

Row Echelon Form

After the Gauss elimination, the row echelon form of the augmented matrix will be

$$\underline{R} = \left[\begin{array}{cccc|c} r_{11} & r_{12} & \dots & r_{1n} & f_1 \\ & r_{22} & \dots & r_{2n} & f_2 \\ & 0 & \dots & r_{rn} & f_r \\ & & & 0 & f_{r+1} \\ & & & & f_m \end{array} \right] \quad r \leq m; r_{11} \neq 0$$

\underline{R} : Row reduced coefficient matrix \underline{R}

" r " : Rank of \underline{R} and also rank of \underline{A}

(a) No solution:

(i) $r < m$ and

(ii) at least one of the numbers $f_{r+1}, f_{r+2}, \dots, f_m$ is not zero.

$\underline{R}\underline{x} = \underline{f}$ is inconsistent then

(b) Unique solution:

System is consistent and $r = n$

(c) Infinitely many solutions

$r < m, f_{r+1} = f_{r+2} = \dots = f_m = 0$

7.4 Linear Independence (Rank of a Matrix.)

Linear Independence and Dependence of vectors

$$\underline{a}_i = [a_{i1} \ a_{i2} \ \dots \ a_{in}] \quad \underline{a}_1, \underline{a}_2, \dots, \underline{a}_m \rightarrow \text{row vectors (row vectors)}$$

$$c_1 \underline{a}_1 + c_2 \underline{a}_2 + \dots + c_m \underline{a}_m \rightarrow \text{Linear combination of the above vectors}$$

Consider the equation

$$c_1 \underline{a}_1 + c_2 \underline{a}_2 + \dots + c_m \underline{a}_m = \underline{0}$$

Trivial solution: all c_i 's are zero.

Linear independent: We cannot find a set of non zero c_i 's.

Linear dependent: At least one set of scalars $\{c_1, c_2, \dots, c_m\}$

or we can express \underline{a}_1 in terms of the other \underline{a}_i 's.

$$\underline{a}_1 = k_2 \underline{a}_2 + k_3 \underline{a}_3 + \dots + k_m \underline{a}_m \text{ where } k_j = -c_j/c_1.$$

Example:

$$\underline{a}_1 = [3 \ 0 \ 2 \ 2]$$

$$\underline{a}_2 = [-6 \ 42 \ 24 \ 54]$$

$$\underline{a}_3 = [21 \ -21 \ 0 \ -15]$$

$$\underline{b}_1 = 2\underline{a}_1 + \underline{a}_2 = [0 \ 42 \ 28 \ 58]$$

$$\underline{b}_2 = 7\underline{a}_1 - \underline{a}_3 = [0 \ -21 \ 14 \ 29]$$

$$\underline{b}_3 = \underline{b}_1 - 2\underline{b}_2 = [0 \ 0 \ 0 \ 0] = \underline{0}$$

$$\underline{b}_1 - 2\underline{b}_2 = \underline{0}$$

$$(2\underline{a}_1 + \underline{a}_2) - 2(7\underline{a}_1 - \underline{a}_3) = \underline{0}$$

$$-12\underline{a}_1 + \underline{a}_2 + 2\underline{a}_3 = \underline{0} \quad \rightarrow \text{Linearly dependent}$$

$$12\underline{a}_1 = \underline{a}_2 + 2\underline{a}_3 \Rightarrow \underline{a}_1 = \frac{1}{12}\underline{a}_2 + \frac{1}{6}\underline{a}_3$$

Rank of a Matrix

Maximum # of linearly independent row vectors of A is the rank of the matrix A .

Example: Consider the above set of row vectors given in the example: Rank is 2. (2 independent row vectors)

Row Equivalent Matrices

A matrix \underline{A}_1 row equivalent to a matrix \underline{A}_2 if \underline{A}_1 can be obtained from \underline{A}_2 by elementary row operations. We have infinitely many \underline{A}_1 .

