# 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.



Table 6.1 Matrix, column vector and row vector.

where  $a_{mn}$  is the element of the matrix at  $m^{\text{th}}$  row and  $n^{\text{th}}$  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.



Table 6.3 Type of matrices.



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







## 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.



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  $n \leq 3$  **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,  $[A][A]^{-1} = [A]^{-1}[A] = [I]$ 

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

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

# 6.2.2 Graphical Method

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



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.



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

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

$$
x_1 = \frac{\begin{vmatrix} b_1 & a_{12} & a_{13} \ b_2 & a_{22} & a_{23} \ b_3 & a_{32} & a_{33} \end{vmatrix}}{\begin{vmatrix} A_1 \ B_2 \ B_3 \end{vmatrix}}, \qquad x_2 = \frac{\begin{vmatrix} a_{11} & b_1 & a_{13} \ a_{21} & b_2 & a_{23} \ a_{31} & b_3 & a_{33} \end{vmatrix}}{\begin{vmatrix} A_1 \ B_2 \ B_3 \end{vmatrix}}, \qquad x_3 = \frac{\begin{vmatrix} a_{11} & a_{12} & b_1 \ a_{21} & a_{22} & b_2 \ a_{31} & a_{32} & b_3 \end{vmatrix}}{\begin{vmatrix} A \ B \end{vmatrix}}
$$

For example,

$$
\begin{bmatrix}\n0.3 & 0.52 & 1 \\
0.5 & 1 & 1.9 \\
0.1 & 0.3 & 0.5\n\end{bmatrix}\n\begin{bmatrix}\nx_1 \\
x_2 \\
x_3\n\end{bmatrix} = \n\begin{bmatrix}\n-0.01 \\
0.67 \\
-0.44\n\end{bmatrix}
$$
\n
$$
x_1 = \frac{\begin{vmatrix}\n-0.01 & 0.52 & 1 \\
0.67 & 1 & 1.9 \\
-0.44 & 0.3 & 0.5\n\end{vmatrix}}{\begin{vmatrix}\n0.3 & 0.52 \\
0.5 & 1 & 1.9 \\
0.5 & 1 & 1.9 \\
0.5 & 1 & 1.9\n\end{vmatrix}} = -14.9, \quad x_2 = \frac{\begin{vmatrix}\n0.3 & -0.01 & 1 \\
0.5 & 0.67 & 1.9 \\
0.1 & -0.44 & 0.5\n\end{vmatrix}}{\begin{vmatrix}\n0.3 & 0.52 & 1 \\
0.5 & 1 & 1.9 \\
0.5 & 1 & 1.9 \\
0.5 & 1 & 1.9\n\end{vmatrix}} = -29.5, \quad x_3 = \frac{\begin{vmatrix}\n0.3 & 0.52 & -0.01 \\
0.5 & 1 & 0.67 \\
0.1 & 0.3 & -0.44 \\
0.5 & 1 & 1.9 \\
0.5 & 1 & 1.9\n\end{vmatrix}} = 19.8
$$

Limitation: Impractical for eqns ( $n > 3$ ).

#### 6.2.4 Method of Elimination (Or Substitution Method)

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

- Step 1:  $x_1 = \cdots$  in  $x_2 \& x_3$  terms for 1<sup>st</sup> 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 3<sup>rd</sup> 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

$$
\begin{bmatrix} a_{11} & a_{12} & a_{13} \ a_{21} & a_{22} & a_{23} \ a_{31} & a_{32} & a_{33} \end{bmatrix} \begin{pmatrix} x_1 \ x_2 \ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \ b_2 \ b_3 \end{pmatrix}
$$
 Forward elimination #1 
$$
\begin{bmatrix} a_{11} & a_{12} & a_{13} \ 0 & a_{22}^{\prime} & a_{23}^{\prime} \ a_{32} & a_{33}^{\prime} \end{bmatrix} \begin{pmatrix} x_1 \ x_2 \ x_3 \end{pmatrix} = \begin{pmatrix} b_1 \ b_2 \ b_3 \end{pmatrix}
$$

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

R2' = R2-R1xf<sub>21</sub> where factor, 
$$
f_{21} = \frac{a_{21}}{a_{11}}
$$
 ; For example,  $a'_{21} = a_{21} - a_{11}f_{21} = 0$   
R3' = R3-R1xf<sub>31</sub> where factor,  $f_{31} = \frac{a_{31}}{a_{11}}$  ; For example,  $a'_{31} = a_{31} - a_{11}f_{31} = 0$ 

