MATRIX ALGEBRA FOR NON-HOMOGENEOUS LINEAR ALGEBRAIC SYSTEM

WEEK 6: MATRIX ALGEBRA FOR NON-HOMOGENEOUS LINEAR ALGEBRAIC SYSTEM 6.1 INTRODUCTION

A *matrix* is an array of *mn* elements (where *m* and *n* are integers) arranged in *m* rows and *n* columns. The difference between matrix and column/ row vector is shown in Table 6.1.

Matrix	Column vector	Row vector
	(i.e. matrix with one column)	(i.e. matrix with one row)
$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \cdots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \cdots & a_{2n} \\ a_{31} & a_{32} & a_{33} & \cdots & a_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \cdots & a_{mn} \end{bmatrix}$	$\boldsymbol{C} = \begin{cases} c_{11} \\ c_{21} \\ c_{31} \\ \vdots \\ c_{m1} \end{cases}$	$\boldsymbol{R} = \{r_{11} \ r_{12} \ r_{13} \ \cdots \ r_{1n}\}$
Size $(A) = m \times n$	Size $(\mathcal{C}) = m \times 1$	Size $(\mathbf{R}) = 1 \times n$

Table 6.1 Matrix, column vector and row vector.

where a_{mn} is the element of the matrix at m^{th} row and n^{th} column. If m = n, it is known as square matrix. Non-square matrix has $m \neq n$.

The common notation of matrix and vector is shown in Table 6.2:

Table 6.2 Common notation of matrix and vector.

Matrix	Vector	
upper-case non-italic bold letter (e.g. $\mathbf{A} = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$)	upper-case italic bold letter (e.g. $A = \begin{cases} 1 \\ 3 \end{cases}$)	
symbol in box bracket (e.g. $[A] = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$)	symbol in curly bracket (e.g. $\{A\} = \begin{cases} 1 \\ 3 \end{cases}$)	

Table 6.3 Type of matrices.

Zero/ Null Matrix	Symmetric Matrix	Diagonal Matrix
$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$	$\begin{bmatrix} 5 & 1 & 2 \\ 1 & 3 & 7 \\ 2 & 7 & 8 \end{bmatrix}$ where $a_{ij} = a_{ji}$	$\begin{bmatrix} 5 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 8 \end{bmatrix}$ • All elements off the main diagonal are equal to 0.
Identity/ Unit Matrix $ \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} $ • Diagonal matrix with element = 1	Upper Triangular matrix $ \begin{bmatrix} 2 & -5 & 6 \\ 0 & 3 & 8 \\ 0 & 0 & 1 \end{bmatrix} $ • All the elements below main diagonal = 0	Lower Triangular matrix $ \begin{bmatrix} 2 & 0 & 0 \\ 9 & 3 & 0 \\ 20 & -3 & 1 \end{bmatrix} $ • All the elements above main diagonal = 0
Banded Matrix $\begin{bmatrix} a_{11} & a_{12} & a_{13} & 0 & 0 \\ a_{21} & a_{22} & a_{23} & a_{24} & 0 \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \\ 0 & a_{42} & a_{43} & a_{44} & a_{45} \\ 0 & 0 & a_{53} & a_{54} & a_{55} \end{bmatrix}$ • All elements = 0, except for a band centered on the main diagonal	Tridiagonal Matrix $\begin{bmatrix} 1 & 3 & 0 & 0 \\ 5 & 2 & 9 & 0 \\ 0 & 6 & 5 & 1 \\ 0 & 0 & -3 & -9 \end{bmatrix}$ • Banded matrix that has bandwidth of 3.	Anti-Symmetric / Skew- Symmetric Matrix $\begin{bmatrix} 5 & 1 & -2 \\ -1 & -3 & 7 \\ 2 & -7 & 8 \end{bmatrix}$ where $a_{ij} = -a_{ji}$

The basic operations of matrices such as trace, transpose, equality, addition/subtraction, scalar multiplication, transpose, multiplication, determinants, cofactor, adjoint, and inverse are provided in Table 6.4.

Table 6.4 Basic Operations of Matrices

Basic Matrix Algebra	Example
Trace	$\mathbf{F} = \begin{bmatrix} 4 & 13 & 3 \\ -2 & 19 & 1 \\ 3 & 2 & 0 \end{bmatrix}$ Trace (F) =Summation of diagonal element= 4+19+0=23
Equality	$\mathbf{A} = \begin{bmatrix} 4 & 13 \\ -2 & 19 \end{bmatrix}; \mathbf{B} = \begin{bmatrix} 4 & 13 \\ -2 & 19 \end{bmatrix}; \mathbf{C} = \begin{bmatrix} 4 & 13 & 2 \\ -2 & 19 & 1 \end{bmatrix} \therefore \mathbf{A} = \mathbf{B}; \mathbf{A} \neq \mathbf{C}$