Example:

$$\underline{A}_2 = \begin{bmatrix} 3 & 0 & 2 & 2 \\ -6 & 42 & 24 & 54 \\ 21 & -21 & 0 & -15 \end{bmatrix}$$

$$\underline{A}_1 = \begin{bmatrix} 3 & 0 & 2 & 2 \\ 0 & 42 & 28 & 58 \\ 0 & 21 & 14 & 29 \end{bmatrix}$$

$$\underline{A}'_1 = \begin{bmatrix} 3 & 0 & 2 & 2 \\ 0 & 42 & 28 & 58 \\ 0 & 0 & 0 & 0 \end{bmatrix} \Rightarrow \text{Rank} = 2$$

Rank in terms Column vectors

The rank r of a matrix \underline{A} equals the number of linearly independent column vectors \underline{A} .

Then \underline{A} and \underline{A}^T have the same rank.

Example

$$\underline{A}^T = \underline{A}'_1 = \begin{bmatrix} 3 & -6 & 21 \\ 0 & 42 & -21 \\ 2 & 24 & 0 \\ 2 & 54 & -15 \end{bmatrix} \Rightarrow \underline{A}_2 = \begin{bmatrix} 3 & -6 & 21 \\ 0 & 42 & -21 \\ 0 & 28 & -14 \\ 0 & 58 & -29 \end{bmatrix} \Rightarrow \underline{A}'_2 = \begin{bmatrix} 3 & -6 & 21 \\ 0 & 2 & -1 \\ 0 & 2 & -1 \\ 0 & 2 & -1 \end{bmatrix}$$

$$\Rightarrow \underline{A}''_2 = \begin{bmatrix} 3 & -6 & 21 \\ 0 & 2 & -1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \quad \text{Rank} : r = 2$$

Linear Dependence

Consider p vectors each having n components.

If $n < p$, then these vectors are linearly dependent.

Assume the matrix A has p vectors as row vectors having n components (A is a $n \times p$ matrix)

If $n < p$, the rank of A : $r \leq n < p$.

Vector Space

The vector space R^n consisting of all vectors with n components (n real numbers) has dimension n .

$$\underline{a}_1 = [1 \ 0 \ \dots \ 0]$$

$$\underline{a}_2 = [0 \ 1 \ \dots \ 0]$$

$$\vdots$$

$$\underline{a}_n = [0 \ 0 \ \dots \ 1]$$

The row vectors obtained by \underline{a}_i 's as

$$c_1 \underline{a}_1 + c_2 \underline{a}_2 + \dots + c_n \underline{a}_n$$

is called the row space of A .

Similarly the span of the column vectors of A is called the column space of A .

7.5 Solutions of Linear Systems

Consider the following linear system of "m" equations in "n" unknowns x_1, x_2, \dots, x_n .

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

⋮

$$a_{m1}x_1 + a_{m2}x_2 + \dots + a_{mn}x_n = b_m$$

with

$$\underline{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{bmatrix}, \quad \underline{\tilde{A}} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} & | & b_1 \\ a_{21} & a_{22} & \dots & a_{2n} & | & b_2 \\ \vdots & \vdots & \ddots & \vdots & | & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & | & b_m \end{bmatrix}$$

i. Existence

Above set of equations is consistent, that is, has solutions (unique or multiple), iff the coefficient matrix \underline{A} and the augmented matrix $\underline{\tilde{A}}$ have the same rank.

ii. Uniqueness

Above set of equations has precisely one solution iff

$$r = n.$$

iii - Infinitely many solutions if $r < n$.

iv - Not solution can be concluded using Gauss Elimination method.

