

Part 3

Chapter 8

Linear Algebraic Equations and Matrices

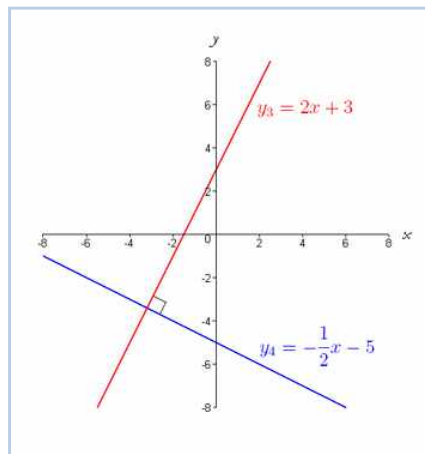
$$\begin{array}{rrcrcl} x & + & y & + & z & = & 6 \\ & & 2y & + & 5z & = & -4 \\ 2x & + & 5y & - & z & = & 27 \end{array}$$

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 2 & 5 \\ 2 & 5 & -1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} 6 \\ -4 \\ 27 \end{bmatrix}$$

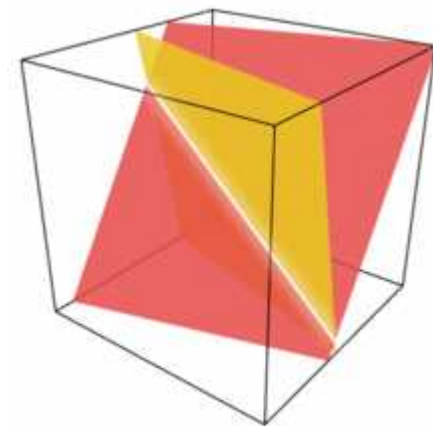
$$AX = B$$

Linear Algebraic Equations

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



solution set for two equations
in two variables : point

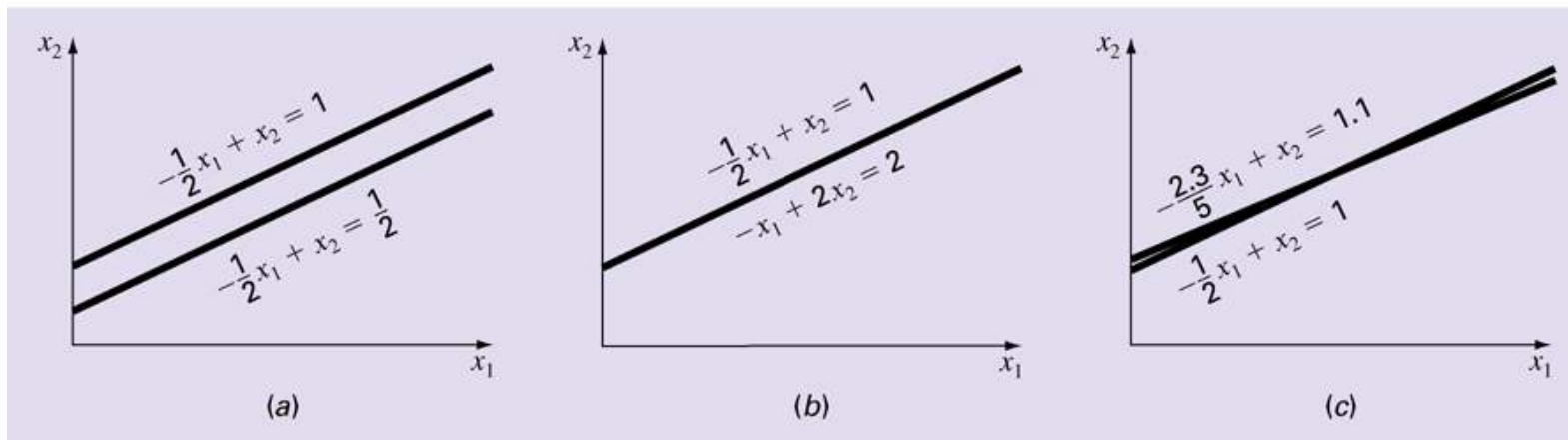


$$\begin{aligned} x + 3y - 2z &= 5 \\ 3x + 5y + 6z &= 7 \end{aligned} \quad x = -7z - 1 \quad \text{and} \quad y = 3z + 2$$

solution for two equations in three variables : line
solution for three equations in three variables : point

Condition of a system

- (a) singular system – no solution
- (b) singular system – infinite solutions
- (c) ill-conditioned system : systems that are extremely sensitive to round-off errors



Direct Solvers

- Gaussian Elimination
- LU factorization, Cholesky Decomposition
- $Ax=b \rightarrow x=A^{-1}b$
 - Analytical, faster
 - Error accumulation.
 - Suitable for small number of unknowns.
 - $O(n^3)$

Iterative Solvers

- Newton-Raphson
 - Gauss-Seidel methods
 - Relaxation
 - Successive Overrelaxation (SOR)
-
- Iterative, slower
 - No error accumulation.
 - Suitable for larger number of unknowns.
 - $O(n^3)$

n 의 크기	CRAY-1 $3n^3$ 나노초	TRS-80 $19,500,000n$ 나노초
10	3×10^{-6} 초	2×10^{-1} 초
100	3×10^{-1} 초	2초
1,000	3초	20초
2,500	50초	50초
10,000	49분	3.2분
1,000,000	95년	5.4시간

The Cray-1 was succeeded in 1982 by the 800 MFLOPS



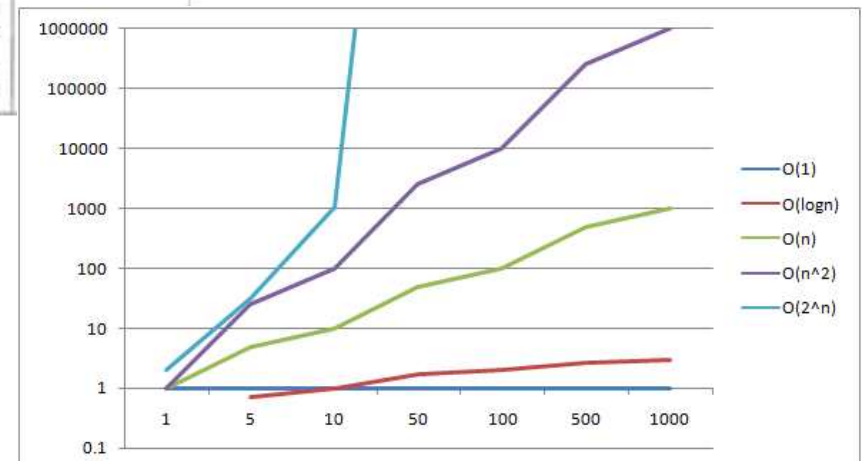
0.89MHz (RAM: 4K)



복잡도함수 (μsec)	$33n$	$46n \log n$	$13n^2$	$3.4n^3$	2^n
10	0.00033초	0.0015초	0.0013초	0.0034초	0.001초
입력 1000	0.003초	0.03초	0.13초	3.4초	4×10^{16} 년
n 의 크기 1,000	0.033초	0.45초	13초	94시간	
10,000	0.33초	6.1초	22분	39일	
100,000	3.3초	1.3분	1.5일	108년	

1000000

100000



Matrix Notation : $A_{m \times n}$

- Row vector (m=1), Column vector (n=1, default)
- Square matrix : principal (or main) diagonal
 - Symmetric matrix
 - Diagonal matrix, Identity matrix
 - Upper (lower) triangular matrix
 - Band matrix : tridiagonal matrix

Symmetric $[A] = \begin{bmatrix} 5 & 1 & 2 \\ 1 & 3 & 7 \\ 2 & 7 & 8 \end{bmatrix}$	Diagonal $[A] = \begin{bmatrix} a_{11} & & \\ & a_{22} & \\ & & a_{33} \end{bmatrix}$	Identity $[A] = \begin{bmatrix} 1 & & \\ & 1 & \\ & & 1 \end{bmatrix}$
Upper Triangular $[A] = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ & a_{22} & a_{23} \\ & & a_{33} \end{bmatrix}$	Lower Triangular $[A] = \begin{bmatrix} a_{11} & & \\ a_{21} & a_{22} & \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$	Banded $[A] = \begin{bmatrix} a_{11} & a_{12} & & \\ a_{21} & a_{22} & a_{23} & \\ & a_{32} & a_{33} & a_{34} \\ & & a_{43} & a_{44} \end{bmatrix}$

Matrix Operations

- $A_{m \times n} + B_{m \times n} = C_{m \times n} \quad c_{ij} = a_{ij} + b_{ij}$

- $A_{m \times k} B_{k \times n} = C_{m \times n} \quad c_{ij} = \sum_{k=1}^n a_{ik} b_{kj}$

$$\begin{bmatrix} 2 & -3 \\ 1 & 0 \\ -1 & 3 \end{bmatrix} + \begin{bmatrix} 9 & -5 \\ 0 & 13 \\ -1 & 3 \end{bmatrix} = \begin{bmatrix} 11 & -8 \\ 1 & 13 \\ -2 & 6 \end{bmatrix}$$

