L_{41}^2 D_1 - L_{42}^2 D_2 - L_{43}^2 D_3 = 1.64366$$

$$L_{54} = \frac{A_{54} - L_{51} D_1 L_{41} - L_{52} D_2 L_{42} - L_{53} D_3 L_{43}}{D_4} = -0.598724$$

# Example

**i=5**

$$D_5 = A_{55} - L_{51}^2 D_1 - L_{52}^2 D_2 - L_{53}^2 D_3 - L_{54}^2 D_4 = 2.3949$$

$$\mathbf{L} \text{ \& } \mathbf{D} \Rightarrow \left( \begin{bmatrix} 3.5120 & 0 & 0 & 0 & 0 \\ 0.21865 & 2.9841 & 0 & 0 & 0 \\ 0 & 0 & 3.5120 & 0 & 0 \\ 0 & -0.670219 & -0.21865 & 1.64366 & 0 \\ 0 & 0 & 0.21865 & -0.598724 & 2.3949 \end{bmatrix} \right)$$

# Forward Substitution

$$\mathbf{L}\mathbf{y} = \mathbf{b}$$

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0.21865 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & -0.670219 & -0.21865 & 1 & 0 \\ 0 & 0 & 0.21865 & -0.598724 & 1 \end{bmatrix} \begin{Bmatrix} y_1 \\ y_2 \\ y_3 \\ y_4 \\ y_5 \end{Bmatrix} = \begin{Bmatrix} 0 \\ 0 \\ 0 \\ -0.04 \\ 0 \end{Bmatrix}$$

$$y_1 = b_1 = 0$$

$$y_2 = b_2 - L_{21}y_1 = 0$$

$$y_3 = b_3 - L_{31}y_1 - L_{32}y_2 = 0$$

$$y_4 = b_4 - L_{41}y_1 - L_{42}y_2 - L_{43}y_3 = -0.04$$

$$y_5 = b_5 - L_{51}y_1 - L_{52}y_2 - L_{53}y_3 - L_{54}y_4 = -0.023949$$

# Backward Substitution

$$\mathbf{DL}^T \mathbf{D} = \mathbf{y}$$

$$\begin{bmatrix} 3.5120 & 0.21865 & 0 & 0 & 0 \\ 0 & 2.9841 & 0 & -0.670219 & 0 \\ 0 & 0 & 3.5120 & -0.21865 & 0.21865 \\ 0 & 0 & 0 & 1.64366 & -0.598724 \\ 0 & 0 & 0 & 0 & 2.3949 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{Bmatrix} = \begin{Bmatrix} 0 \\ 0 \\ 0 \\ -0.04 \\ -0.023949 \end{Bmatrix}$$

$$x_5 = \frac{y_5}{D_5} = -0.01$$

$$x_4 = \frac{y_4}{D_4} - L_{45}x_5 = -0.0303232$$

$$x_3 = \frac{y_3}{D_3} - L_{34}x_4 - L_{35}x_5 = -0.00444367$$

$$x_2 = \frac{y_2}{D_2} - L_{23}x_3 - L_{24}x_4 - L_{25}x_5$$

$$= -0.0203232$$

$$x_1 = \frac{y_1}{D_1} - L_{12}y_2 - L_{13}y_3 - L_{14}y_4 - L_{15}y_5$$

$$= 0.00444367$$

# Handling Constraints

**Solve equations with special conditions**

$$\mathbf{Ax} = \mathbf{b}$$

$$(a) \ x_i = c$$

$$(b) \ c_i x_i + c_j x_j = c$$



## Case (a)

$$\text{Solve } \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} b_1 \\ b_2 \\ b_3 \end{Bmatrix}$$

with  $x_2 = c$

### Solution

$$A_{11}x_1 + A_{12}x_2 + A_{13}x_3 = b_1$$

$$A_{21}x_1 + A_{22}x_2 + A_{23}x_3 = b_2$$

$$A_{31}x_1 + A_{32}x_2 + A_{33}x_3 = b_3$$

$$\Rightarrow$$

$$A_{11}x_1 + (0)x_2 + A_{13}x_3 = b_1 - A_{12}c$$

$$(0)x_1 + (1)x_2 + (0)x_3 = c$$

$$A_{31}x_1 + (0)x_2 + A_{33}x_3 = b_3 - A_{32}c$$

## Case (a)

$$\begin{bmatrix} A_{11} & 0 & A_{13} \\ 0 & 1 & 0 \\ A_{31} & 0 & A_{33} \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} b_1 - A_{12}c \\ c \\ b_3 - A_{32}c \end{Bmatrix}$$