Addition/Subtraction	$\mathbf{D} = \mathbf{A} + \mathbf{B} = \begin{bmatrix} 8 & 26 \\ -4 & 38 \end{bmatrix}$		
	$\mathbf{E} = \mathbf{A} - \mathbf{B} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$		
Scalar Multiplication	$2\mathbf{D} = \begin{bmatrix} 16 & 52\\ -8 & 76 \end{bmatrix}$		
Transpose, $[\bullet]^T$	$\mathbf{C} = \begin{bmatrix} 4 & 13 & 2 \\ -2 & 19 & 1 \end{bmatrix}; \mathbf{C}^T = \begin{bmatrix} 4 & -2 \\ 13 & 19 \\ 2 & 1 \end{bmatrix}$		
Matrix Multiplication	$\begin{bmatrix} 3 & 1 \\ 8 & 6 \\ 0 & 4 \\ (\text{Size } 3x2) \end{bmatrix} \begin{bmatrix} 5 & 9 \\ 7 & 2 \\ (\text{Size } 2x2) \end{bmatrix} = \begin{bmatrix} 22 & 29 \\ 82 & 84 \\ 28 & 8 \\ (\text{Size } 3x2) \end{bmatrix}$		
Note: AB ≠ BA	$\begin{bmatrix} 5 & 9 \\ 7 & 2 \\ (Size 2x2) \end{bmatrix} \begin{pmatrix} 3 & 1 \\ 8 & 6 \\ 0 & 4 \end{bmatrix} = \underbrace{error}_{(because non-equal interior dimensions)}$		
Determinant, •	$\mathbf{A} = \begin{bmatrix} 4 & 13 \\ -2 & 19 \end{bmatrix}; \ \mathbf{A} = \begin{vmatrix} 4 & 13 \\ -2 & 19 \end{vmatrix} = 4(19) - (-2)(13) = 102$		
Note: $ \bullet $ is determinant, not absolute in this case. Note: It is inefficient to calculate determinant manually for 4x4	$\mathbf{F} = \begin{bmatrix} 4 & 13 & 3 \\ -2 & 19 & 1 \\ 3 & 2 & 0 \end{bmatrix}; \ \mathbf{G} = \begin{bmatrix} 4 & 13 & 3 & 3 \\ -2 & 19 & 1 & 1 \\ 3 & 2 & 0 & 0 \\ 0 & 2 & 1 & 0 \end{bmatrix}$ $ \mathbf{F} = \begin{vmatrix} 4 & 13 & 3 & 3 \\ -2 & 19 & 1 & 1 \\ 3 & 2 & 0 \end{vmatrix} = 4 \begin{vmatrix} 19 & 1 \\ 2 & 0 \end{vmatrix} - 13 \begin{vmatrix} -2 & 1 \\ 3 & 0 \end{vmatrix} + 3 \begin{vmatrix} -2 & 19 \\ 3 & 2 \end{vmatrix} = -152$ $ \mathbf{G} = \begin{vmatrix} 4 & 13 & 3 & 3 \\ -2 & 19 & 1 & 1 \\ 3 & 2 & 0 & 0 \\ 0 & 2 & 1 & 0 \end{vmatrix}$		
matrix and above.	$= 4 \begin{vmatrix} 19 & 1 & 1 \\ 2 & 0 & 0 \\ 2 & 1 & 0 \end{vmatrix} - 13 \begin{vmatrix} -2 & 1 & 1 \\ 3 & 0 & 0 \\ 0 & 1 & 0 \end{vmatrix} + 3 \begin{vmatrix} -2 & 19 & 1 \\ 3 & 2 & 0 \\ 0 & 2 & 0 \end{vmatrix} - 3 \begin{vmatrix} -2 & 19 & 1 \\ 3 & 2 & 0 \\ 0 & 2 & 1 \end{vmatrix}$ $= 4(2) - 13(3) + 3(6) - 3(-55) = 152$		
Cofactor & Adjoint Note: adjoint = cofactor ^T	$\mathbf{A} = \begin{bmatrix} 4 & 13 \\ -2 & 19 \end{bmatrix}; \text{ cofactor}(\mathbf{A}) = \begin{bmatrix} 19 & - -2 \\ - 13 & 4 \end{bmatrix} = \begin{bmatrix} 19 & 2 \\ -13 & 4 \end{bmatrix}$ adjoint(\mathbf{A}) = (\cofactor(\mathbf{A}))^T = \begin{bmatrix} 19 & 2 \\ -13 & 4 \end{bmatrix}^T = \begin{bmatrix} 19 & -13 \\ 2 & 4 \end{bmatrix}		

Note: It is inefficient	
to calculate cofactor	
& adjoint manually	$\mathbf{F} = \begin{bmatrix} -2 & 19 & 1 \end{bmatrix}$; cofactor(\mathbf{F})= $\begin{bmatrix} -13 & 3 \\ 2 & 0 \end{bmatrix}$ $\begin{bmatrix} 4 & 3 \\ 2 & 0 \end{bmatrix}$ $\begin{bmatrix} -4 & 13 \\ 2 & 2 \end{bmatrix}$ =
for 4x4 matrix and	$\begin{bmatrix} 1 & 2 & 0 \end{bmatrix}$ $\begin{bmatrix} 1 & 2 & 0 & 13 & 01 & 13 & 21 \\ 113 & 31 & 14 & 31 & 14 & 131 \end{bmatrix}$
above.	$\begin{bmatrix} 1 & 1 & - \\ 19 & 1 & - \\ -2 & 1 & -2 & 19 \end{bmatrix}$
	$\begin{bmatrix} -2 & 3 & -61 \\ 6 & -9 & 31 \\ -44 & -10 & 102 \end{bmatrix}; \text{adjoint}(\mathbf{F}) = \begin{bmatrix} -2 & 6 & -44 \\ 3 & -9 & -10 \\ -61 & 31 & 102 \end{bmatrix}$
	$\mathbf{G} = \begin{bmatrix} 4 & 13 & 3 & 3 \\ -2 & 19 & 1 & 1 \\ 3 & 2 & 0 & 0 \\ 0 & 2 & 1 & 0 \end{bmatrix};$
	$\operatorname{cofactor}(\mathbf{G}) = \begin{bmatrix} \begin{vmatrix} 2 & 0 & 0 \\ 2 & 1 & 0 \\ 2 & 1 & 0 \end{vmatrix} & \begin{vmatrix} 4 & 3 & 3 \\ 2 & 0 & 0 \\ 2 & 1 & 0 \end{vmatrix} & \begin{vmatrix} 4 & 3 & 3 \\ 3 & 0 & 0 \\ 2 & 1 & 0 \end{vmatrix} & \begin{vmatrix} 4 & 3 & 3 \\ 3 & 0 & 0 \\ 2 & 1 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 3 & 3 \\ 3 & 2 & 0 \\ 2 & 1 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 3 & 3 \\ 3 & 2 & 0 \\ 0 & 1 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 3 & 3 \\ -2 & 1 & 0 \\ 13 & 3 & 3 \\ 19 & 1 & 1 \\ 2 & 1 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 3 & 3 \\ -2 & 1 & 1 \\ 0 & 1 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 3 & 3 \\ -2 & 1 & 1 \\ 0 & 1 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 3 & 3 \\ -2 & 1 & 0 \\ -2 & 1 & 1 \\ 2 & 0 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 3 & 3 \\ -2 & 1 & 1 \\ -2 & 1 & 0 \end{vmatrix} = \begin{vmatrix} 4 & 13 & 3 \\ -2 & 1 & 0 \\ -2 & 19 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1 \\ -2 & 10 & 1$
	$\operatorname{cofactor}(\mathbf{G}) = \begin{bmatrix} 2 & -3 & 6 & 55 \\ -6 & 9 & -18 & -13 \\ 44 & 10 & -20 & -82 \\ 0 & 0 & 152 & -152 \end{bmatrix}; \operatorname{adjoint}(\mathbf{G}) = \begin{bmatrix} 2 & -6 & 44 & 0 \\ -3 & 9 & 10 & 0 \\ 6 & -18 & -20 & 152 \\ 55 & -13 & -82 & -152 \end{bmatrix}$
Inverse, $[\bullet]^{-1}$	$\mathbf{A} = \begin{bmatrix} 4 & 13 \\ -2 & 19 \end{bmatrix}; \mathbf{A}^{-1} = \frac{1}{ \mathbf{A} } adjoint(\mathbf{A}) = \frac{1}{102} \begin{bmatrix} 19 & -13 \\ 2 & 4 \end{bmatrix}$
	$\mathbf{F} = \begin{bmatrix} 4 & 13 & 3 \\ -2 & 19 & 1 \\ 3 & 2 & 0 \end{bmatrix}; \ \mathbf{F}^{-1} = \frac{1}{ \mathbf{F} } adjoint(\mathbf{F}) = \frac{1}{-152} \begin{bmatrix} -2 & 6 & -44 \\ 3 & -9 & -10 \\ -61 & 31 & 102 \end{bmatrix}$
	$\mathbf{G} = \begin{bmatrix} 4 & 13 & 3 & 3\\ -2 & 19 & 1 & 1\\ 3 & 2 & 0 & 0\\ 0 & 2 & 1 & 0 \end{bmatrix}; \mathbf{G}^{-1} = \frac{1}{ \mathbf{G} } adjoint(\mathbf{G}) = \frac{1}{152} \begin{bmatrix} 2 & -6 & 44 & 0\\ -3 & 9 & 10 & 0\\ 6 & -18 & -20 & 152\\ 55 & -13 & -82 & -152 \end{bmatrix}$
	$0.1x_1 + 7x_2 - 0.3x_3 = -19.3$
	$3x_1 - 0.1x_2 - 0.2x_3 = 7.85$
	$0.3x_1 - 0.2x_2 + 10x_3 = 71.4$
	(i) Conventional Matrix form
	$\begin{bmatrix} 0.1 & 7 & -0.3 \\ 2 & 0.1 & 0.2 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \begin{pmatrix} -19.3 \\ 7 & 0.5 \end{bmatrix}$
	$\begin{bmatrix} 5 & -0.1 & -0.2 \\ 0.3 & -0.2 & 10 \end{bmatrix} \begin{bmatrix} x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 7.85 \\ 714 \end{bmatrix}$
	(ii) Augmented Matrix form
	$\begin{bmatrix} 0.1 & 7 & -0.3 & -19.3 \\ 3 & -0.1 & -0.2 & 7.85 \\ 0.3 & -0.2 & 10 & 71.4 \end{bmatrix}$