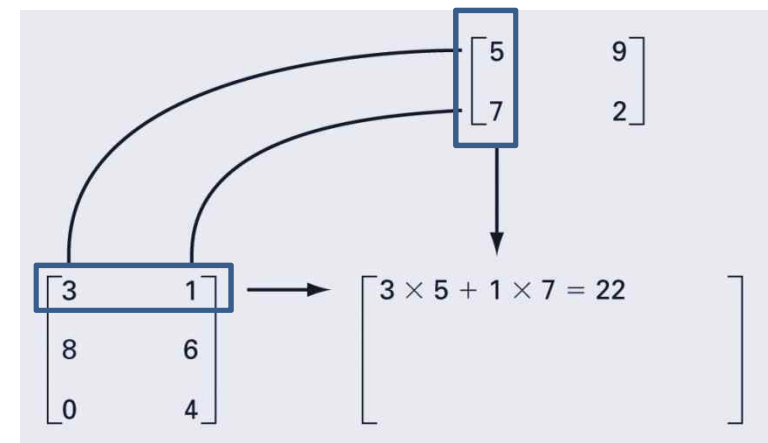
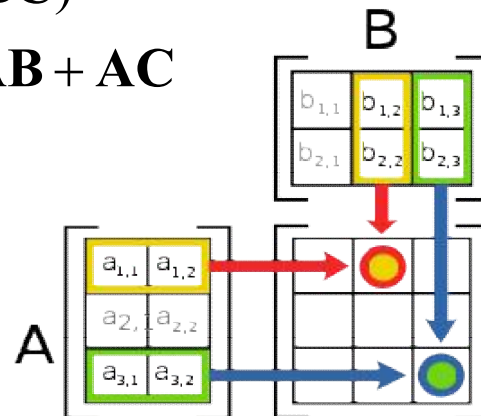
$$4 \begin{bmatrix} 1 & 3 & 5 \\ -1 & -8 & 10 \\ -7 & -5 & 13 \end{bmatrix} = \begin{bmatrix} 4 & 12 & 20 \\ -4 & -32 & 40 \\ -28 & -20 & 52 \end{bmatrix}$$

- $\mathbf{AB} \neq \mathbf{BA}$

$$(\mathbf{AB})\mathbf{C} = \mathbf{A}(\mathbf{BC})$$

$$\mathbf{A}(\mathbf{B} + \mathbf{C}) = \mathbf{AB} + \mathbf{AC}$$

$$\mathbf{AA}^{-1} = \mathbf{I}$$



Matrix Application

- Suppose you make the following sales:

- Monday

- 3 T-shirts at \$10 each
- 4 hats at \$15 each
- 1 pair of shorts at \$20

- Tuesday

- 4 T-shirts at \$10 each
- 2 hats at \$15 each
- 3 pairs of shorts at \$20.



- Then your total revenue for the two days is

$$\begin{array}{ccc} [10 & 15 & 20] & \times & \begin{bmatrix} 3 & 4 \\ 4 & 2 \\ 1 & 3 \end{bmatrix} & = & [110 & 130] \\ \text{Price} & & \text{Quantity} & & \text{Revenue} \end{array}$$

대부분의 의사결정 문제는 연립방정식 문제 → matrix 문제

Representing System of Linear Algebraic Equations

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

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

$$[A]=\begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nm} \end{bmatrix}, [x]=\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}, [b]=\begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

augmented matrix

$$\begin{array}{l} x + 3y - 2z = 5 \\ 3x + 5y + 6z = 7 \\ 2x + 4y + 3z = 8 \end{array} \quad \begin{bmatrix} 1 & 3 & -2 \\ 3 & 5 & 6 \\ 2 & 4 & 3 \end{bmatrix} \quad \begin{bmatrix} 5 \\ 7 \\ 8 \end{bmatrix} \quad \left[\begin{array}{ccc|c} 1 & 3 & -2 & 5 \\ 3 & 5 & 6 & 7 \\ 2 & 4 & 3 & 8 \end{array} \right]$$

MATLAB Matrix Operations

- Creation

```
>> A=[1 2 3; 4 5 6; 7 8 9]
```

```
A =
```

1	2	3
4	5	6
7	8	9

- Matrix augmentation, column-wise

```
>> [A A]
```

```
ans =
```

1	2	3	1	2	3
4	5	6	4	5	6
7	8	9	7	8	9

- Matrix augmentation, row-wise

```
>> [A;A]
```

```
ans =
```

1	2	3
4	5	6
7	8	9
1	2	3
4	5	6
7	8	9

MATLAB Matrix Operations (cont.)

- Creation

```
>> A=[1 2 3; 4 5 6; 7 8 9]
```

```
A =
```

1	2	3
4	5	6
7	8	9

- Matrix Multiplication

```
>> A*A
```

```
ans =
```

30	36	42
66	81	96
102	126	150

- Transpose

```
>> A'
```

```
ans =
```

1	4	7
2	5	8
3	6	9

- Element-by-element Multiplication

```
>> A.*A
```

```
ans =
```

1	4	9
16	25	36
49	64	81

Solving With MATLAB

- MATLAB provides two direct ways to solve systems of linear algebraic equations $[A]\{x\}=\{b\}$:
 - Left-division : $x = A \backslash b$
 - Matrix inversion : $x = \text{inv}(A) * b$
- The matrix inverse is less efficient than left-division and also only works for square, non-singular systems.

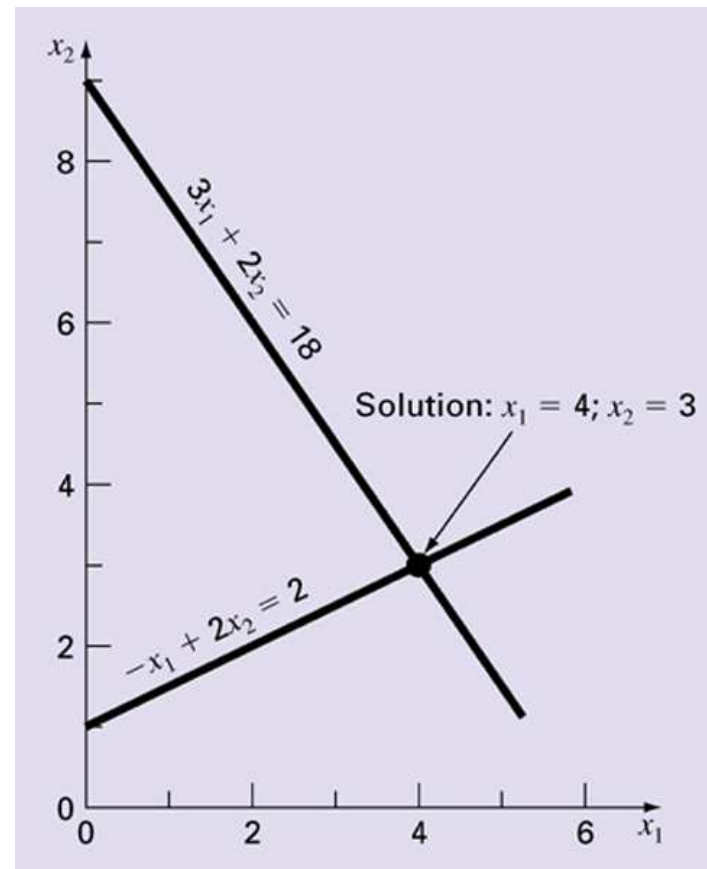
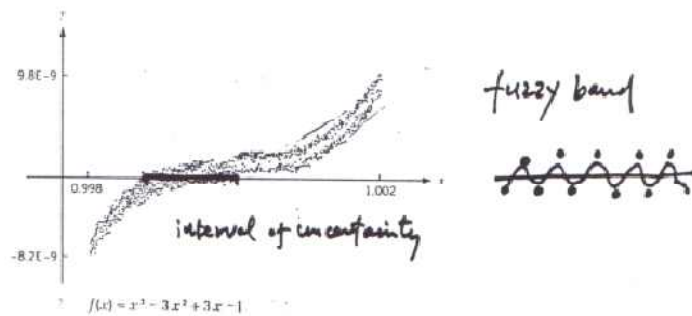
Solving Small Number of Equations

- Graphical methods
- Eliminations of unknowns
- Cramer's rule

Graphical method

$$3x_1 + 2x_2 = 18$$

$$-x_1 + 2x_2 = 2$$



Eliminations of unknowns

$$a_{11}x_1 + a_{12}x_2 = b_1$$

$$a_{21}x_1 + a_{22}x_2 = b_2$$

$$a_{21}a_{11}x_1 + a_{21}a_{12}x_2 = a_{21}b_1$$

$$-) \quad a_{11}a_{21}x_1 + a_{11}a_{22}x_2 = a_{11}b_2$$

$$x_2 = \frac{a_{11}b_2 - a_{21}b_1}{a_{11}a_{22} - a_{21}a_{12}}$$

$$x_1 = \frac{a_{22}b_1 - a_{12}b_2}{a_{11}a_{22} - a_{21}a_{12}}$$



Gauss Elimination

14-12-2008. Solving sets of equations by Gaussian elimination. Further Problems 12, Page 514.
 Problem 12. A network engineer, Peterstone, fourth edition by S.H. Steward with assistance by
 solve the following set of equations by Gaussian elimination. (81) (9)

108. (c)
$$\begin{aligned} x_1 + 2x_2 + 5x_3 + x_4 &= 4 \\ 3x_1 - 4x_2 + 3x_3 - 2x_4 &= 7 \\ 4x_1 + 3x_2 + 2x_3 - x_4 &= 1 \\ x_1 - 2x_2 - 4x_3 - x_4 &= 2. \end{aligned}$$

Explain an error to convert the left-hand side to an upper-triangular matrix. There is no set way of doing this, the coefficient to make yourself begins heavy work. Here's one way, you may well have taken quite a different route.