Forward elimination #2

$$
\begin{bmatrix} a_{11} & a_{12} & a_{13} \ 0 & a_{22}' & a_{23}' \ 0 & a_{32}' & a_{33}' \end{bmatrix} \begin{bmatrix} x_1 \ x_2 \ x_3 \end{bmatrix} = \begin{Bmatrix} b_1 \ b_2' \ b_3' \end{Bmatrix} \quad \text{Forward elimination #2} \quad \begin{bmatrix} a_{11} & a_{12} & a_{13} \ 0 & a_{22}' & a_{23}' \ 0 & 0 & a_{33}' \end{bmatrix} \begin{bmatrix} x_1 \ x_2 \ x_3 \end{bmatrix} = \begin{Bmatrix} b_1 \ b_2' \ b_3' \end{Bmatrix}
$$

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

R3''=R3'-R2'xf<sub>32</sub> 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

$$
\begin{bmatrix} a_{11} & a_{12} & a_{13} \ a'_{22} & a'_{23} \ a''_{33} \end{bmatrix} \begin{bmatrix} x_1 \ x_2 \ x_3 \end{bmatrix} = \begin{Bmatrix} b_1 \ b'_2 \ b''_3 \end{Bmatrix}
$$
  
\nBack substitution #1  
\nSolution,  $x_3 = \frac{b_3''}{a_3''}$   
\n
$$
\begin{bmatrix} a_{11} & a_{12} & a_{13} \ a'_{22} & a'_{23} \ a''_{33} \end{bmatrix} \begin{bmatrix} x_1 \ x_2 \ x_3 \end{bmatrix} = \begin{Bmatrix} b_1 \ b'_2 \ b''_3 \end{Bmatrix}
$$
  
\nBack substitution #2  
\nSolution,  $x_2 = \frac{b_2' - a_{23}' x_3}{a_{22}'}$   
\nSolution,  $x_2 = \frac{b_2' - a_{23}' x_3}{a_{22}'}$   
\n
$$
\begin{bmatrix} a_{11} & a_{12} & a_{13} \ a'_{22} & a'_{23} \ a'_{33} \end{bmatrix} \begin{bmatrix} x_1 \ x_2 \ x_3 \end{bmatrix} = \begin{Bmatrix} b_1 \ b'_2 \ b''_3 \end{Bmatrix}
$$
  
\nBack substitution #3  
\nSolution,  $x_1 = \frac{b_1 - a_{12}x_2 - a_{13}x_3}{a_{11}'}$ 

For example:

$$
\begin{bmatrix}\n3 & -0.1 & -0.2 \\
0.1 & 7 & -0.3 \\
0.3 & -0.2 & 10\n\end{bmatrix}\n\begin{bmatrix}\nx_1 \\
x_2 \\
x_3\n\end{bmatrix} =\n\begin{bmatrix}\n7.85 \\
-19.3 \\
71.4\n\end{bmatrix}
$$
\n
$$
\begin{bmatrix}\n3 & -0.1 & -0.2 \\
7.00333 & -0.293333 \\
0 & 7.00333 & -0.293333 \\
10 & -0.190000 & 10.0200\n\end{bmatrix}\n\begin{bmatrix}\nx_1 \\
x_2 \\
x_3\n\end{bmatrix} =\n\begin{bmatrix}\n7.85 \\
-19.5617 \\
70.6150\n\end{bmatrix}
$$
\n
$$
\begin{bmatrix}\n83' = R3 - R1 \times \frac{0.3}{3} & 0 \\
R3' = R3 - R1 \times \frac{0.3}{3} & 0 \\
R3'' = R3' - R2' \times \frac{-0.19}{3} & 0 \\
R3'' = R3' - R2' \times \frac{-0.19}{7.00333} & 0 \\
R3'' = R3' - R2' \times \frac{-0.19}{7.00333} & 0 \\
R3'' = \frac{70.0843}{10.0120} = 7.0000\n\end{bmatrix}
$$
\n
$$
\begin{bmatrix}\n3 & -0.1 & -0.2 \\
0 & 7.00333 & -0.293333 \\
10 & 0 & 10.0120\n\end{bmatrix}\n\begin{bmatrix}\nx_1 \\
x_2 \\
x_3\n\end{bmatrix} =\n\begin{bmatrix}\n7.85 \\
-19.5617 \\
70.0843\n\end{bmatrix}
$$
\n
$$
\begin{bmatrix}\n3 & -0.1 & -0.2 \\
0 & 7.00333 & -0.293333 \\
0 & 0 & 10.0120\n\end{bmatrix}\n\begin{bmatrix}\nx_1 \\
x_2 \\
x_3\n\end{bmatrix} =\n\begin{bmatrix}\n7.85 \\
-19.5617 \\
70.0843\n\end{bmatrix}
$$
\n
$$
\begin{bmatrix}\n5 & -0.1 & -0.2 \\
0 & 7.00333 & -
$$