6.2 Solving Non-Homogeneous System of Linear Equations

Multicomponent systems result in *n* set(s) of mathematical equations that must be solved simultaneously. It can be represented by the following matrix format: $[A]{X} = {B}$. If ${B} \neq {0}$, it is known as non-homogeneous system of linear equations. In this study, several methods used to solve the unknown ${X}$ by using the [A] & non-zero ${B}$ will be discussed next.

Linear Algebraic Equations	Coefficient Matrix, $[A]$	Unknown, {X}	Non-zero { <i>B</i> }
$4x_1 + 13x_2 = 8$ $-2x_1 + 19x_2 = 2$ n=2 where 2 sets of eqns. are given to solve $x_1 \& x_2$ respectively.	$\begin{bmatrix} 4 & 13 \\ -2 & 19 \end{bmatrix}$	${x_1 \\ x_2}$	{ ⁸ ₂ }
$\begin{array}{c} 0.5x_1 + 2.5x_2 - 9x_3 = -6\\ -4.5x_1 + 3.5x_2 - 2x_3 = 5\\ -8x_1 - 9x_2 + 22x_3 = 2\\ \end{array}$ n=3 where 3 sets of eqns. are given to solve x_1 , x_2 and x_3 respectively.	$\begin{bmatrix} 0.5 & 2.5 & -9 \\ -4.5 & 3.5 & -2 \\ -8 & -9 & 22 \end{bmatrix}$	$ \begin{cases} x_1 \\ x_2 \\ x_3 \end{cases} $	$ \begin{pmatrix} -6\\5\\2 \end{pmatrix} $

For $n \leq 3$, methods frequently used to solve the non-homogeneous system of linear equations are given below:

- (i) Matrix Inversion (Moderate efficiency for n = 3 & high efficiency for n = 2)
- (ii) Graphical method (Less efficiency but useful for visualizing & enhancing intuition)
- (iii) Cramer's rule (High efficiency for $n \le 3$ **Main Focus**)
- (iv) Method of elimination (Less efficiency)

However, method (i)- (iv) are less efficiency for n > 3, thus more advanced methods are introduced:

- (a) Gaussian Elimination (Naïve vs Partial Pivoting) --- Main Focus
- (b) LU Decomposition --- Out of scope
- (c) Thomas algorithm --- Out of scope
- (d) Gauss Seidel Method --- Out of scope

6.2.1 Matrix Inversion Approach

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

If [A] is a square and non-singular matrix, $[A][A]^{-1} = [A]^{-1}[A] = [I]$

$$\{X\} = [A]^{-1}\{B\}$$

$$\mathbf{A} = \begin{bmatrix} 4 & 13 \\ -2 & 19 \end{bmatrix}; \mathbf{A}^{-1} = \frac{1}{|\mathbf{A}|} adjoint(\mathbf{A}) = \frac{1}{102} \begin{bmatrix} 19 & -13 \\ 2 & 4 \end{bmatrix}; \{X\} = \frac{1}{102} \begin{bmatrix} 19 & -13 \\ 2 & 4 \end{bmatrix} \{ 8 \\ 2 \end{bmatrix} = \{ \begin{array}{c} 126/102 \\ 24/102 \\ 24/102 \\ 3 \end{bmatrix}$$

6.2.2 Graphical Method

Rearrange the equations into linear plot format and then plot it.