Expanding from bottom row we have

$$\begin{aligned} 15/2 x_4 &= 75/2 \rightarrow x_4 = 5 \\ -36x_3 - 15/2(5) &= -75/2 \\ \rightarrow -36x_3 &= -75/2 + 75/2 \\ \therefore x_3 &= \frac{0}{-36} \rightarrow x_3 = 0 \\ -25x_2 - 30(0) - 25/2(5) &= -25/2 \\ \rightarrow -25x_2 - 0 &= -125/2 = -25/2 \\ &= -25x_2 = 50 \\ \therefore x_2 &= \frac{50}{-25} \rightarrow x_2 = -2 \end{aligned}$$

Returning the right-hand column to its original position

$$\begin{aligned} 12x_1 + 24(-2) + 60(0) + 12(5) &= 48 \\ \rightarrow 12x_1 - 48 + 0 + 60 &= 48 \\ \rightarrow 12x_1 &= 48 - 60 \therefore x_1 = \frac{36}{12} \\ &\rightarrow x_1 = 3 \end{aligned}$$

So $x_1 = 3$, $x_2 = -2$, $x_3 = 0$, $x_4 = 5$.

M.D. 304/734.

Gauss Elimination

- Based on the principle that elementary row operations retain the solution.
 - swaps two rows
 - scales a row (multiply each element in a row by a non-zero number)
 - subtracts a scaled version of one row from another
- Two steps
 - Forward elimination
 - Back substitution

$$R_i \leftrightarrow R_j$$

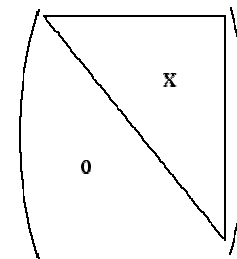
$$sR_i \rightarrow R_i$$

$$sR_i + R_j \rightarrow R_j$$

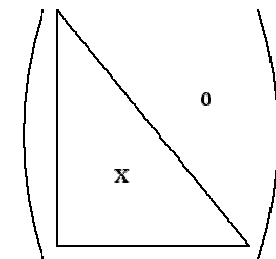
$$\begin{array}{c}
 \left[\begin{array}{ccc|c} a_{11} & a_{12} & a_{13} & b_1 \\ a_{21} & a_{22} & a_{23} & b_2 \\ a_{31} & a_{32} & a_{33} & b_3 \end{array} \right] \\
 \downarrow \\
 \left[\begin{array}{ccc|c} a_{11} & a_{12} & a_{13} & b_1 \\ & a'_{22} & a'_{23} & b'_2 \\ & & a''_{33} & b''_3 \end{array} \right] \\
 \downarrow \\
 \begin{array}{l} x_3 = b''_3 / a''_{33} \\ x_2 = (b'_2 - a'_{23}x_3) / a'_{22} \\ x_1 = (b_1 - a_{13}x_3 - a_{12}x_2) / a_{11} \end{array}
 \end{array}
 \left. \begin{array}{l} \\ \\ \end{array} \right\} \begin{array}{l} \text{(a) Forward} \\ \text{elimination} \\ \\ \text{(b) Back} \\ \text{substitution} \end{array}$$

Forward Elimination

- Goal: to trianglize the matrix by elementary row operations
- Procedure
 - Remove column 1 from row i by
 - subtracting a_{i1}/a_{11} *(row 1) from row i for $2 \leq i \leq n$.
 - (or by adding $-a_{i1}/a_{11}$ *(row 1))
 - Repeat for column 2 for i for $3 \leq i \leq n$.
 - ...
 - Repeat for column $n-1$ for i for $i=n$.



upper triangular



lower triangular

M
U
L
T
I
P
L
I
E
R
S

$$\begin{array}{rrrrrr} 4x_0 & +6x_1 & +2x_2 & -2x_3 & = & 8 \\ 2x_0 & & +5x_2 & -2x_3 & = & 4 \\ -4x_0 & -3x_1 & -5x_2 & +4x_3 & = & 1 \\ 8x_0 & +18x_1 & -2x_2 & +3x_3 & = & 40 \end{array}$$

Multipliers: $-(2/4)$ (blue), $-(-4/4)$ (red), $-(8/4)$ (green)

$$4x_0 + 6x_1 + 2x_2 - 2x_3 = 8$$

M
U
L
T
I
P
L
I
E
R
S

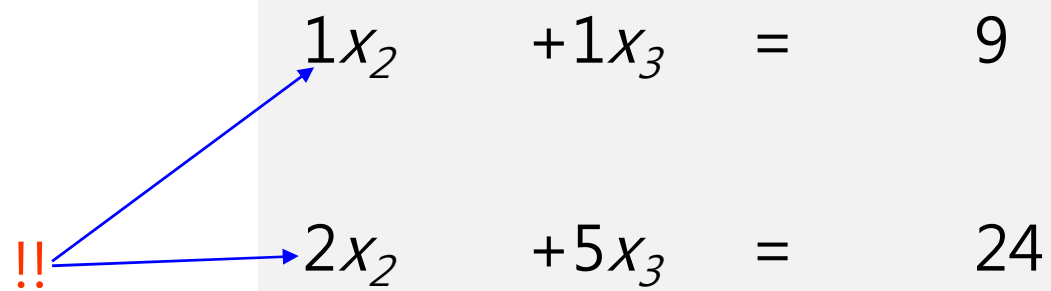
$$-(3/-3)$$

$$-(6/-3)$$

$$\begin{array}{rclcrcl} -3x_1 & +4x_2 & -1x_3 & = & 0 \\ +3x_1 & -3x_2 & +2x_3 & = & 9 \\ +6x_1 & -6x_2 & +7x_3 & = & 24 \end{array}$$

$$4x_0 \quad +6x_1 \quad +2x_2 \quad -2x_3 \quad = \quad 8$$

$$\quad -3x_1 \quad +4x_2 \quad -1x_3 \quad = \quad 0$$



$1x_2 \quad +1x_3 \quad = \quad 9$
 $2x_2 \quad +5x_3 \quad = \quad 24$

$$4x_0 + 6x_1 + 2x_2 - 2x_3 = 8$$

$$-3x_1 + 4x_2 - 1x_3 = 0$$

$$1x_2 + 1x_3 = 9$$

$$3x_3 = 6$$

Forward Elimination (cont.)

- Pseudo code

```
for k = 1:(n-1) % column index
    for i = (k+1):n % row index
         $A(\text{row } i) = A(\text{row } i) - a_{ik}/a_{kk} * A(\text{row } k);$ 
    end
end
```

- MATLAB Code

```
[m,n] = size(A);
if m~=n, error('Matrix A must be square'); end
nb = n+1;
Aug = [A b];
% forward elimination
for k = 1:n-1
    for i = k+1:n
        factor = Aug(i,k)/Aug(k,k);
        Aug(i,k:nb) = Aug(i,k:nb)-factor*Aug(k,k:nb);
    end
end
```


Back Substitution

$$1x_0 + 1x_1 - 1x_2 + 4x_3 = 8$$

$$- 2x_1 - 3x_2 + 1x_3 = 5$$

$$2x_2 - 3x_3 = 0$$

$$2x_3 = 4$$

$$x_3 = 2$$

Back Substitution

$$1x_0 + 1x_1 - 1x_2 = 0$$

$$- 2x_1 - 3x_2 = 3$$

$$2x_2 = 6$$

$$2x_3 = 4$$

Back Substitution

$$1x_0 + 1x_1 - 1x_2 = 0$$

$$-2x_1 - 3x_2 = 3$$

$$2x_2 = 6 \quad x_2 = 3$$

$$2x_3 = 4$$

Back Substitution

$$1x_0 + 1x_1 = 3$$

$$- 2x_1 = 12$$

$$2x_2 = 6$$

$$2x_3 = 4$$

Back Substitution

$$1x_0 + 1x_1 = 3$$

$$-2x_1 = 12 \quad x_1 = -6$$

$$2x_2 = 6$$

$$2x_3 = 4$$

Back Substitution

$$1x_0 = 9$$

$$-2x_1 = 12$$

$$2x_2 = 6$$

$$2x_3 = 4$$

Back Substitution

$$1x_0 = 9 \quad x_0 = 9$$

$$-2x_1 = 12$$

$$2x_2 = 6$$

$$2x_3 = 4$$

Back Substitution

- Goal: to find $\{x_1, x_2, x_2, \dots, x_n\}$
- Procedure:
 - From row $i=n$ to 1

$$x_i = \frac{b'_i - \sum_{j=i+1}^n a'_{ij} x_j}{a'_{ii}}$$

```
% back substitution
x = zeros(n,1);
x(n) = Aug(n,nb)/Aug(n,n);
for i = n-1:-1:1
    x(i) = (Aug(i,nb)-Aug(i,i+1:n)*x(i+1:n))/Aug(i,i);
end
```