Example:

$$3x_1 + 2x_2 + 2x_3 = 13$$

$$2x_1 + x_2 + x_3 = 7$$

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

$$\underline{A} = \begin{bmatrix} 3 & 2 & 2 \\ 2 & 1 & 1 \\ 1 & 2 & -1 \end{bmatrix} \quad \underline{\tilde{A}} = \begin{bmatrix} 3 & 2 & 2 & | & 13 \\ 2 & 1 & 1 & | & 7 \\ 1 & 2 & -1 & | & 2 \end{bmatrix}$$

$$\underline{\tilde{A}} = \begin{bmatrix} 3 & 2 & 2 & | & 13 \\ 2 & 1 & 1 & | & 7 \\ 1 & 2 & -1 & | & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & -1 & | & 2 \\ 2 & 1 & 1 & | & 7 \\ 3 & 2 & 2 & | & 13 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & -1 & | & 2 \\ 0 & 3 & -3 & | & -3 \\ 0 & 4 & -5 & | & -7 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 2 & -1 & | & 2 \\ 0 & 1 & -1 & | & -1 \\ 0 & 4 & -5 & | & -7 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & -1 & | & 2 \\ 0 & 1 & -1 & | & -1 \\ 0 & 0 & -1 & | & -3 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & -1 & | & 2 \\ 0 & 1 & -1 & | & -1 \\ 0 & 0 & 1 & | & 3 \end{bmatrix}$$

$$x_3 = 3$$

$$x_2 - x_3 = -1 \Rightarrow x_2 - 3 = -1 \Rightarrow x_2 = 2$$

$$x_1 + 2x_2 - x_3 = 2 \Rightarrow x_1 + 4 - 3 = 2 \Rightarrow x_1 = 1$$

Unique solution! $\{1, 2, 3\}$

Example

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

$$2x_1 + x_2 + x_3 = 7$$

$$3x_1 + 2x_2 + x_3 = 10$$

$$\underline{A} = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & 1 \\ 3 & 2 & 1 \end{bmatrix} \quad \underline{\tilde{A}} = \begin{bmatrix} 1 & 2 & -1 & | & 2 \\ 2 & 1 & 1 & | & 7 \\ 3 & 2 & 1 & | & 10 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 2 & -1 & | & 2 \\ 0 & 3 & -3 & | & -3 \\ 0 & 4 & -4 & | & -4 \end{bmatrix}$$

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

Multiple solution exists.

$$x_2 - x_3 = -1 \quad ; \quad x_3 = 2 \Rightarrow x_2 = 1$$

$$x_1 + 2x_2 - x_3 = 2 \Rightarrow x_1 + 2 - 2 = 2 \Rightarrow x_1 = 2$$

$$x_2 - x_3 = -1 \quad ; \quad x_3 = 3 \Rightarrow x_2 = 2$$

$$x_1 + 2x_2 - x_3 = 2 \Rightarrow x_1 + 4 - 3 = 2 \Rightarrow x_1 = 1$$

$\{2, 1, 2\}$ is
a solution

$\{1, 2, 3\}$ is also
a solution

7.6 Second and Third order Determinants

The determinant of a 2×2 matrix \underline{A} :

$$D = \det \underline{A} = \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} = a_{11}a_{22} - a_{12}a_{21}$$

Determinant is a useful tool to determine the solutions of linear system of equations

Consider linear system of two equations in two unknowns:

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

Hence

$$\underline{A} = \begin{bmatrix} a_{11} & a_{12} & | & b_1 \\ a_{21} & a_{22} & | & b_2 \end{bmatrix}$$

$$x_1 = \frac{\begin{vmatrix} b_1 & a_{12} \\ b_2 & a_{22} \end{vmatrix}}{D} = \frac{b_1 a_{22} - b_2 a_{12}}{a_{11} a_{22} - a_{12} a_{21}}$$

$$x_2 = \frac{\begin{vmatrix} a_{11} & b_1 \\ a_{21} & b_2 \end{vmatrix}}{D} = \frac{a_{11} b_2 - a_{21} b_1}{D}$$

(Replace the column with b)
; $D = a_{11}a_{22} - a_{12}a_{21}$

provided that $D \neq 0$.

If $D = 0$, then we have multiple solutions.