Original linear equations	$a_{11}x_1 + a_{12}x_2 = b_1$	$3x_1 + 2x_2 = 18$
	$a_{21}x_1 + a_{22}x_2 = b_2$	$-x_1 + 2x_2 = 2$
Linear plot format	$x_2 = -\left(\frac{a_{11}}{a}\right)x_1 + \left(\frac{b_1}{a}\right)$	$x_2 = -\left(\frac{3}{2}\right)x_1 + \left(\frac{18}{2}\right)$
$x_2 = mx_1 + c$ where $m = slope$; $c = intercept$	$x_2 = -\left(\frac{a_{21}}{a_{22}}\right)x_1 + \left(\frac{b_2}{a_{22}}\right)$	$x_2 = -\left(\frac{-1}{2}\right)x_1 + \left(\frac{2}{2}\right)$

Using graphical method, the solution that satisfies both equations is the intersection point.



For singular system, the slopes of the equations are equal (or zero determinant), and it leads to

- (a) No solution case when there is no interception between the lines or
- (b) Infinite solutions case when there are infinite interception points between the lines.

(a) No solution case	(b) Infinite solutions case
$x_{2} = +\left(\frac{1}{2}\right)x_{1} + (1)$ $x_{2} = +\left(\frac{1}{2}\right)x_{1} + \left(\frac{1}{2}\right)$	$x_{2} = +\left(\frac{1}{2}\right)x_{1} + (1)$ $x_{2} = +\left(\frac{1}{2}\right)x_{1} + (1)$
x_2 $\frac{1}{2}x_1 + x_2 = 1$ $\frac{1}{2}x_1 + x_2 = \frac{1}{2}$ x_1	$\frac{x_2}{1 + \frac{x_1 + x_2}{2}}$
Diff of slope = $\begin{vmatrix} -\frac{1}{2} & 1 \\ -\frac{1}{2} & 1 \end{vmatrix} = -\frac{1}{2} - \left(-\frac{1}{2}\right) = 0$	Diff of slope $\begin{vmatrix} -\frac{1}{2} & 1 \\ -1 & 2 \end{vmatrix} = -1 - (-1) = 0$

For ill-conditioned system (also known as ill-posed system), the slopes of the equations are almost equal (or close-to-zero determinant), and it leads to



(c) Many solutions case and it is sensitive to round-off error.

6.2.3 Cramer's Rule

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

For example,

$$x_{1} = \frac{\begin{vmatrix} -0.01 & 0.52 & 1 \\ 0.5 & 1 & 1.9 \\ 0.67 & 1 & 1.9 \\ -0.44 & 0.3 & 0.5 \\ \hline 0.3 & 0.52 & 1 \\ 0.5 & 1 & 1.9 \\ 0.1 & 0.3 & 0.5 \\ \hline 0.5 & 1 & 1.9 \\ 0.5 & 1 & 0.5 \\ \hline 0.5 & 1 & 0.5 \\$$

Limitation: Impractical for eqns (n > 3).

6.2.4 Method of Elimination (Or Substitution Method)

$$\begin{bmatrix} 0.3 & 0.52 & 1 \\ 0.5 & 1 & 1.9 \\ 0.1 & 0.3 & 0.5 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -0.01 \\ 0.67 \\ -0.44 \end{bmatrix}$$

- Step 1: $x_1 = \cdots$ in $x_2 \& x_3$ terms for 1st eqn
- Step 2: Substitute $x_1 = \cdots$ to $2^{nd} \& 3^{rd}$ eqns.

Obtain $x_2 = \cdots$ in x_3 term

• Step 3: Substitute $x_2 = \cdots$ to 3rd eqn.

Obtain x_3 solution

• Step 4: Back Substitute to obtain $x_1 \& x_2$ solutions

Limitation: Extremely tedious to solve manually. However, the elimination approach can be extended and made more systematically to improve the efficiency such as Gauss Elimination method.

6.2.5 Naïve Gauss Elimination

It is an extension of the method of elimination which has a systematic scheme with forward elimination & back substitution procedure.

Forward elimination #1

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix} \quad \overrightarrow{\text{Forward elimination #1}} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a'_{22} & a'_{23} \\ 0 & a'_{32} & a'_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b'_2 \\ b'_3 \end{pmatrix}$$

R1 is the pivot equation, where a_{11} is the pivot element to turn a'_{21} & a'_{31} into 0

R2'=R2-R1x
$$f_{21}$$
 where factor, $f_{21} = \frac{a_{21}}{a_{11}}$; For example, $a'_{21} = a_{21} - a_{11}f_{21} = 0$
R3'=R3-R1x f_{31} where factor, $f_{31} = \frac{a_{31}}{a_{11}}$; For example, $a'_{31} = a_{31} - a_{11}f_{31} = 0$

Forward elimination #2

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a'_{22} & a'_{23} \\ 0 & a'_{32} & a'_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{cases} b_1 \\ b'_2 \\ b'_3 \end{pmatrix} \qquad \overline{\text{Forward elimination #2}} \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a'_{22} & a'_{23} \\ 0 & 0 & a''_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{cases} b_1 \\ b'_2 \\ b''_3 \end{pmatrix}$$

R2 is the pivot equation, where a_{22}' is the pivot element to turn a_{32}'' to be 0

R3''=R3'-R2'x
$$f_{32}$$
 where factor, $f_{32} = \frac{a'_{32}}{a'_{22}}$; For example, $a''_{32} = a'_{32} - a'_{22}f_{32} = 0$

Note: •' and •'' indicate change of value after first and second elimination procedures, respectively.

Back Substitution

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a'_{22} & a'_{23} \\ 0 & 0 & a''_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b'_2 \\ b''_3 \end{pmatrix} \quad \overrightarrow{Back substitution \#1} \qquad Solution, \ x_3 = \frac{b''_3}{a''_{33}}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a'_{22} & a'_{23} \\ 0 & 0 & a''_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b'_2 \\ b''_3 \end{pmatrix} \quad \overrightarrow{Back substitution \#2} \qquad Solution, \ x_2 = \frac{b'_2 - a'_{23} x_3}{a'_{22}}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ 0 & a'_{22} & a'_{23} \\ 0 & 0 & a''_{33} \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \\ b'_2 \\ b''_3 \end{pmatrix} \quad \overrightarrow{Back substitution \#3} \qquad Solution, \ x_1 = \frac{b_1 - a_{12} x_2 - a_{13} x_3}{a_{11}}$$