Operation Counting

- Forward Elimination: $O(n^3)$
- Back substitution: $O(n^2)$

$$\begin{array}{rcl}
 \text{Forward Elimination} & \frac{2n^3}{3} + O(n^2) & \\
 \text{Back Substitution} & n^2 + O(n) & \\
 \hline
 \text{Total} & \frac{2n^3}{3} + O(n^2) &
 \end{array}$$

$$\begin{array}{lcl}
 \text{add/sub:} & \sum_{i=1}^{i=n-1} i(i+1) = \sum_{i=1}^{i=n-1} (i^2 + i) = \frac{(n-1)n(2n-1)}{6} + \frac{n(n-1)}{2} \approx \frac{n^3}{6} \\
 \text{mul/div:} & \sum_{i=1}^{i=n-1} i(i+2) = \sum_{i=1}^{i=n-1} (i^2 + 2i) = \frac{(n-1)n(2n-1)}{6} + n(n-1) \approx \frac{n^3}{6}
 \end{array}$$

Outer Loop k	Inner Loop i	Addition/Subtraction flops	Multiplication/Division flops
1	2, n	$(n-1) n $	$(n-1) n+1 $
2	3, n	$(n-2) n-1 $	$(n-2) n $
\vdots	\vdots		
k	$k+1, n$	$(n-k) n+1-k $	$(n-k) n+2-k $
\vdots	\vdots		
$n-1$	n, n	$(1) 2 $	$(1) 3 $

Forward Elimination

Pitfalls of Elimination Methods

- Round-off errors.

Forward Elimination: $O(n^3)$
Back substitution: $O(n^2)$

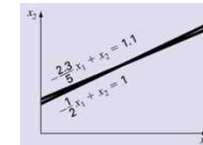
- Division by zero.

- It is possible that during both elimination and back-substitution phases a division by zero can occur.

$$x_i = \frac{b'_i - \sum_{j=i+1}^n a'_{ij} x_j}{a'_{ii}}$$

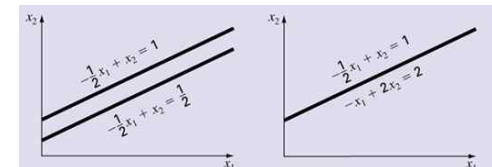
- Ill-conditioned systems.

- Systems where small changes in coefficients result in large changes in the solution.



- Singular systems.

- When two equations are identical, we would lose one degree of freedom and be dealing with the impossible case of $n-1$ equations for n unknowns.
- For large sets of equations, it may not be obvious.
 - The fact that the determinant of a singular system is zero can be used.



Techniques for Improving Solutions

- Use of more significant figures (**pivoting**)
- Pivoting.
 - Errors occur in the divisions of forward elimination due to limited number of significant digits.
 - The larger the divider the smaller the error.
 - If a pivot element is zero(or close to zero), normalization step leads to division by zero. Problem can be avoided by
 - Complete (full) pivoting : Searching for the largest element in all rows and columns then switching.
 - Partial pivoting. : Switching the rows so that the largest element is the pivot element.

Example

$$0.0003 x_1 + 3.0000 x_2 = 2.0001$$

$$1.0000 x_1 + 1.0000 x_2 = 1.0000$$

Significant Figures	x_2	x_1	Absolute Value of Percent Relative Error for x_1
3	0.667	-3.33	1099
4	0.6667	0.0000	100
5	0.66667	0.30000	10
6	0.666667	0.330000	1
7	0.6666667	0.3330000	0.1

After pivoting

$$1.0000 x_1 + 1.0000 x_2 = 1.0000$$

$$0.0003 x_1 + 3.0000 x_2 = 2.0001$$

Significant Figures	x_2	x_1	Absolute Value of Percent Relative Error for x_1
3	0.667	0.333	0.1
4	0.6667	0.3333	0.01
5	0.66667	0.33333	0.001
6	0.666667	0.333333	0.0001
7	0.6666667	0.3333333	0.0000

MATLAB M-File: GaussPivot

```
function x = GaussPivot(A,b)
% GaussPivot(A,b) :
% Solve Ax =b using Gaussian elimination with pivoting
% Input:
%   A = coefficient matrix
%   b = right-hand-side matrix
%
% Output:
%   x = solution matrix

% compute the matrix sizes
[m, n] = size(A);
if m ~= n, error('Matrix A must be square'); end
nb = n + 1;
Aug = [A b];

% forward elimination
for k = 1 : n-1
    % partial pivoting
    [big, i] = max(abs(Aug(k:n,k)));
    ipr = i+k-1;
    if ipr ~= k
        %pivot the rows
        Aug([k,ipr], :) = Aug([ipr,k], :);
    end
    for i = k+1 : n
        factor = Aug(i,k) / Aug(k,k);
        Aug(i,k:nb) = Aug(i,k:nb) - factor*Aug(k,k:nb);
    end;
end

% back-substitution
x = zeros(n,1);
x(n) = Aug(n,nb) / Aug(n,n);
for i = n-1 : -1 : 1
    x(i) = (Aug(i,nb) - Aug(i,i+1:n)*x(i+1:n)) / Aug(i,i);
end
```

Partial Pivoting
(switch rows)

largest element in {x}

`[big,i] = max(x)`

index of the largest element
(from diagonal, not from 1)

Partial
Pivoting

MATLAB Matrix Indexing

```
v = [16 5 9 4 2 11 7 14];
```

```
v(3)      % Extract the third element  
ans = 9
```

```
v([1 5 6]) % Extract the first, fifth, and sixth elements  
ans = 16    2    11
```

```
v(1:2:end) % Extract all the odd elements  
ans = 16    9    2    7
```

```
v(end:-1:1) % Reverse the order of elements  
ans = 14    7    11    2    4    9    5    16
```

```
v([2 3 4]) = [10 15 20] % Replace some elements of v  
v = 16    10    15    20    2    11    7    14
```

```
v([2 3]) = 30 % Replace second and third elements by 30  
v = 16    30    30    20    2    11    7    14
```

MATLAB Matrix Indexing

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

A

From the diagram you can see that $A(14)$ is the same as $A(2, 4)$.

The single subscript can be a vector containing more than one linear index, as in:

```
A([6 12 15])
```

```
ans =
```

```
11 15 12
```

A([2 3 4], [1 2 4])

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

A



5	11	8
9	7	12
4	14	1


```
>> format short
>> x=GaussPivot0(A,b)
```

```
Aug =
```

1	0	2	3	1
-1	2	2	-3	-1
0	1	1	4	2
6	2	2	4	1

Aug = [A b]

```
big =
```

```
6
```

```
i =
```

```
4
```

Find the first pivot element and its index

```
ipr =
```

```
4
```

```
Aug =
```

6	2	2	4	1
-1	2	2	-3	-1
0	1	1	4	2
1	0	2	3	1

Interchange rows 1 and 4

```
factor =
```

```
-0.1667
```

```
Aug =
```

6.0000	2.0000	2.0000	4.0000	1.0000
0	2.3333	2.3333	-2.3333	-0.8333
0	1.0000	1.0000	4.0000	2.0000
1.0000	0	2.0000	3.0000	1.0000

```
factor =
```

```
0
```

```
Aug =
```

6.0000	2.0000	2.0000	4.0000	1.0000
0	2.3333	2.3333	-2.3333	-0.8333
0	1.0000	1.0000	4.0000	2.0000
1.0000	0	2.0000	3.0000	1.0000

```
factor =
```

```
0.1667
```

Eliminate first column

```
Aug =
```

6.0000	2.0000	2.0000	4.0000	1.0000
0	2.3333	2.3333	-2.3333	-0.8333
0	1.0000	1.0000	4.0000	2.0000
0	-0.3333	1.6667	2.3333	0.8333

No need to interchange

```

big =
    2.3333
i =
    1
ipr =
    2
factor =
    0.4286
Aug =
    6.0000    2.0000    2.0000    4.0000    1.0000
           0    2.3333    2.3333   -2.3333   -0.8333
           0         0         0    5.0000    2.3571
           0   -0.3333    1.6667    2.3333    0.8333

```

Second pivot element and index

No need to interchange

```

factor =
   -0.1429
Aug =
    6.0000    2.0000    2.0000    4.0000    1.0000
           0    2.3333    2.3333   -2.3333   -0.8333
           0         0         0    5.0000    2.3571
           0         0         0    2.0000    0.7143

```

Eliminate second column

```

big =
    2
i =
    2
ipr =
    4
Aug =
    6.0000    2.0000    2.0000    4.0000    1.0000
           0    2.3333    2.3333   -2.3333   -0.8333
           0         0    2.0000    2.0000    0.7143
           0         0         0    5.0000    2.3571

```

Third pivot element and index

Interchange rows 3 and 4

```

factor =
    0
Aug =
    6.0000    2.0000    2.0000    4.0000    1.0000
           0    2.3333    2.3333   -2.3333   -0.8333
           0         0    2.0000    2.0000    0.7143
           0         0         0    5.0000    2.3571