**Algorithm**  $x_j = c$

1. Modify the right-hand side vector as  $b_i = b_i - A_{ij}c, i = 1, \dots, n$

2. Modify the coefficient matrix as  $A_{ij} = 0, i = 1, \dots, n$

$$A_{ji} = 0, i = 1, \dots, n$$

3. Set  $A_{jj} = 1$

# Example

$$\begin{bmatrix} 10 & -5 & 2 \\ 3 & 20 & 5 \\ -2 & 7 & 15 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} 6 \\ 58 \\ 57 \end{Bmatrix} \text{ with } x_2 = 3$$

## Modified Equations

$$\begin{bmatrix} 10 & 0 & 2 \\ 0 & 1 & 0 \\ -2 & 0 & 15 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} 6 + 5(3) \\ 3 \\ 57 - 7(3) \end{Bmatrix} = \begin{Bmatrix} 21 \\ 3 \\ 36 \end{Bmatrix}$$

## Case (b)

### Minimization Problem

$$\Pi(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x} - \mathbf{x}^T \mathbf{b}$$

leads to the solution to

$$\frac{\partial \Pi}{\partial \mathbf{x}} = 0 = \mathbf{A} \mathbf{x} - \mathbf{b}$$

## Case (b)

$$\Pi(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T \mathbf{A} \mathbf{x} - \mathbf{x}^T \mathbf{b} + \frac{1}{2} C \left( c_i x_i - c_j x_j - c \right)^2$$

 Large number

**Minimum is when**  $c_i D_i - c_j D_j - c = 0$

$$\frac{\partial \Pi}{\partial \mathbf{x}} = 0$$

$$C = 10^4 \max |A_{pq}|, 1 \leq p, q \leq n$$

## Case (b)

$$\begin{bmatrix}
 A_{11} & & A_{1i} & & A_{1j} & & A_{1n} \\
 & \ddots & & & & & \\
 A_{i1} & & A_{ii} + Cc_i^2 & & A_{ij} + Cc_i c_j & & A_{in} \\
 & & & \ddots & & & \\
 A_{j1} & & A_{ji} + Cc_i c_j & & A_{jj} + Cc_j^2 & & A_{jn} \\
 & & & & \ddots & & \\
 A_{n1} & & A_{ni} & & A_{nj} & & A_{nn}
 \end{bmatrix}
 \begin{Bmatrix}
 x_1 \\
 \\
 x_i \\
 \\
 x_j \\
 \\
 x_n
 \end{Bmatrix}
 =
 \begin{Bmatrix}
 b_1 \\
 \\
 b_i + Ccc_i \\
 \\
 b_j + Ccc_j \\
 \\
 b_n
 \end{Bmatrix}$$

# Example

$$\begin{bmatrix} 10 & -5 & 2 \\ -5 & 20 & 5 \\ 2 & 5 & 15 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} 6 \\ 58 \\ 57 \end{Bmatrix} \text{ with } 2x_1 + x_3 = 3$$

## Modified Equations

$$\begin{bmatrix} 10 + 20(10^4)2^2 & -5 & 2 + 20(10^4)(2)(1) \\ -5 & 20 & 5 \\ 2 + 20(10^4)(2)(1) & 5 & 15 + 20(10^4)1^2 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \\ x_3 \end{Bmatrix} = \begin{Bmatrix} 6 + 20(10^4)(3)(2) \\ 58 \\ 57 + 20(10^4)(3)(1) \end{Bmatrix}$$

# Matrix Toolbox

- Specifications