For example:

$$\begin{bmatrix} 3 & -0.1 & -0.2 \\ 0.1 & 7 & -0.3 \\ 0.3 & -0.2 & 10 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{cases} 7.85 \\ -19.3 \\ 71.4 \end{pmatrix}$$
Forward elimination #1

$$R2' = R2 - R1 \times \frac{0.1}{3}$$

$$R3' = R3 - R1 \times \frac{0.3}{3}$$
Forward elimination #2

$$R3'' = R3' - R2' \times \frac{-0.19}{7.00333}$$

$$\begin{bmatrix} 3 & -0.1 & -0.2 \\ 0 & 7.00333 & -0.293333 \\ 0 & -0.190000 & 10.0200 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{cases} 7.85 \\ -19.5617 \\ 70.6150 \end{pmatrix}$$
Forward elimination #2

$$R3'' = R3' - R2' \times \frac{-0.19}{7.00333}$$

$$\begin{bmatrix} 3 & -0.1 & -0.2 \\ 0 & 7.00333 & -0.293333 \\ 0 & 0 & 10.0120 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{cases} 7.85 \\ -19.5617 \\ 70.0843 \end{pmatrix}$$
Forward elimination #1

$$R3'' = R3' - R2' \times \frac{-0.19}{7.00333}$$

$$R3'' = R3' - R2' \times \frac{-0.19}{7.00333}$$

$$R3'' = 7.0000$$
Forward elimination #1

$$R3 = \frac{70.0843}{10.0120} = 7.0000$$
Forward elimination #2

$$R3 = \frac{7.85(-0.1)(-2.5)(-0.2)(7)}{3} = 3.00000$$
Forward elimination #3

$$R1 = \frac{7.85(-0.1)(-2.5)(-0.2)(7)}{3} = 3.00000$$

Limitation: Suffer the division by zero issue or the solution is sensitive to round-off error

For example,

$$\begin{bmatrix} 0 & 2 & 3 \\ 4 & 6 & 7 \\ 2 & 1 & 6 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 8 \\ -3 \\ 5 \end{pmatrix} \xrightarrow{\text{Forward elimination #1}} R2' = R2 - R1 \times \frac{4}{0} \text{ Error!}$$

$$R3' = R3 - R1 \times \frac{2}{0}$$

$$\begin{bmatrix} 2 & 100000 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 100000 \\ 2 \end{bmatrix} \xrightarrow{\text{Forward elimination #2}} R2' = R2 - R1 \times \frac{1}{2}$$
$$\begin{bmatrix} 2 & 100000 \\ -499999 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 100000 \\ -49998 \end{bmatrix} \overrightarrow{\text{Back substitution #1}} \quad x_2 = 1$$
$$\overrightarrow{\text{Back substitution #2}} \quad x_1 = \frac{100000 - 100000x_2}{2} = 0$$

Verification of solution:

LHS:RHS:
$$\begin{bmatrix} 2 & 100000 \\ 1 & 1 \end{bmatrix} \begin{cases} 0 \\ 1 \end{bmatrix} = \begin{cases} 100000 \\ 1 \end{bmatrix}$$
 $\begin{cases} 100000 \\ 2 \end{bmatrix}$ $\therefore LHS \neq RHS$ as percentage of error, $\begin{cases} \% \ error_b_1 \\ \% \ error_b_2 \end{cases} = \begin{cases} \frac{100000-100000}{100000} x100\% \\ \frac{2-1}{2} x100\% \end{cases} = \begin{cases} 0\% \\ 50\% \end{cases}$ Thus, $\begin{cases} x_1 \\ x_2 \end{bmatrix} = \begin{cases} 0 \\ 1 \end{bmatrix}$ is a poor solution as it is different from the actual solution. The solution is sensitive to the round-off error which leads to high error discrepancy.

6.2.5 Gauss Elimination with Partial Pivoting (GEwPP)

The limitation of Naïve Gauss Elimination can be improved by using GEwPP that consists of scaling analysis & pivoting strategy:

(a) Scaling analysis: Indicates the requirement of having pivoting to avoid divide by zero issue.

[2	100000 $\{x_1\} =$	{100000}	Scaling the coeffient matrix	0.00002	$\begin{bmatrix} 1\\1 \end{bmatrix} \begin{bmatrix} x_1\\x_2 \end{bmatrix} = \begin{bmatrix} 1\\2 \end{bmatrix}$	}
L1	$1](x_2)$	$\begin{pmatrix} 2 \end{pmatrix}$	to have max value of 1	pivot elem	ent is smalle	r

Rule of thumbs: If the pivot element is smaller than other rows, then pivoting is needed.

(b) Pivoting strategy: Switch row/ column to avoid pivot element to be zero or close to zero

(i) Naïve Gauss Elimination - Gaussian Elimination (GE) without pivoting strategy

 $\begin{bmatrix} 2 & 100000 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 100000 \\ 2 \end{bmatrix}$

Note: Previously we remain the original formulation and get poor solution after solving it.

(ii) **Gauss Elimination with Partial Pivoting (GEwPP)** -Switch row so that largest element is the pivot element (Main Focus).

 $\begin{bmatrix} 0.00002 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \overrightarrow{Partial \ pivoting} \quad \begin{bmatrix} 1 & 1 \\ 0.00002 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ $pivot \ element \ is \ the \ largest$

Example of GEwPP

 $\begin{bmatrix} 2 & 100000 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 100000 \\ 2 \end{bmatrix}$ Scaling $\begin{bmatrix} 0.00002 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ Note: Scaling indicates partial pivoting is needed $\overrightarrow{Partial Pivoting} \begin{bmatrix} 1 & 1 \\ 0.00002 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ Note: Pivot element is the largest after PP. $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$ Note: Round-off error happens when we use approximate value Back substitution #1 $x_2 = 1$ Back substitution #2 $x_1 = 2 - x_2 = 1$

Verification of solution:

LHS:RHS:
$$\begin{bmatrix} 2 & 100000\\ 1 & 1 \end{bmatrix} \begin{cases} 1\\ 1 \end{bmatrix} = \begin{cases} 100002\\ 2 \end{bmatrix}$$
 $\begin{cases} 100000\\ 2 \end{bmatrix}$ $\therefore LHS \approx RHS$ as percentage of error, $\begin{cases} \% \ error_b_1\\ \% \ error_b_2 \end{cases} = \begin{cases} \frac{|100000-100002|}{100000} x100\%\\ \frac{|2-2|}{2} x100\% \end{cases} = \begin{cases} 0.002\%\\ 0\% \end{cases}$ Thus, $\begin{cases} x_1\\ x_2 \end{bmatrix} = \begin{cases} 1\\ 1 \end{cases}$ is an accurate solution as it is close to the actual solution. The solution is less sensitive to the round-off error by using the GEwPP, as compares to naïve GE.