```

Eliminate third column

Back
substitution

```

x =
    0
    0
    0
    0.4714
x =
    0
    0
   -0.1143
    0.4714
x =
    0
    0.2286
   -0.1143
    0.4714
x =
   -0.1857
    0.2286
   -0.1143
    0.4714

```

Tridiagonal Systems

- Sparse, only a few elements each row.
- Tridiagonal systems can be solved using the same method as Gauss elimination,
 - but with much less effort because most of the matrix elements are already 0.
 - $O(n)$

$$\begin{bmatrix}
 f_1 & g_1 & & & & & & \\
 e_2 & f_2 & g_2 & & & & & \\
 & e_3 & f_3 & g_3 & & & & \\
 & & \cdot & \cdot & \cdot & & & \\
 & & & \cdot & \cdot & \cdot & & \\
 & & & & \cdot & \cdot & \cdot & \\
 & & & & & e_{n-1} & f_{n-1} & g_{n-1} \\
 & & & & & & e_n & f_n
 \end{bmatrix}
 \begin{bmatrix}
 x_1 \\
 x_2 \\
 x_3 \\
 \cdot \\
 \cdot \\
 \cdot \\
 x_{n-1} \\
 x_n
 \end{bmatrix}
 =
 \begin{bmatrix}
 r_1 \\
 r_2 \\
 r_3 \\
 \cdot \\
 \cdot \\
 \cdot \\
 r_{n-1} \\
 r_n
 \end{bmatrix}$$

Gauss-Jordan

- It is a variation of Gauss elimination. The major differences are
 - All rows are **normalized** by dividing them by their pivot elements.
 - Elimination step results in an **identity matrix**.
 - Consequently, it is not necessary to employ back substitution to obtain solution.

The diagram illustrates the Gauss-Jordan elimination process for a system of three linear equations with three variables:

$$\begin{cases} x_1 + 2x_2 - x_3 = 0 \\ 2x_1 + 3x_2 + 4x_3 = 1 \\ -x_1 + x_2 = 2 \end{cases}$$

The system is represented as (A, b) , where A is the coefficient matrix and b is the constant vector:

$$(A, b) = \begin{bmatrix} 1 & 2 & -1 & 0 \\ 2 & 3 & 4 & 1 \\ -1 & 1 & 0 & 2 \end{bmatrix}$$

The matrix is then transformed into row echelon form, resulting in an identity matrix for the coefficients:

$$\begin{bmatrix} 1 & 0 & 0 & -21/17 \\ 0 & 1 & 0 & 13/17 \\ 0 & 0 & 1 & 5/17 \end{bmatrix}$$

The solutions are:

$$\begin{cases} x_1 = -\frac{21}{17} \\ x_2 = \frac{13}{17} \\ x_3 = \frac{5}{17} \end{cases}$$

	B	C	D	E	F	G	H
Gauss-Jordan elimination for 3 by 3 matrices							
Starting Matrix							
A=		1	2	3			1
		4	5	6	b=		7
		7	8	-9			2

Step 1							
		1	2	3			1
		0	-0.75	-1.5	b=		0.75
		0	-0.85714	-4.28571			-0.71429

$(1/4)R_2 - R_1$

Step 2							
		-0.375	0	0.375			-1.125
		0	1	2	b=		-1
		0	0	-2.25			-1.375

Step 3							
		2.25	0	0			8.125
		0	-1.125	0	b=		2.5
		0	0	-2.25			-1.375

x3=	3.611111
x2=	-2.22222
x1=	0.611111

Verification					
	-1.72222	0.777778	-0.05556		3.611111
A ⁻¹ =	1.444444	-0.55556	0.111111	x=	-2.22222
	-0.05556	0.111111	-0.05556		0.611111

Gauss-Jordan elimination for 3 by 3 matrices					
Starting Matrix					
A=	1	2	3	b=	1
	4	5	6		7
	7	8	-9		2

$R_2 - (4/1)R_1$

Step 1	(Normalize Pivot)				
	1	2	3		1
	4	5	6	b=	7
	7	8	-9		2

Step 2	(eliminate)				
	1	2	3		1
	0	-3	-6	b=	3
	0	-6	-30		-5

Step 3	(Normalize pivot)				
	1	2	3		1
	0	1	2	b=	-1
	0	-6	-30		-5

Step 4	(Eliminate)				
	1	0	-1		3
	0	1	2	b=	-1
	0	0	-18		-11

Step 5	(Normalize pivot)				
	1	0	-1		3
	0	1	2	b=	-1
	0	0	1		0.611111

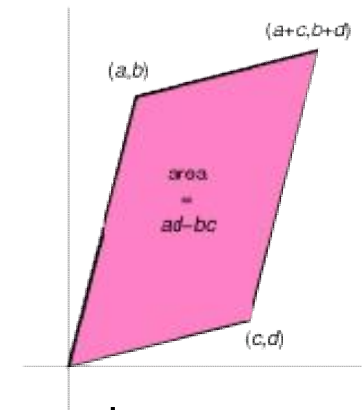
Step 5	(Normalize pivot)				
	1	0	0		3.611111
	0	1	0	b=	-2.22222
	0	0	1		0.611111

x1=	3.611111
x2=	-2.22222
x3=	0.611111

Verification					
	-1.72222	0.777778	-0.05556		3.611111
A ⁻¹ =	1.444444	-0.55556	0.111111	x=	-2.22222
	-0.05556	0.111111	-0.05556		0.611111

Determinant

- Unique number associated with each square matrix
- Required for many multivariate procedures
- Determinant is a scalar
- Determinant denoted as $\det X$ or $|X|$
- Geometrically, the determinant of a 2×2 matrix corresponds to the area spanned by the two (e.g., column) vectors of the matrix.
 - When the vectors are aligned (i.e., linearly dependent), the determinant is 0;
 - when the vectors have some angle between each other, the determinant grows as the area spanned between the vectors grows.
 - At 90° , the determinant is maximal (because the area of the rectangle spanned by the two vectors is maximal).



Determinant

- Rank of matrix
 - Number of linearly independent rows (or columns)
 - Matrix is full rank if all rows (columns) linearly independent
 - Rank reduced if rows linearly dependent
 - Quick check of rank of matrix,
 - calculate determinant,
 - if 0, then matrix is not of full rank

Input matrix:

```
1.000 2.000 3.000 2.000
2.000 3.000 4.000 2.000
3.000 4.000 5.000 2.000
1.000 1.000 1.000 1.000
```

Trace:

10.000

Determinant:

0.000

Rank:

3

Theorem

- if A has a row of zeros or a column of zeros, then $\det(A) = 0$.
- $\det(A) = \det(A^T)$
- If A is an $n \times n$ triangular(upper, lower) or diagonal, then $\det(A)$ is the product of the entries on the main diagonal of the matrix ; that is, $\det(A) = a_{11}a_{22}\dots a_{nn}$.

$$\begin{vmatrix} 2 & 7 & -3 & 8 & 3 \\ 0 & -3 & 7 & 5 & 1 \\ 0 & 0 & 6 & 7 & 6 \\ 0 & 0 & 0 & 9 & 8 \\ 0 & 0 & 0 & 0 & 4 \end{vmatrix} = (2)(-3)(6)(9)(4) = -1296$$

Theorem

- If B is the matrix that results when a single row or single column of A is multiplied by a scalar k , then $\det(B) = k \det(A)$
- If B is the matrix that results when two rows or two columns of A are interchanged, then $\det(B) = -\det(A)$
- If B is the matrix that results when a multiple of one row of A is added to another row or when a multiple column is added to another column, then $\det(B) = \det(A)$. (pivoting)

$$\begin{vmatrix} ka_{11} & ka_{12} & ka_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = k \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

$$\det(B) = k \det(A)$$

$$\begin{vmatrix} a_{21} & a_{22} & a_{23} \\ a_{11} & a_{12} & a_{13} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = - \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

$$\det(B) = -\det(A)$$

$$\begin{vmatrix} a_{11} + ka_{21} & a_{12} + ka_{22} & a_{13} + ka_{23} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} = \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix}$$

$$\det(B) = \det(A)$$

$$\det(kA) = k^n \det(A)$$

Theorem

- If A is a square matrix with two proportional rows or two proportional column, then $\det(A) = 0$.

$$\begin{vmatrix} 1 & 3 & -2 & 4 \\ 2 & 6 & -4 & 8 \\ 3 & 9 & 1 & 5 \\ 1 & 1 & 4 & 8 \end{vmatrix} = \begin{vmatrix} 1 & 3 & -2 & 4 \\ 0 & 0 & 0 & 0 \\ 3 & 9 & 1 & 5 \\ 1 & 1 & 4 & 8 \end{vmatrix} = 0$$

The second row is 2 times the first, so we added -2 times the first row to the second to introduce a row of zeros

$$\begin{vmatrix} -1 & 4 \\ -2 & 8 \end{vmatrix}, \begin{vmatrix} 3 & -1 & 4 & -5 \\ 6 & -2 & 5 & 2 \\ 5 & 8 & 1 & 4 \\ -9 & 3 & -12 & 15 \end{vmatrix}$$

Theorem

- If A and B are square matrices of the same size, then $\det(AB) = \det(A)\det(B)$
- If A and B are square matrices of the same size, then $\det(A+B) \neq \det(A) + \det(B)$

$$A = \begin{bmatrix} 3 & 1 \\ 2 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} -1 & 3 \\ 5 & 8 \end{bmatrix}, \quad AB = \begin{bmatrix} 2 & 17 \\ 3 & 14 \end{bmatrix}$$
$$\det(A) = 1 \quad \det(B) = -23 \quad \det(AB) = -23$$

$$A = \begin{bmatrix} 1 & 2 \\ 2 & 5 \end{bmatrix}, \quad B = \begin{bmatrix} 3 & 1 \\ 1 & 3 \end{bmatrix}, \quad A + B = \begin{bmatrix} 4 & 3 \\ 3 & 8 \end{bmatrix}$$
$$\det(A) = 1 \quad \det(B) = 8 \quad \det(A + B) = 23$$

Theorem

- Let A , B , and C be $n \times n$ matrices that differ only in a single row, say the r th, and assume that the r th row of C can be obtained by adding corresponding entries in the r th rows of A and B . Then

$$\det(C) = \det(A) + \det(B)$$

The same result holds for columns.