Proof: Solve the above equations. (H.W.)

Example : $4x_1 + 3x_2 = 10$
 $2x_1 + 5x_2 = 12$

$$x_1 = \frac{\begin{vmatrix} 10 & 3 \\ 12 & 5 \end{vmatrix}}{\begin{vmatrix} 4 & 3 \\ 2 & 5 \end{vmatrix}} = \frac{50 - 36}{20 - 6} = \frac{14}{14} = 1$$

$$x_2 = \frac{\begin{vmatrix} 4 & 10 \\ 2 & 12 \end{vmatrix}}{\begin{vmatrix} 4 & 3 \\ 2 & 5 \end{vmatrix}} = \frac{48 - 20}{14} = \frac{28}{14} = 2$$

Third order Determinants

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

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

$$= a_{11}(a_{22}a_{33} - a_{32}a_{23}) - a_{21}(a_{12}a_{33} - a_{32}a_{13}) + a_{31}(a_{12}a_{23} - a_{22}a_{13})$$

Linear System of Three Equations

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

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

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

$$x_1 = \frac{D_1}{D}, \quad x_2 = \frac{D_2}{D}, \quad x_3 = \frac{D_3}{D}$$

$$D_1 = \begin{vmatrix} b_1 & a_{12} & a_{13} \\ b_2 & a_{22} & a_{23} \\ b_3 & a_{32} & a_{33} \end{vmatrix}, \quad D_2 = \begin{vmatrix} a_{11} & b_1 & a_{13} \\ a_{21} & b_2 & a_{23} \\ a_{31} & b_3 & a_{33} \end{vmatrix}, \quad D_3 = \begin{vmatrix} a_{11} & a_{12} & b_1 \\ a_{21} & a_{22} & b_2 \\ a_{31} & a_{32} & b_3 \end{vmatrix}$$

Note :

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

$$= a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{13}a_{32}) + a_{13}(a_{12}a_{23} - a_{13}a_{22})$$

$$= a_{11}(a_{22}a_{33} - a_{23}a_{32}) - a_{12}(a_{21}a_{33} - a_{31}a_{23}) + a_{13}(a_{21}a_{32} - a_{31}a_{22})$$

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

So first column or first row may be used. The same approach for other columns or rows.

Example :

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

$$2x_1 + x_2 + x_3 = 7$$

$$3x_1 + 2x_2 + 2x_3 = 13$$

$$D = \begin{vmatrix} 1 & 2 & -1 \\ 2 & 1 & 1 \\ 3 & 2 & 2 \end{vmatrix} = (1)(2 \cdot 2) - 2(4 + 3) + (3)(2 + 1) = -12 + 9 = -3$$

$$D_1 = \begin{vmatrix} 2 & 2 & -1 \\ 7 & 1 & 1 \\ 13 & 2 & 2 \end{vmatrix} = 2(2 \cdot 2) - 7(4 + 2) + 13(2 + 1) = -42 + 39 = -3$$

$$D_2 = \begin{vmatrix} 1 & 2 & -1 \\ 2 & 7 & 1 \\ 3 & 13 & 2 \end{vmatrix} = (1)(14 - 13) - 2(4 + 13) + 3(2 + 7) \\ = 1 - 34 + 27 = -6$$

$$D_3 = \begin{vmatrix} 1 & 2 & 2 \\ 2 & 1 & 7 \\ 3 & 2 & 13 \end{vmatrix} = (1)(13 - 14) - 2(26 - 4) + 3(14 - 2) = \\ = -1 - 44 + 36 = -9$$

$$x_1 = \frac{D_1}{D} = \frac{-3}{-3} = 1; \quad x_2 = \frac{-6}{-3} = 2, \quad x_3 = \frac{-9}{-3} = 3$$

7.7 Cramer's Rule : (293)

Consider an $n \times n$ (square) matrix

$$D = \det \underline{A} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{bmatrix}$$