Determinant analysis can be done before GEwPP to know if you have well-conditioned system or singular system. Precaution: scaling is performed to standardize matrix before calculating determinant.

Well-conditioned system, $ \bullet \neq 0$	Singular system, $ \bullet = 0$	
$-1x_1 + 1x_2 + 2x_3 = 2$	$-1x_1 + 1x_2 + 2x_3 = 2$	$-1x_1 + 1x_2 + 2x_3 = 2$
$3x_1 - 1x_2 + 1x_3 = 6$	$3x_1 - 1x_2 + 1x_3 = 6$	$3x_1 - 1x_2 + 1x_3 = 6$
$-1x_1 + 3x_2 + 4x_3 = 4$	$-2x_1 + 2x_2 + 4x_3 = 4$	$-2x_1 + 2x_2 + 4x_3 = 8$
Unique solution for $\begin{cases} x_1 \\ x_2 \\ x_3 \end{cases}$ exists	$\begin{vmatrix} -\frac{1}{2} & \frac{1}{2} & 1\\ 1 & -\frac{1}{3} & \frac{1}{3}\\ -\frac{2}{4} & \frac{2}{4} & 1 \end{vmatrix} = 0$	
as $\begin{vmatrix} -\frac{1}{2} & \frac{1}{2} & 1 \\ 1 & -\frac{1}{3} & \frac{1}{3} \\ -\frac{1}{2} & \frac{3}{2} & 1 \end{vmatrix} = 0.4 \neq 0$	We get no solution or infinite solutions for $\begin{cases} x_1\\ x_2 \end{cases}$, we can know	
	either it is no solution or infinite solutions by using GEwPP.	

Rule of thumb: Assume that $-0.1 \le |\bullet| \le 0.1$ is considered as ill-conditioned system in this study.

Example: Solving a well-conditioned system using GEwPP

$$\begin{bmatrix} -1 & 1 & 2 \\ 3 & -1 & 1 \\ -1 & 3 & 4 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{cases} 2 \\ 6 \\ 4 \end{pmatrix}$$

$$\overline{\text{Scaling}} \begin{bmatrix} -1/2 & 1/2 & 1 & 1 \\ 1 & -1/3 & 1/3 & 2 \\ -1/4 & 3/4 & 1 & 1 \end{bmatrix} \xrightarrow{\text{Pivoting}} \begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ -1/2 & 1/2 & 1 & 1 \\ -1/4 & 3/4 & 1 & 1 \end{bmatrix}$$
Forward elimination #1
$$R2' = R2 - R1 \times \frac{(-\frac{1}{2})}{1} \qquad \begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ 0 & 1/3 & 7/6 & 2 \\ 0 & 2/3 & 13/12 & 1.5 \end{bmatrix}$$

$$R3' = R3 - R1 \times \frac{(-\frac{1}{4})}{1}$$

$$\overline{\text{Scaling}} \begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ 0 & 2/7 & 1 & 1 \\ 0 & 8/13 & 1 & 18/13 \\ 0 & 8/13 & 1 & 12/7 \end{bmatrix} \xrightarrow{\text{Pivoting}} \begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ 0 & 8/13 & 1 & 12/7 \\ 1 & 18/13 \end{bmatrix}$$
Forward elimination #2
$$R3'' = R3' - R2' \times \frac{(\frac{7}{2})}{(\frac{1}{23})} \qquad \begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ 0 & 8/13 & 1 & 12/7 \\ 0 & 8/13 & 1 & 18/13 \\ 0 & 0 & 15/28 & 15/14 \end{bmatrix}$$
Back substitution
$$x_3 = \frac{15/14}{15/28} = 2$$

$$x_2 = \frac{18/13 - (1)x_3}{8/3} = -1$$

$$x_1 = \frac{2 - (-1/3)x_2 - (1/3)x_3}{1} = 1$$

$$\therefore \begin{cases} x_1 \\ x_2 \\ x_2 \end{cases} = \begin{cases} -1 \\ 2 \\ 1 \\ 2 \\ 3 \end{bmatrix} \text{ is an accurate solution}$$

Example: Solving a singular system (infinite solutions case) using GEwPP

$\begin{bmatrix} -1 & 1 & 2 \\ 3 & -1 & 1 \\ -2 & 2 & 4 \end{bmatrix} \begin{pmatrix} 2 \\ 2 \\ 3 \end{pmatrix}$	$ \begin{cases} x_1 \\ x_2 \\ x_3 \end{cases} = \begin{cases} 2 \\ 4 \\ 4 \end{cases} $	}
$\overrightarrow{\text{Scaling}} \begin{bmatrix} -1/2 \\ 1 \\ -2/4 \end{bmatrix}$	1/2 -1/3 2/4	$ \begin{bmatrix} 1 & 1 \\ 1/3 & 2 \\ 1 & 1 \end{bmatrix} $
$\overrightarrow{\text{Pivoting}}\begin{bmatrix}1\\-1/2\\-2/4\end{bmatrix}$	-1/3 1/2 2/4	$\begin{array}{ccc} 1/3 & 2 \\ 1 & 1 \\ 1 & 1 \end{array}$

Forward elimination #1

$$R2' = R2 - R1 \times \frac{(-\frac{1}{2})}{1}$$

$$\begin{bmatrix} 1 & -1/3 & 1/3 & | & 2 \\ 0 & 1/3 & 7/6 & | & 2 \\ 0 & 1/3 & 7/6 & | & 2 \end{bmatrix}$$

$$R3' = R3 - R1 \times \frac{(-\frac{2}{4})}{1}$$

$$\begin{bmatrix} 1 & -1/3 & 1/3 & | & 2 \\ 0 & 1/3 & 7/6 & | & 2 \end{bmatrix}$$

 Forward elimination #2

$$\begin{bmatrix} 1 & -1/3 & 1/3 & | & 2 \\ 0 & 1/3 & 7/6 & | & 2 \end{bmatrix}$$

 R3'' = R3' - R2' $\times \frac{(\frac{1}{3})}{(\frac{1}{3})}$

$$\begin{bmatrix} 1 & -1/3 & 1/3 & | & 2 \\ 0 & 1/3 & 7/6 & | & 2 \\ 0 & 0 & 0 & | & 0 \end{bmatrix}$$

 Back substitution
 $0x_3 = 0$
 $x_3 = t$, where $-\infty \le t \le \infty$
 $x_2 = \frac{2 - (7/6)x_3}{1/3}$
 $x_1 = \frac{2 - (-1/3)x_2 - (1/3)x_3}{1}$

 Infinite solutions that can satisfy the eqns.