A	B	C
$\begin{bmatrix} 1 & 7 & 5 \\ 2 & 0 & 3 \\ 1 & 4 & 7 \end{bmatrix}$	$\begin{bmatrix} 1 & 7 & 5 \\ 2 & 0 & 3 \\ 0 & 1 & -1 \end{bmatrix}$	$\begin{bmatrix} 1 & 7 & 5 \\ 2 & 0 & 3 \\ 1+0 & 4+1 & 7+(-1) \end{bmatrix}$

$$\det \begin{bmatrix} 1 & 7 & 5 \\ 2 & 0 & 3 \\ 1+0 & 4+1 & 7+(-1) \end{bmatrix} = \det \begin{bmatrix} 1 & 7 & 5 \\ 2 & 0 & 3 \\ 1 & 4 & 7 \end{bmatrix} + \det \begin{bmatrix} 1 & 7 & 5 \\ 2 & 0 & 3 \\ 0 & 1 & -1 \end{bmatrix}$$

Theorem

- A square matrix A is invertible if and only if $\det(A) \neq 0$

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 1 & 0 & 1 \\ 2 & 4 & 6 \end{bmatrix}$$

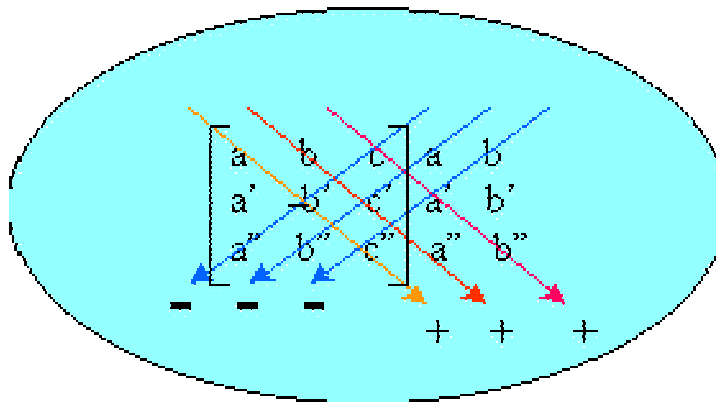
Since the first and third rows of A are proportional, $\det(A) = 0$.
Thus, A is not invertible

- If A is invertible, then

$$\det(A^{-1}) = \frac{1}{\det(A)}$$

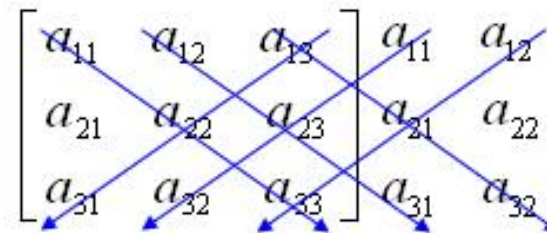
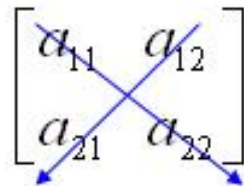
Determinant

- Computation
 - Mnemonic method
 - Permutation method
 - Gauss Jordan method
 - Minor method (method of Cofactors)



Mnemonic Method

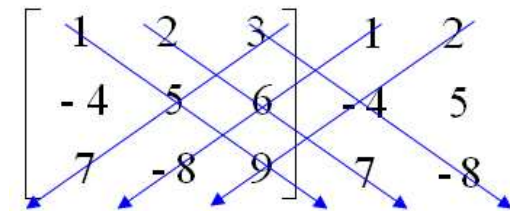
- The determinant is computed by summing the products on the rightward arrows and subtracting the products on the leftward arrows.



$$A = \begin{bmatrix} 3 & 1 \\ 4 & -2 \end{bmatrix} \text{ and } B = \begin{bmatrix} 1 & 2 & 3 \\ -4 & 5 & 6 \\ 7 & -8 & 9 \end{bmatrix}$$

$$\det(A) = (3)(-2) - (1)(4) = -10$$

$$\det(B) = (45) + (84) + (96) - (105) - (-48) - (-72) = 240$$



Permutation method : **Permutation**

- The set of integers $\{1, 2, \dots, n\}$ is an arrangement of these integers in some order without omission and repetition
 - Permutations of three integers
 - (ex) six different permutations of the set of integers $\{1, 2, 3\}$.
 - $(1, 2, 3)$ $(2, 1, 3)$ $(3, 1, 2)$ $(1, 3, 2)$ $(2, 3, 1)$ $(3, 2, 1)$
 - Permutations of four integers
 - $(1, 2, 3, 4)$ $(2, 1, 3, 4)$ $(3, 1, 2, 4)$ $(4, 1, 2, 3)$ $(1, 2, 4, 3)$ $(2, 1, 4, 3)$
 $(3, 1, 4, 2)$ $(4, 1, 3, 2)$ $(1, 3, 2, 4)$ $(2, 3, 1, 4)$ $(3, 2, 1, 4)$ $(4, 2, 1, 3)$
 $(1, 3, 4, 2)$ $(2, 3, 4, 1)$ $(3, 2, 4, 1)$ $(4, 2, 3, 1)$ $(1, 4, 2, 3)$ $(2, 4, 1, 3)$
 $(3, 4, 1, 2)$ $(4, 3, 1, 2)$ $(1, 4, 3, 2)$ $(2, 4, 3, 1)$ $(3, 4, 2, 1)$ $(4, 3, 2, 1)$

Permutation method : Inversion

- In a permutation $(j_1, j_2, j_3, \dots, j_n)$ a larger integer precedes a smaller one. $(\textcolor{red}{2}, \textcolor{blue}{1}, 3)$
- Counting Inversions
 - $(6, 1, 3, 4, 5, 2) : 5 + 0 + 1 + 1 + 1 = 8$.
 - $(2, 4, 1, 3) : 1 + 2 + 0 = 3$.
 - $(1, 2, 3, 4) : \text{no inversions in this permutation}$
- A permutation is called **even** if the total number of inversions is an even integer and is called **odd** if the total inversions is an odd integer.

Permutation	Number of Inversions	classification
(1,2,3)	0	even
(1,3,2)	1	odd
(2,1,3)	1	odd
(2,3,1)	2	even
(3,1,2)	2	even
(3,2,1)	3	odd

Permutation method : Elementary Products

- From an $n \times n$ matrix A , any product of n entries from A , no two of which come from the same row or same column
- An $n \times n$ matrix A has $n!$ elementary products.

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \quad a_1 - a_2 - \quad a_{11} a_{22} \quad a_{12} a_{21}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \quad a_1 - a_2 - a_3 - \quad \begin{matrix} a_{11} a_{22} a_{33} & a_{12} a_{21} a_{33} & a_{13} a_{21} a_{32} \\ a_{11} a_{23} a_{32} & a_{12} a_{23} a_{31} & a_{13} a_{22} a_{31} \end{matrix}$$

Permutation method : Signed Elementary Products

- *Signed elementary product* is an elementary product multiplied by +1 or -1 .
 - (+1) : if elementary product is an even permutation
 - (– 1) : if elementary product is an odd permutation.

