

LINEAR OPTIMIZATION

LECTURE 5

06/05/2021

Goals:

Understanding the set of solutions to $Ax = b$

$A: m \times n$

$x: n \times 1$

$b: m \times 1$

To show:

$\{x \mid Ax = b\}$ is an affine subspace of \mathbb{R}^n



$\{x \mid Ax = 0\}$ is a subspace of \mathbb{R}^n



$$\dim \{x \mid Ax = 0\} = n - r \quad (r: \text{no. of linearly independent columns of } A)$$

$$= n - r \quad (r: \text{no. of linearly independent rows of } A)$$

Part 1: $\dim \{x : Ax = 0\} = n - k$ (k : no. of linearly independent columns of A)

Consider the system $Ax = 0$.

$$a_{11} x_1 + a_{12} x_2 + \dots + a_{1n} x_n = 0$$

$$a_{21} x_1 + a_{22} x_2 + \dots + a_{2n} x_n = 0$$

:

$$a_{m1} x_1 + a_{m2} x_2 + \dots + a_{mn} x_n = 0$$

Define $A^i = \begin{bmatrix} a_{1i} \\ a_{2i} \\ \vdots \\ a_{mi} \end{bmatrix}$ the i^{th} column of A

The system $Ax = 0$ can be rewritten as:

$$x_1 A^1 + x_2 A^2 + \dots + x_n A^n = \bar{0}$$

Notice that A^1, A^2, \dots, A^n are in \mathbb{R}^m

The system $Ax=0$ can be rewritten as:

$$x_1 A^1 + x_2 A^2 + \dots + x_n A^n = \bar{0}$$

Column space of A:

A^1, A^2, \dots, A^n are the column vectors of A.

The subspace $\text{span}(A^1, A^2, \dots, A^n)$ is called column space of A.

Our hypothesis: dimension of column space = k

- There are k linearly independent columns in A.

↳ every other column can be written as a linear combination of these k columns.

Assume A^1, A^2, \dots, A^k are linearly independent. Other columns can be written as a combination of these columns.

$$A^{k+1} = \beta_{1,k+1} A^1 + \beta_{2,k+1} A^2 + \dots + \beta_{k,k+1} A^k$$

:

$$A^n = \beta_{1,n} A^1 + \beta_{2,n} A^2 + \dots + \beta_{k,n} A^k$$

Consider again the system $Ax=0$ which we view as:

$$x_1 A^1 + x_2 A^2 + \dots + x_n A^n = 0 \quad \text{--- } \oplus$$

Recall that we are interested in showing that

$$\dim \{x : Ax=0\} = n-k.$$

Substituting $(*)$ in \oplus gives:

$$\begin{aligned} & (x_1 + \beta_{1,k+1} x_{k+1} + \beta_{1,k+2} x_{k+2} + \dots + \beta_{1,n} x_n) A^1 \\ & + (x_2 + \beta_{2,k+1} x_{k+1} + \beta_{2,k+2} x_{k+2} + \dots + \beta_{2,n} x_n) A^2 \\ & + \dots \\ & + (x_k + \beta_{k,k+1} x_{k+1} + \beta_{k,k+2} x_{k+2} + \dots + \beta_{k,n} x_n) A^k \\ & = 0 \end{aligned}$$

(*)

Since A^1, \dots, A^k are linearly independent, each of the coefficients:

$$x_i + \beta_{i,k+1} x_{k+1} + \dots + \beta_{i,n} x_n = 0, \quad i=1, \dots, k$$

$$\begin{aligned}
 & (x_1 + \beta_{1,k+1} x_{k+1} + \beta_{1,k+2} x_{k+2} + \dots + \beta_{1,n} x_n) A^1 \\
 & + (x_2 + \beta_{2,k+1} x_{k+1} + \beta_{2,k+2} x_{k+2} + \dots + \beta_{2,n} x_n) A^2 \\
 & + \dots \\
 & + (x_k + \beta_{k,k+1} x_{k+1} + \beta_{k,k+2} x_{k+2} + \dots + \beta_{k,n} x_n) A^k \\
 & = 0
 \end{aligned}$$

(*)

Since A^1, \dots, A^k are linearly independent, each of the coefficients:

$$x_i + \beta_{i,k+1} x_{k+1} + \dots + \beta_{i,n} x_n = 0, \quad i=1, \dots, k$$