Example: Solving a singular system (no solution case) using GEwPP

$$\begin{bmatrix} -1 & 1 & 2 \\ 3 & -1 & 1 \\ -2 & 2 & 4 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 2 \\ 6 \\ 8 \end{pmatrix}$$

$$\boxed{\text{Scaling}} \begin{bmatrix} -1/2 & 1/2 & 1 & 1 \\ 1 & -1/3 & 1/3 & 2 \\ -2/4 & 2/4 & 1 & 2 \end{bmatrix}$$

$$\overrightarrow{\text{Pivoting}} \begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ -1/2 & 1/2 & 1 & 2 \\ -1/2 & 1/2 & 1 & 2 \\ -2/4 & 2/4 & 1 & 2 \end{bmatrix}$$

$$\overrightarrow{\text{Forward elimination #1}}$$

$$R2' = R2 - R1 \times \frac{(-\frac{2}{2})}{1}$$

$$\begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ 0 & 1/3 & 7/6 & 2 \\ 0 & 1/3 & 7/6 & 3 \end{bmatrix}$$

$$\overrightarrow{\text{R3'}} = R3 - R1 \times \frac{(-\frac{2}{4})}{1}$$

$$\begin{bmatrix} 1 & -1/3 & 1/3 & 2 \\ 0 & 1/3 & 7/6 & 3 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$\overrightarrow{\text{Back substitution}} \quad 0x_3 = 1 \qquad \therefore \text{ No solutions that satisfy the eqns.}$$

6.3 Row Echelon Form, Reduced Row Echelon Form, Rank, & Linear Dependency

After the GEwPP, the coefficient matrix will be in the **Row Echelon Form (REF)**. From the previous example, we obtain:

Well-conditioned system, $ \bullet \neq 0$	Singular system, $ \bullet = 0$	
$-1x_1 + 1x_2 + 2x_3 = 2$ $3x_1 - 1x_2 + 1x_3 = 6$ $-1x_1 + 3x_2 + 4x_3 = 4$	$-1x_1 + 1x_2 + 2x_3 = 2$ $3x_1 - 1x_2 + 1x_3 = 6$ $-2x_1 + 2x_2 + 4x_3 = 4$	$-1x_1 + 1x_2 + 2x_3 = 2$ $3x_1 - 1x_2 + 1x_3 = 6$ $-2x_1 + 2x_2 + 4x_3 = 8$
Coefficient matrix, $[A] = \begin{bmatrix} -1 & 1 & 2 \\ 3 & -1 & 1 \\ -1 & 3 & 4 \end{bmatrix}$	Coefficient matrix, $[A] = \begin{bmatrix} -3 \\ -3 \end{bmatrix}$	$ \begin{array}{cccc} 1 & 1 & 2 \\ 3 & -1 & 1 \\ 2 & 2 & 4 \end{array} $
Coefficient matrix after GEwPP is in REF ,	Coefficient matrix after GEwP	P is in REF ,
$[A]_{GEWPP} = \begin{bmatrix} 1 & -1/3 & 1/3 \\ 0 & 8/3 & 1 \\ 0 & 0 & 15/28 \end{bmatrix}$	$[A]_{GEwPP} = \left[\right]$	$ \begin{bmatrix} 1 & -1/3 & 1/3 \\ 0 & 1/3 & 7/6 \\ 0 & 0 & 0 \end{bmatrix} $

REF has the following characteristics:

- Zero row(s) are always below non-zero rows if there is any.
- Pivot element of the non-zero rows at the bottom must be at the right of the pivot element above it.
- Non-unique; can be in different scale

REF can be further reduced to **Reduced Row Echelon Form (RREF)** by using Gauss-Jordan Elimination with Partial Pivoting (GJEwPP) as shown in the example below:

$[A]_{GEWPP} = \begin{bmatrix} 1 & -1/3 & 1/3 \\ 0 & 8/3 & 1 \\ 0 & 0 & 15/28 \end{bmatrix}$	$[A]_{GEwPP} = \begin{bmatrix} 1 & -1/3 & 1/3 \\ 0 & 1/3 & 7/6 \\ 0 & 0 & 0 \end{bmatrix}$
Scale the pivot element to 1 $\overrightarrow{R2 \rightarrow R2 \times \frac{3}{8}}$ $\begin{bmatrix} 1 & -1/3 & 1/3 \\ 0 & 1 & 3/8 \\ 0 & 0 & 1 \end{bmatrix}$ $R3 \rightarrow R3 \times \frac{28}{15}$ $\begin{bmatrix} 0 & 1 & 3/8 \\ 0 & 0 & 1 \end{bmatrix}$	Scale the pivot element to 1 1 $-1/3$ $1/3$ $R2 \rightarrow R2 \times 3$ 0 1 3.5 $R3 \rightarrow R3 \times \frac{-1}{1.25}$ 0 0 0
Forward elimination $\begin{bmatrix} 1 & 0 & 11/24 \\ 0 & 1 & 3/8 \\ 0 & 0 & 1 \end{bmatrix}$ $R1 \rightarrow R1 - R2 \times \frac{(-\frac{1}{3})}{1}$ $\begin{bmatrix} 1 & 0 & 11/24 \\ 0 & 1 & 3/8 \\ 0 & 0 & 1 \end{bmatrix}$	Forward elimination $\begin{bmatrix} 1 & 0 & 1.5 \\ 0 & 1 & 3.5 \\ 0 & 0 & 0 \end{bmatrix}$ $R1 \rightarrow R1 - R2 \times \frac{(-\frac{1}{3})}{1}$ $\begin{bmatrix} 0 & 1.5 \\ 0 & 1 & 3.5 \\ 0 & 0 & 0 \end{bmatrix}$
Forward elimination $R1 \rightarrow R1 - R3 \times \frac{\binom{11}{24}}{1}$ $R2 \rightarrow R2 - R3 \times \frac{\binom{3}{8}}{1}$	

RREF has the following characteristics:

- Also a REF
- Unique; Scale the pivot element to 1
- Element above the pivot element is 0

Once RREF is obtained, **rank** of a matrix can be evaluated by counting the number of <u>non-zero rows</u> of RREF. Note: Rank is the maximum number of **linearly independent vector**.