Elementary Product	Associated Permutation	Even or Odd	Signed Elementary Product
$a_{11}a_{22}$	(1,2)	even	$a_{11}a_{22}$
$a_{12}a_{21}$	(2,1)	odd	$-a_{12}a_{21}$

Elementary Product	Associated Permutation	Even or Odd	Signed Elementary Product
$a_{11}a_{22}a_{33}$	(1,2,3)	even	$a_{11}a_{22}a_{33}$
$a_{11}a_{23}a_{32}$	(1,3,2)	odd	$-a_{11}a_{23}a_{32}$
$a_{12}a_{21}a_{33}$	(2,1,3)	odd	$-a_{12}a_{21}a_{33}$
$a_{12}a_{23}a_{31}$	(2,3,1)	even	$a_{12}a_{23}a_{31}$
$a_{13}a_{21}a_{32}$	(3,1,2)	even	$a_{13}a_{21}a_{32}$
$a_{13}a_{22}a_{31}$	(3,2,1)	odd	$-a_{13}a_{22}a_{31}$

Permutation method : Determinant

- The *determinant* is denoted by $|A|$ or $\det(A)$ to be the sum of all signed elementary products from A .

$$\det(A) = \sum \pm a_{1j_1} a_{2j_2} \dots a_{nj_n}$$

$$\det \begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} = a_{11}a_{22} - a_{12}a_{21}$$

$$\det \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} = a_{11}a_{22}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} - a_{13}a_{22}a_{31} - a_{12}a_{21}a_{33} - a_{11}a_{23}a_{32}$$

Gauss Elimination Method

- Evaluate $\det(A)$ where

$$A = \begin{bmatrix} 0 & 1 & 5 \\ 3 & -6 & 9 \\ 2 & 6 & 1 \end{bmatrix}$$

$$\det(A) = \begin{vmatrix} 0 & 1 & 5 \\ 3 & -6 & 9 \\ 2 & 6 & 1 \end{vmatrix} = - \begin{vmatrix} 3 & -6 & 9 \\ 0 & 1 & 5 \\ 2 & 6 & 1 \end{vmatrix}$$

$$= -3 \begin{vmatrix} 1 & -2 & 3 \\ 0 & 1 & 5 \\ 2 & 6 & 1 \end{vmatrix}$$

The first and second rows of A are interchanged.

A common factor of 3 from the first row was taken through the determinant sign

Gauss Elimination Method

$$= -3 \begin{vmatrix} 1 & -2 & 3 \\ 0 & 1 & 5 \\ 0 & 10 & -5 \end{vmatrix}$$

← -2 times the first row was added to the third row.

$$= -3 \begin{vmatrix} 1 & -2 & 3 \\ 0 & 1 & 5 \\ 0 & 0 & -55 \end{vmatrix}$$

← -10 times the second row was added to the third row

$$= (-3)(-55) \begin{vmatrix} 1 & -2 & 3 \\ 0 & 1 & 5 \\ 0 & 0 & 1 \end{vmatrix}$$

← A common factor of -55 from the last row was taken through the determinant sign.

$$= (-3)(-55)(1) = 165$$

.

● Gauss-Jordan 소거법

● 단위행렬 변환

(ex) $A = \begin{bmatrix} 4 & 3 \\ 2 & 1 \end{bmatrix}$

$$|A| = \begin{vmatrix} 4 & 3 \\ 2 & 1 \end{vmatrix}$$

$$\frac{1}{4}|A| = \begin{vmatrix} 1 & \frac{3}{4} \\ 2 & 1 \end{vmatrix} = \begin{vmatrix} 1 & \frac{3}{4} \\ 0 & -\frac{1}{2} \end{vmatrix}$$

$$(-2)\left(\frac{1}{4}\right)|A| = \begin{vmatrix} 1 & \frac{3}{4} \\ 0 & 1 \end{vmatrix} = \begin{vmatrix} 1 & 0 \\ 0 & 1 \end{vmatrix} = 1$$

$$\therefore |A| = \left(-\frac{1}{2}\right) \cdot (4) = 2$$

Minor/Cofactor Method : Minors

- The determinant of a 3x3 matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \quad \begin{aligned} \det(A) &= a_{11}a_{22}a_{33} + a_{12}a_{23}a_{31} + a_{13}a_{21}a_{32} \\ &\quad - a_{13}a_{22}a_{31} - a_{12}a_{21}a_{33} - a_{11}a_{23}a_{32} \\ \Rightarrow \det(A) &= a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{23}a_{31}) \\ &\quad + a_{13}(a_{21}a_{32} - a_{22}a_{31}) \end{aligned}$$

$$M_{11} = \begin{vmatrix} a_{22} & a_{23} \\ a_{32} & a_{33} \end{vmatrix}, \quad M_{12} = \begin{vmatrix} a_{21} & a_{23} \\ a_{31} & a_{33} \end{vmatrix}, \quad M_{13} = \begin{vmatrix} a_{21} & a_{22} \\ a_{31} & a_{32} \end{vmatrix}$$

Minor/Cofactor Method : Cofactors

- If A is a square matrix, then the **minor** of a_{ij} is denoted by M_{ij} and
 - is defined to be the determinant of the submatrix that remains after the i th row and j th column are deleted from A .
 - The number $(-1)^{i+j}M_{ij}$ is denoted by C_{ij} and is called the **cofactor** of a_{ij}

Let $A = \begin{bmatrix} 3 & 1 & -4 \\ 2 & 5 & 6 \\ 1 & 4 & 8 \end{bmatrix}$

The minor of entry a_{11} is $M_{11} = \begin{vmatrix} \cancel{3} & \cancel{1} & \cancel{-4} \\ 2 & 5 & 6 \\ 1 & 4 & 8 \end{vmatrix} = \begin{vmatrix} 5 & 6 \\ 4 & 8 \end{vmatrix} = 16$

+	-	+
-	+	-
+	-	+

The cofactor of a_{11} is $C_{11} = (-1)^{1+1}M_{11} = M_{11} = 16$

Minor/Cofactor Method : Cofactor Expansions

- The expression can be written in terms of minors and cofactors as

$$\begin{aligned}\det(A) &= a_{11}M_{11} + a_{12}(-M_{12}) + a_{13}M_{13} \\ &= a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13}\end{aligned}$$

- This method is called *cofactor expansion* along the first row of A .
- By rearranging, we can obtain

$$\begin{aligned}\det(A) &= a_{11}C_{11} + a_{12}C_{12} + a_{13}C_{13} \\ &= a_{11}C_{11} + a_{21}C_{21} + a_{31}C_{31} \\ &= a_{21}C_{21} + a_{22}C_{22} + a_{23}C_{23} \\ &= a_{12}C_{12} + a_{22}C_{22} + a_{32}C_{32} \\ &= a_{31}C_{31} + a_{32}C_{32} + a_{33}C_{33} \\ &= a_{13}C_{13} + a_{23}C_{23} + a_{33}C_{33}\end{aligned}$$

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}$$

Minor/Cofactor Method : Cofactor Expansions

$$\text{Let } A = \begin{bmatrix} 3 & 1 & 0 \\ -2 & -4 & 3 \\ 5 & 4 & -2 \end{bmatrix} \quad \det(A) = \begin{vmatrix} 3 & 1 & 0 \\ -2 & -4 & 3 \\ 5 & 4 & -2 \end{vmatrix} = 3 \begin{vmatrix} -4 & 3 \\ 4 & -2 \end{vmatrix} - 1 \begin{vmatrix} -2 & 3 \\ 5 & -2 \end{vmatrix} + 0 \begin{vmatrix} -2 & -4 \\ 5 & 4 \end{vmatrix}$$

$$= 3(-4) - (1)(-11) + 0 = -1$$

$$\det(A) = \begin{vmatrix} 3 & 1 & 0 \\ -2 & -4 & 3 \\ 5 & 4 & -2 \end{vmatrix} = 3 \begin{vmatrix} -4 & 3 \\ 4 & -2 \end{vmatrix} - (-2) \begin{vmatrix} 1 & 0 \\ 4 & -2 \end{vmatrix} + 5 \begin{vmatrix} 1 & 0 \\ -4 & 3 \end{vmatrix}$$

$$= 3(-4) - (-2)(-2) + 5(3) = -1$$

Sparse Matrices

$$\begin{pmatrix} 0 & 3 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 7 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 7 \\ 0 & 0 & 0 & 5 & 0 & 0 & 0 & 8 & 0 & 0 \\ 2 & 0 & 0 & 0 & 0 & 0 & 6 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 9 & 0 & 0 & 0 & 4 & 0 \\ 0 & 1 & 0 & 7 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 4 \\ 0 & 0 & 7 & 0 & 0 & 0 & 0 & 5 & 0 & 0 \end{pmatrix}$$

Cramer's Rule

- If $Ax=b$ is a system of n linear equations in n unknowns such that $\det(A) \neq 0$, then the system has a unique solution. This solution is

$$x_1 = \frac{\det(A_1)}{\det(A)}, \quad x_2 = \frac{\det(A_2)}{\det(A)}, \quad x_3 = \frac{\det(A_3)}{\det(A)}$$

Where A_j is the matrix obtained by replacing the entries in the column of A by vector b .