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

For example,

$$
\begin{bmatrix} 0 & 2 & 3 \\ 4 & 6 & 7 \\ 2 & 1 & 6 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{Bmatrix} 8 \\ -3 \\ 5 \end{Bmatrix} \xrightarrow[R2' = R2 - R1 \times \frac{4}{0} \text{ Error}] \text{Error}
$$

$$
\begin{bmatrix} 2 & 100000 \\ 1 & 1 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix} = \begin{Bmatrix} 100000 \\ 2 \end{Bmatrix} \frac{\text{Forward elimination #2}}{\text{R2'} = \text{R2} - \text{R1} \times \frac{1}{2}}
$$
  

$$
\begin{bmatrix} 2 & 100000 \\ 0 & -49999 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix} = \begin{Bmatrix} 100000 \\ -49998 \end{Bmatrix} \frac{\text{Back substitution #1}}{\text{Back substitution #2}} \quad x_2 = 1
$$
  

$$
\text{Back substitution #2} \quad x_1 = \frac{100000 - 100000x_2}{2} = 0
$$

Verification of solution:

LHS:  
\n
$$
\begin{bmatrix}\n2 & 100000 \\
1 & 1\n\end{bmatrix}\n\begin{bmatrix}\n0 \\
1\n\end{bmatrix} =\n\begin{bmatrix}\n100000 \\
1\n\end{bmatrix}
$$
\n
$$
\begin{bmatrix}\n2 & 100000 \\
1 & 1\n\end{bmatrix}\n\begin{bmatrix}\n0 \\
1\n\end{bmatrix} =\n\begin{bmatrix}\n100000 \\
1\n\end{bmatrix}
$$
\n
$$
\begin{bmatrix}\n\frac{100000 - 100000}{100000}x100\% \\
\frac{2-1}{2}x100\% \\
\frac{2-1}{2}x100\% \\
\end{bmatrix} =\n\begin{bmatrix}\n0\% \\
50\% \\
50\% \\
\end{bmatrix}
$$
\nThus,  $\begin{bmatrix}\nx_1 \\
x_2\n\end{bmatrix} = \begin{bmatrix}\n0 \\
1\n\end{bmatrix}$  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.



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

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

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).

 $\begin{bmatrix} 0.00002 & 1 \\ 1 & 1 \end{bmatrix}$  $\begin{bmatrix} 2 & 0 & 1 \\ 1 & 1 & 1 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix}$  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$  $\frac{1}{2}$ }  $\frac{1}{Partial}$   $\frac{1}{10.00002}$   $\frac{1}{10.00002}$   $\frac{1}{10}$  $\begin{bmatrix} 1 & 1 \\ 0.00002 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$  $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$ pivot element is the largest Example of GEwPP

 $\begin{bmatrix} 2 & 100000 \\ 1 & 1 \end{bmatrix}$  $\begin{bmatrix} 2 & 100000 \\ 1 & 1 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix}$  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 100000 \\ 2 \end{pmatrix}$  $\begin{bmatrix} 2 \end{bmatrix}$  $\overrightarrow{Scaling}$   $\begin{bmatrix} 0.00002 & 1 \\ 1 & 1 \end{bmatrix}$ 0002  $\begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 1 \\ 2 \end{pmatrix}$ 2 Note: Scaling indicates partial pivoting is needed  $\overrightarrow{Partual}$  Partial Pivoting  $\begin{bmatrix} 1 & 1 \\ 0 & 0 & 0 & 1 \end{bmatrix}$  $\begin{bmatrix} 1 & 1 \\ 0.00002 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ 1 } Note: Pivot element is the largest after PP.  $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$  $\begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix} \begin{Bmatrix} x_1 \\ x_2 \end{Bmatrix}$  $\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$ 1 } Note: Round-off error happens when we use approximate value Back substitution  $#$  $x_2$  = 1 Back substitution #2  $x_1 = 2-x_2 = 1$ Verification of solution:

LHS:  
\nRHS:  
\n
$$
\left[\begin{array}{ccc}2&100000\\1&1\end{array}\right]\left\{\begin{array}{ccc}1\\1\end{array}\right\} = \left\{\begin{array}{ccc}100002\\2\end{array}\right\} & \left\{\begin{array}{ccc}100000\\2\end{array}\right\}
$$
\n
$$
\therefore LHS \approx RHS \text{ as percentage of error, } \begin{array}{ccc} \frac{96}{6} \text{ error\_b}_1 \\ \frac{96}{6} \text{ error\_b}_2 \end{array}\right\} = \left\{\begin{array}{ccc} \frac{|100000-100002|}{100000} \times 100\% \\ \frac{|2-2|}{2} \times 100\% \end{array}\right\} = \left\{\begin{array}{ccc} 0.002\% \\ 0\% \end{array}\right\}
$$
\nThus, 
$$
\left\{\begin{array}{ccc} x_1 \\ x_2 \end{array}\right\} = \left\{\begin{array}{ccc} 1 \\ 1 \end{array}\right\} \text{ is an accurate solution as it is close to the actual solution. The solution is less sensitive to the round-off error by using the GEwPP, as compares to naïve GE.}
$$

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.



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

Example: Solving a well-conditioned system using GEwPP

$$
\begin{bmatrix} -1 & 1 & 2 \ 3 & -1 & 1 \ 4 & 3 & 4 \end{bmatrix} \begin{bmatrix} x_1 \ x_2 \ x_3 \end{bmatrix} = \begin{Bmatrix} 2 \ 6 \ 4 \end{Bmatrix}
$$
  
\n
$$
\frac{1}{3}
$$
  
\n
$$
\frac{1}{
$$

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



Forward ⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗elimination ⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗#1⃗⃗⃗⃗ 1 (− ) 1 −1/3 1/3 2 2 2′ = 2 − 1 × 0 1/3 7/6 2 [ ] 1 2 0 1/3 7/6 2 (− ) 4 3′ = 3 − 1 × 1 Forward ⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗elimination ⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗#2⃗⃗⃗⃗ 1 −1/3 1/3 2 1 ( ) [ 0 1/3 7/6 2 ] 3 3′′ = 3′ − 2′ × 1 0 0 0 0 ( ) 3 Back ⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗substitution ⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗⃗ 0<sup>3</sup> = 0 2−(7/6) 2−(−1/3)[ ]−(1/3) 1 1/3 <sup>3</sup> = , ℎ − ∞ ≤ ≤ ∞ 1 2 ∴ { } = 2−(7/6) 3 1/3 2−(7/6)<sup>3</sup> <sup>2</sup> = } { 1/3 Infinite solutions that can satisfy the 2−(−1/3)2−(1/3)<sup>3</sup> eqns. <sup>1</sup> = 1 

Example: Solving a singular system (no solution case) using GEwPP  
\n
$$
\begin{bmatrix}\n-1 & 1 & 2 \\
3 & -1 & 1 \\
-2 & 2 & 4\n\end{bmatrix}\n\begin{bmatrix}\nx_1 \\
x_2 \\
x_3\n\end{bmatrix} = \n\begin{bmatrix}\n2 \\
6 \\
6\n\end{bmatrix}
$$
\n
$$
\frac{1}{\text{Scaling}}\n\begin{bmatrix}\n-1/2 & 1/2 & 1 & 1 \\
1 & -1/3 & 1/3 & 2 \\
-2/4 & 2/4 & 1 & 2\n\end{bmatrix}
$$
\n
$$
\frac{1}{\text{Pivoting}}\n\begin{bmatrix}\n1 & -1/3 & 1/3 & 2 \\
-1/2 & 1/2 & 1 & 1 \\
-2/4 & 2/4 & 1 & 2\n\end{bmatrix}
$$
\n
$$
\frac{1}{\text{Forward elimination #1}}
$$
\n
$$
R2' = R2 - R1 \times \frac{-\frac{1}{2}}{1}
$$
\n
$$
\begin{bmatrix}\n1 & -1/3 & 1/3 & 2 \\
0 & 1/3 & 7/6 & 2 \\
0 & 1/3 & 7/6 & 3\n\end{bmatrix}
$$
\n
$$
R3' = R3 - R1 \times \frac{-\frac{2}{4}}{1}
$$
\n
$$
\frac{1}{\text{Forward elimination #2}}
$$
\n
$$
R3'' = R3' - R2' \times \frac{\frac{1}{3}}{1}
$$
\n
$$
\frac{1}{(\frac{1}{3})}
$$
\n
$$
\frac{1}{(\frac{1}{3})}
$$
\n
$$
\frac{1}{\text{Back substitution}} \quad 0x_3 = 1 \quad \therefore \text{ No solutions that satisfy the eqns.}
$$

## 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:



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:



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 non-zero rows of RREF. Note: Rank is the maximum number of **linearly independent vector**.



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.



# (b) Electrical system





(d) Dynamic system



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.