(i) $n=1 \Rightarrow D = a_{11}$

(ii) $n \geq 2$

$$D = a_{j1} C_{j1} + a_{j2} C_{j2} + \dots + a_{jn} C_{jn} \quad \begin{array}{l} \text{(any row)} \\ j=1, 2, \dots \text{ or } n \end{array}$$

or

$$D = a_{1k} C_{1k} + a_{2k} C_{2k} + \dots + a_{nk} C_{nk} \quad \begin{array}{l} \text{(any column)} \\ (k=1, 2, \dots \text{ or } n) \end{array}$$

$$C_{jk} = (-1)^{j+k} M_{jk}$$

where M_{jk} is a determinant of order $n-1$.

$$D = \sum_{k=1}^n a_{jk} (-1)^{j+k} M_{jk} \quad (j=1, 2, \dots \text{ or } n = \text{any row})$$

or

$$D = \sum_{j=1}^n a_{jk} (-1)^{j+k} M_{jk} \quad (k=1, 2, \dots \text{ or } n = \text{any column})$$

Example : $n=3$; $j=2$

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

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

$$C_{21} = (-1)^3 M_{21} = -M_{21}$$

$$D = a_{21} C_{21} + a_{22} C_{22} + a_{23} C_{23}$$

$$C_{22} = (-1)^4 M_{22} = M_{22}$$

$$= -a_{21} M_{21} + a_{22} M_{22} - a_{23} M_{23}$$

$$C_{23} = (-1)^5 M_{23} = -M_{23}$$

Sign pattern:

$$\begin{bmatrix} + & - & + \\ - & + & - \\ + & - & + \end{bmatrix}$$

Example - Triangular Matrix

$$D = \begin{vmatrix} -3 & 0 & 0 \\ 4 & 2 & 0 \\ 2 & 1 & 6 \end{vmatrix} = -3 \begin{vmatrix} 2 & 0 \\ 1 & 6 \end{vmatrix} = (-3)(12) = -36$$

Properties:

- i. Interchange of two rows multiplies the value of the determinant by -1 .
- ii. Addition of a multiple of a row to another row does not alter the value of the determinant.
- iii. Multiplication of a row by a nonzero constant c , multiplies the value of the determinant by c .
- iv. An $n \times n$ matrix A has a rank n iff $\det A \neq 0$.
- v. Above statements are true for column operations.

Example: $A = \begin{bmatrix} 4 & 6 \\ 2 & 4 \end{bmatrix}$

i. $\det A = \begin{vmatrix} 4 & 6 \\ 2 & 4 \end{vmatrix} = 16 - 12 = 4$

ii. $\det(A') = \begin{vmatrix} 4 & 6 \\ 0 & 4 \end{vmatrix} = 4$

iii. $\det A = \begin{vmatrix} 4 & 6 \\ 2 & 4 \end{vmatrix} = 2 \begin{vmatrix} 2 & 3 \\ 2 & 4 \end{vmatrix} = (2)(8-6) = 4$
 $= (2)(2) \begin{vmatrix} 2 & 3 \\ 1 & 2 \end{vmatrix} = (4)(1) = 4$

iv. $\det(A) = \begin{vmatrix} 4 & 6 \\ 2 & 4 \end{vmatrix} = 2 \begin{vmatrix} 2 & 3 \\ 2 & 4 \end{vmatrix} = 2 \begin{vmatrix} 2 & 3 \\ 0 & 1 \end{vmatrix} = (2)(2) = 4$

7.8 Inverse of a Matrix

$$\underline{A}\underline{A}^{-1} = \underline{A}^{-1}\underline{A} = \underline{I}$$

where

$$\underline{I} = \begin{bmatrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \\ \vdots & & \ddots & \\ 0 & & & 1 \end{bmatrix} \quad (n \times n \text{ unit matrix})$$

If \underline{A} has an inverse, then \underline{A} is called a nonsingular matrix.