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

$$[A] = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \text{ coefficient matrix}$$

$$x_1 = \frac{\begin{vmatrix} b_1 & a_{12} & a_{13} \\ b_2 & a_{22} & a_{23} \\ b_3 & a_{32} & a_{33} \end{vmatrix}}{|A|}, \quad x_2 = \frac{\begin{vmatrix} a_{11} & b_1 & a_{13} \\ a_{21} & b_2 & a_{23} \\ a_{31} & b_3 & a_{33} \end{vmatrix}}{|A|}, \quad x_3 = \frac{\begin{vmatrix} a_{11} & a_{12} & b_1 \\ a_{21} & a_{22} & b_2 \\ a_{31} & a_{32} & b_3 \end{vmatrix}}{|A|}$$

Cramer's Rule

- Use Cramer's rule to solve
$$\begin{aligned}x_1 + \quad + 2x_3 &= 6 \\ -3x_1 + 4x_2 + 6x_3 &= 30 \\ -x_1 - 2x_2 + 3x_3 &= 8\end{aligned}$$

$$A = \begin{bmatrix} 1 & 0 & 2 \\ -3 & 4 & 6 \\ -1 & -2 & 3 \end{bmatrix} \quad A_1 = \begin{bmatrix} 6 & 0 & 2 \\ 30 & 4 & 6 \\ 8 & -2 & 3 \end{bmatrix} \quad A_2 = \begin{bmatrix} 1 & 6 & 2 \\ -3 & 30 & 6 \\ -1 & 8 & 3 \end{bmatrix} \quad A_3 = \begin{bmatrix} 1 & 0 & 6 \\ -3 & 4 & 30 \\ -1 & -2 & 8 \end{bmatrix}$$

$$x_1 = \frac{\det(A_1)}{\det(A)} = \frac{-40}{44} = \frac{-10}{11} \quad x_2 = \frac{\det(A_2)}{\det(A)} = \frac{72}{44} = \frac{18}{11} \quad x_3 = \frac{\det(A_3)}{\det(A)} = \frac{152}{44} = \frac{38}{11}$$

Matlab : Determinant

- $d = \det(X)$
 - Using $\det(X) == 0$ as a test for matrix singularity ($==$, $\sim=$)
 - The function `cond(X)` can check for singular and nearly singular matrices.
 - The determinant is computed from the triangular factors obtained by Gaussian elimination
 - Examples
 - $A = [1 \ 2 \ 3; 4 \ 5 \ 6; 7 \ 8 \ 9]$
 - $d = \det(A)$ produces $d = 0$ (singular matrix)
 - Changing $A(3,3)$ with $A(3,3) = 0$
 - $d = \det(A)$ produces $d = 27$

Inverse Matrix

- if $AB = I$, then $A = B^{-1}$ or $B = A^{-1}$ ($\det(A) \neq 0$)

- Gauss-Jordan 소거법

$$A \cdot X = I$$

$$(A^{-1})A \cdot X = (A^{-1})I$$

$$I \cdot X = A^{-1}$$

$$\Rightarrow [A \rightarrow I; I \rightarrow A^{-1}]$$

$$\begin{bmatrix} 1 & 2 & 1 & 1 & 0 & 0 \\ 2 & 9 & 2 & 0 & 1 & 0 \\ 1 & 7 & 2 & 0 & 0 & 1 \end{bmatrix} \quad \begin{bmatrix} 1 & 0 & 0 & 0.8 & 0.6 & -1 \\ 0 & 1 & 0 & -0.4 & 0.2 & 0 \\ 0 & 0 & 1 & 1 & -1 & 1 \end{bmatrix}$$

the second part is the INVERSE MATRIX of the blue part

$$(2X) \quad A = \begin{bmatrix} 1 & -1 & 2 \\ 3 & 0 & 1 \\ 1 & 0 & 2 \end{bmatrix}$$

$$\begin{bmatrix} \textcircled{1} & -1 & 2 & | & 1 & 0 & 0 \\ 3 & 0 & 1 & | & 0 & 1 & 0 \\ 1 & 0 & 2 & | & 0 & 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1 & 2 & | & 1 & 0 & 0 \\ 0 & 3 & -5 & | & -3 & 1 & 0 \\ 0 & 1 & \textcircled{0} & | & -1 & 0 & 1 \end{bmatrix}$$

$$\begin{array}{l} R_2 \leftrightarrow R_3 \\ ([A|B] \sim A, b, c) \\ \text{for } \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 0 \\ 0 & 3 & -5 \end{bmatrix} \end{array} \rightarrow \begin{bmatrix} 1 & -1 & 2 & | & 1 & 0 & 0 \\ 0 & \textcircled{1} & 0 & | & -1 & 0 & 1 \\ 0 & 3 & -5 & | & -3 & 1 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 & | & 0 & 0 & 1 \\ 0 & 1 & 0 & | & -1 & 0 & 1 \\ 0 & 0 & -5 & | & 0 & 1 & -3 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 0 & 2 & | & 0 & 0 & 1 \\ 0 & 1 & 0 & | & -1 & 0 & 1 \\ 0 & 0 & 1 & | & 0 & -\frac{1}{5} & \frac{3}{5} \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 & | & 0 & \frac{2}{5} & -\frac{1}{5} \\ 0 & 1 & 0 & | & -1 & 0 & 1 \\ 0 & 0 & 1 & | & 0 & -\frac{1}{5} & \frac{3}{5} \end{bmatrix}$$

$$(Ref) \quad \begin{bmatrix} 1 & -1 & 2 & | & 1 & 0 & 0 \\ 0 & \textcircled{1} & -\frac{5}{3} & | & -\frac{1}{3} & 0 & 0 \\ 0 & 1 & 0 & | & -1 & 0 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & \frac{1}{3} & | & 0 & \frac{1}{3} & 0 \\ 0 & 1 & -\frac{5}{3} & | & -\frac{1}{3} & 0 & 0 \\ 0 & 0 & \textcircled{\frac{1}{3}} & | & 0 & -\frac{1}{3} & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & \frac{1}{3} & | & 0 & \frac{1}{3} & 0 \\ 0 & 1 & -\frac{5}{3} & | & -\frac{1}{3} & 0 & 0 \\ 0 & 0 & \textcircled{1} & | & 0 & -\frac{1}{3} & \frac{2}{3} \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 0 & | & 0 & \frac{2}{5} & -\frac{1}{5} \\ 0 & 1 & 0 & | & -1 & 0 & 1 \\ 0 & 0 & 1 & | & 0 & -\frac{1}{5} & \frac{3}{5} \end{bmatrix}$$

Inverse Matrix : Adjoint Matrix

- The number $(-1)^{i+j}M_{ij}$ is denoted by C_{ij} and is called the **cofactor** of entry a_{ij}
- If A is any $n \times n$ matrix and C_{ij} is the cofactor of a_{ij} then the matrix

$$\begin{bmatrix} C_{11} & C_{12} & \cdots & C_{1n} \\ C_{21} & C_{22} & \cdots & C_{2n} \\ \vdots & \vdots & & \vdots \\ C_{n1} & C_{n2} & \cdots & C_{nn} \end{bmatrix}$$

is called the *matrix of cofactors from A* .

- The **transpose** of this matrix is called the **adjoint** of A and is denoted by $\text{adj}(A)$

Inverse Matrix : Adjoint Matrix

The cofactors of A are

$$\text{Let } A = \begin{bmatrix} 3 & 2 & -1 \\ 1 & 6 & 3 \\ 2 & -4 & 0 \end{bmatrix} \quad \begin{array}{lll} C_{11} = 12 & C_{12} = 6 & C_{13} = -16 \\ C_{21} = 4 & C_{22} = 2 & C_{23} = 16 \\ C_{31} = 12 & C_{32} = -10 & C_{33} = 16 \end{array}$$

cofactor

$$\begin{bmatrix} 12 & 6 & -16 \\ 4 & 2 & 16 \\ 12 & -10 & 16 \end{bmatrix} \quad \text{adj}(A) = \begin{bmatrix} 12 & 4 & 12 \\ 6 & 2 & -10 \\ -16 & 16 & 16 \end{bmatrix}$$

Inverse Matrix : Adjoint Matrix

If A is an invertible matrix, then

$$A^{-1} = \frac{1}{\det(A)} \text{adj}(A)$$

$$\begin{aligned} \text{Let } A &= \begin{bmatrix} 3 & 2 & -1 \\ 1 & 6 & 3 \\ 2 & -4 & 0 \end{bmatrix} & \det(A) &= 64 \\ A^{-1} &= \frac{1}{64} \begin{bmatrix} 12 & 4 & 12 \\ 6 & 2 & -10 \\ -16 & 16 & 16 \end{bmatrix} = \begin{bmatrix} \frac{12}{64} & \frac{4}{64} & \frac{12}{64} \\ \frac{6}{64} & \frac{2}{64} & \frac{-10}{64} \\ \frac{-16}{64} & \frac{16}{64} & \frac{16}{64} \end{bmatrix} \end{aligned}$$

Inverse Matrix : Matlab

- $y = \text{inv}(X)$ returns the inverse of the square matrix X .
- A warning message is printed if X is badly scaled or nearly singular.

$$Ax=b$$
$$x=A^{-1}b$$

- Matrix inversion

```
>> x=inv(A)*b
```

- Left division

```
>> x=A\b
```

/ (slash) : matrix right division, A/B is $A*\text{inv}(B)$
 \ (backslash) : matrix left division, $A\backslash B$ is $\text{inv}(A)*B$

$$\begin{aligned} 3w-2y+4z &= 8 \\ 5w+8y-6z &= -5 \\ 9w-2y+7z &= -17 \end{aligned}$$

b=	A=
8	3 -2 4
-5	5 8 -6
-17	9 -2 7

```
>> X=inv(A)*b  
or  
>> X=A\b
```

X=	
-36.7778	
71.2778	
65.2222	

Gaussian elimination

THE END

Homework : 추후 공지

Report : 추후 공지