Well-conditioned system, $ \bullet \neq 0$	Singular system, $ \bullet = 0$
$REF = \begin{bmatrix} 1 & -1/3 & 1/3 \\ 0 & 8/3 & 1 \\ 0 & 0 & 15/28 \end{bmatrix}$	$REF = \begin{bmatrix} 1 & -1/3 & 1/3 \\ 0 & 1/3 & 7/6 \\ 0 & 0 & 0 \end{bmatrix}$
$RREF = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$	$RREF = \begin{bmatrix} 1 & 0 & 1.5 \\ 0 & 1 & 3.5 \\ 0 & 0 & 0 \end{bmatrix}$
Rank=3	Rank=2
It means that all the 3 equations given are linear independent, therefore finding 3 unknowns from 3 linear independent equations are possible.	It means that only 2 out of 3 equations given are linear independent, therefore finding 3 unknowns from 2 linear independent equations are difficult.
$-1x_1 + 1x_2 + 2x_3 = 2$ $3x_1 - 1x_2 + 1x_3 = 6$ $-1x_1 + 3x_2 + 4x_3 = 4$	$-1x_1 + 1x_2 + 2x_3 = 2$ $3x_1 - 1x_2 + 1x_3 = 6$ $-2x_1 + 2x_2 + 4x_3 = 4$
\therefore [A] = Full rank matrix	\therefore [A] = Rank – deficient matrix

As a rule of thumbs, n linearly independent equations are required to solve n unknowns. To solve 3 unknowns, if we have less than 3 linearly independent equations, i.e. more unknowns than knowns, then we get the singular system issue.

6.4 Engineering Application of Non-Homogeneous System of Linear Equations

(a) Transform information into multiple linear algebraic equations to be solved simultaneously.

The amo rubber compone shown	ounts of needed ents types in the	metal, pl for #1, #2, a followin	astic, and electrical and #3 are g Table.	If totals of 2120, 43.4 and 164 g of metal, plastic, and rubber respectively are available each day. How many components can be produced per day?
Component	Metal (g/ component)	Plastic (g/ component) 0.25	Rubber (g/ component)	$15Comp_1 + 17Comp_2 + 19Comp_3 = 2120$ $0.25Comp_1 + 0.33Comp_2 + 0.42Comp_3 = 43.4$ $1.0Comp_1 + 1.2Comp_2 + 1.6Comp_3 = 164$
2 3 Note: It is information equations of	17 19 important fo n into mu & matrix forr	0.33 0.42 or student to ltiple linea mat.	1.2 1.6	$\begin{bmatrix} 15 & 17 & 19 \\ 0.25 & 0.33 & 0.42 \\ 1.0 & 1.2 & 1.6 \end{bmatrix} \begin{cases} Comp_1 \\ Comp_2 \\ Comp_3 \end{cases} = \begin{cases} 2120 \\ 43.4 \\ 164 \end{cases}$ Then, it can be solved by using the GEwPP, Cramer's rule, etc.

(b) Electrical system



$\begin{bmatrix} R_A + R_B & -R_B & -R_A \\ -R_B & R_B + R_C & -R_C \\ -R_A & -R_C & R_A + R_C + R_D \end{bmatrix} \begin{bmatrix} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \begin{cases} +V_1 \\ -V_2 \\ +V_3 \end{bmatrix}$ Eg. If the resistance, *R* and voltage, *V* are given, estimate the output currents, *I* of the 3 dof electrical circuit

$$\begin{bmatrix} 4+2 & -2 & -4 \\ -2 & 2+8 & -8 \\ -4 & -8 & 4+6+8 \end{bmatrix} \begin{cases} I_1 \\ I_2 \\ I_3 \end{bmatrix} = \begin{cases} 16 \\ -40 \\ 0 \end{cases}$$

Note: The derivation of the eqns involves theory of circuit, thus it is not examined in this study.

(c) Mechanical vibration system

Imp I	Assume $f_1 = F_1 cos \omega t$, $f_2 = F_2 cos \omega t$, $x_1 = X_1 cos \omega t$, $x_2 = X_2 cos \omega t$, $\ddot{x}_1 = -X_1 \omega^2$, $\ddot{x}_2 = -X_2 \omega^2$, $\omega = 10$ $\begin{bmatrix} k_1 + k_2 - \omega^2 m_1 & -k_2 \\ -k_2 & k_2 + k_3 - \omega^2 m_2 \end{bmatrix} \begin{cases} X_1 \\ X_2 \end{cases} = \begin{cases} F_1 \\ F_2 \end{cases}$ Eg. If the stiffness, k , mass, m , force, F , and excitation frequency, ω are given, estimate the output response of the 2 dof mass-spring vibration system. $\begin{bmatrix} 400 - 10^2(40) & -200 \\ -200 & 400 - 10^2(40) \end{bmatrix} \begin{cases} X_1 \\ X_2 \end{cases} = \begin{cases} F_1 \\ F_2 \end{cases}$ Note: The derivation of the eqns involves theory of vibration, thus it is not examined in this study.
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(d) Dynamic system

 $\begin{bmatrix} m_1 & 1 & 0 \\ m_2 & -1 & 1 \\ m_3 & 0 & -1 \end{bmatrix} \begin{pmatrix} a \\ T \\ R \end{pmatrix} = \begin{pmatrix} m_1g - c_1v \\ m_2g - c_2v \\ m_3g - c_3v \end{pmatrix}$ Eg. If the mass, m, drag coeffient, c, and free fall velocity, v are given, estimate the output tension & acceleration of the 3 dof falling parachutists. $\begin{bmatrix} 70 & 1 & 0 \\ 60 & -1 & 1 \\ 40 & 0 & -1 \end{bmatrix} \begin{pmatrix} a \\ T \\ R \end{pmatrix} = \begin{cases} 70(9.81) - 10(5) \\ 60(9.81) - 14(5) \\ 40(9.81) - 17(5) \end{pmatrix}$ Note: The derivation of the eqns involves theory of dynamic, thus it is not examined in this study.

Advanced applications of matrix algebra including transformation matrix, image processing, signal processing, finite element simulation, page rank algorithm, Hill Cipher encryption, etc. Thus, mastering matrix algebra is important and it has huge impact.