If \underline{A} has no inverse, then \underline{A} is called a singular matrix.

If \underline{A} has an inverse, the inverse is unique.

The importance of \underline{A}^{-1} :

$$\underline{A}\underline{x} = \underline{b}$$

$$\underline{A}^{-1}\underline{A}\underline{x} = \underline{A}^{-1}\underline{b} \Rightarrow \underline{I}\underline{x} = \underline{A}^{-1}\underline{b}$$

$$\underline{x} = \underline{A}^{-1}\underline{b}$$

Gauss-Jordan Elimination

$$\underline{A} = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & 1 \\ 3 & 2 & 2 \end{bmatrix}$$

Not consider the augmented matrix:

$$[\underline{A} \ \underline{I}] = \left[\begin{array}{ccc|ccc} 1 & 2 & -1 & 1 & 0 & 0 \\ 2 & 1 & 1 & 0 & 1 & 0 \\ 3 & 2 & 2 & 0 & 0 & 1 \end{array} \right]$$

$$\begin{array}{l} -2R_1 + R_2 \rightarrow R_2 \\ -3R_1 + R_3 \rightarrow R_3 \end{array} \rightarrow \left[\begin{array}{ccc|ccc} 1 & 2 & -1 & 1 & 0 & 0 \\ 0 & -3 & 3 & -2 & 1 & 0 \\ 0 & -4 & 5 & -3 & 0 & 1 \end{array} \right]$$

$$\begin{array}{l} -\frac{4}{3}R_2 + R_3 \rightarrow R_3 \\ -3R_3 + R_2 \rightarrow R_2 \\ R_3 + R_1 \rightarrow R_1 \end{array} \rightarrow \left[\begin{array}{ccc|ccc} 1 & 2 & -1 & 1 & 0 & 0 \\ 0 & -3 & 3 & -2 & 1 & 0 \\ 0 & 0 & 1 & -\frac{1}{3} & -\frac{4}{3} & \frac{1}{3} \end{array} \right]$$

$$\rightarrow \left[\begin{array}{ccc|ccc} 1 & 2 & 0 & 2/3 & -4/3 & 1 \\ 0 & -3 & 0 & -1 & 5 & -3 \\ 0 & 0 & 1 & -1/3 & -4/3 & 1 \end{array} \right] \xrightarrow{-\frac{1}{3}R_2 \rightarrow R_2}$$

$$\rightarrow \left[\begin{array}{ccc|ccc} 1 & 2 & 0 & 2/3 & -4/3 & 1 \\ 0 & -1 & 0 & -1/3 & 5/3 & -1 \\ 0 & 0 & 1 & -1/3 & -4/3 & 1 \end{array} \right] \xrightarrow{\begin{array}{l} 2R_2 + R_1 \rightarrow R_1 \\ -R_2 \rightarrow R_2 \end{array}}$$

$$\rightarrow \left[\begin{array}{ccc|ccc} 1 & 0 & 0 & 0 & 2 & -1 \\ 0 & 1 & 0 & 1/3 & -5/3 & 1 \\ 0 & 0 & 1 & -1/3 & -4/3 & 1 \end{array} \right]$$

I
 A^{-1}

Verification

$$\underline{AA^{-1}} = \begin{bmatrix} 1 & 2 & -1 \\ 2 & 1 & 1 \\ 3 & 2 & 2 \end{bmatrix} \begin{bmatrix} 0 & 2 & -1 \\ 1/3 & -5/3 & 1 \\ -1/3 & -4/3 & 1 \end{bmatrix} = \begin{bmatrix} \frac{2}{3} + \frac{1}{3} & 2 - \frac{10}{3} + \frac{4}{3} & -1 + 2 - 1 \\ \frac{1}{3} - \frac{1}{3} & 4 - \frac{5}{3} - \frac{4}{3} & \dots \\ \frac{2}{3} - \frac{2}{3} & \dots & -3 + 2 + 2 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$