* Therefore, every solution x to $Ax=0$ satisfies:

$$x_1 = -\beta_{1,k+1} x_{k+1} - \beta_{1,k+2} x_{k+2} \dots - \beta_{1,n} x_n$$

$$x_2 = -\beta_{2,k+1} x_{k+1} - \beta_{2,k+2} x_{k+2} \dots - \beta_{2,n} x_n$$

:

$$x_k = -\beta_{k,k+1} x_{k+1} - \beta_{k,k+2} x_{k+2} \dots - \beta_{k,n} x_n$$

* Similarly, every solution to the above system (in blue) is a solution to $Ax=0$

x is a solution to $Ax=0$ iff

$$x_1 = -\beta_{1,k+1} x_{k+1} - \beta_{1,k+2} x_{k+2} \dots \rightarrow \beta_{1,n} x_n$$

$$x_2 = -\beta_{2,k+1} x_{k+1} - \beta_{2,k+2} x_{k+2} \dots - \beta_{2,n} x_n$$

:

$$x_k = -\beta_{k,k+1} x_{k+1} - \beta_{k,k+2} x_{k+2} \dots - \beta_{k,n} x_n$$

- Notice that once we fix a value for x_{k+1}, \dots, x_n , the values for $x_1 \dots x_k$ get fixed.
- This indicates a dimension of $n-k$ for $\{x : Ax=0\}$. We will now exhibit a basis of size $n-k$.

x is a solution to $Ax=0$ iff

$$x_1 = -\beta_{1,k+1} x_{k+1} - \beta_{1,k+2} x_{k+2} \dots \rightarrow \beta_{1,n} x_n$$

$$x_2 = -\beta_{2,k+1} x_{k+1} - \beta_{2,k+2} x_{k+2} \dots - \beta_{2,n} x_n$$

:

$$x_k = -\beta_{k,k+1} x_{k+1} - \beta_{k,k+2} x_{k+2} \dots - \beta_{k,n} x_n$$

define $n-k$ vectors: $U_{k+1}, U_{k+2}, \dots, U_n$ as follows:

- U_j , for $k+1 \leq j \leq n$, is obtained by setting $x_j = 1$ and the rest in x_{k+1}, \dots, x_n to 0.

$$U_{k+1} = \begin{bmatrix} -\beta_{1,k+1} \\ -\beta_{2,k+1} \\ \vdots \\ -\beta_{k,k+1} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$U_{k+2} = \begin{bmatrix} -\beta_{1,k+2} \\ -\beta_{2,k+2} \\ \vdots \\ -\beta_{k,k+2} \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

$$\dots$$

$$U_n = \begin{bmatrix} -\beta_{1,n} \\ -\beta_{2,n} \\ \vdots \\ -\beta_{k,n} \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

Claim 1: $U_{k+1}, U_{k+2}, \dots, U_n$ are linearly independent.

Claim 2: $U_{k+1}, \dots, U_n \in \{x : Ax=0\}$

Claim 3: Every solution to $Ax=0$ can be written as a linear combination of U_{k+1}, \dots, U_n .

$$U_{k+1} = \begin{bmatrix} -\beta_{1,k+1} \\ -\beta_{2,k+1} \\ \vdots \\ -\beta_{k,k+1} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad U_{k+2} = \begin{bmatrix} -\beta_{1,k+2} \\ -\beta_{2,k+2} \\ \vdots \\ -\beta_{k,k+2} \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad \dots \quad U_n = \begin{bmatrix} -\beta_{1,n} \\ -\beta_{2,n} \\ \vdots \\ -\beta_{k,n} \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

Claim 1: U_{k+1}, \dots, U_n are linearly independent.

$$\text{Suppose } \alpha_{k+1} U_{k+1} + \dots + \alpha_n U_n = 0$$

$\alpha_{k+1} = 0$ due to the $k+1^{\text{th}}$ coordinate in U_{k+1}

\vdots

$\alpha_n = 0$ due to the n^{th} coordinate in U_n

$$U_{k+1} = \begin{bmatrix} -\beta_{1,k+1} \\ -\beta_{2,k+1} \\ \vdots \\ -\beta_{k,k+1} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad U_{k+2} = \begin{bmatrix} -\beta_{1,k+2} \\ -\beta_{2,k+2} \\ \vdots \\ -\beta_{k,k+2} \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad \dots \quad U_n = \begin{bmatrix} -\beta_{1,n} \\ -\beta_{2,n} \\ \vdots \\ -\beta_{k,n} \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

Claim 2: $AU_{k+1} = AU_{k+2} = \dots = AU_n = 0$.

Recall:

x is a solution to $Ax=0$ iff

$$x_1 = -\beta_{1,k+1} x_{k+1} - \beta_{1,k+2} x_{k+2} \dots \rightarrow \beta_{1,n} x_n$$

$$x_2 = -\beta_{2,k+1} x_{k+1} - \beta_{2,k+2} x_{k+2} \dots - \beta_{2,n} x_n$$

\vdots

$$x_k = -\beta_{k,k+1} x_{k+1} - \beta_{k,k+2} x_{k+2} \dots - \beta_{k,n} x_n$$

Notice that $U_{k+1}, U_{k+2}, \dots, U_n$ satisfy the above equations.

$$U_{k+1} = \begin{bmatrix} -\beta_{1,k+1} \\ -\beta_{2,k+1} \\ \vdots \\ -\beta_{k,k+1} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad U_{k+2} = \begin{bmatrix} -\beta_{1,k+2} \\ -\beta_{2,k+2} \\ \vdots \\ -\beta_{k,k+2} \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad \dots \quad U_n = \begin{bmatrix} -\beta_{1,n} \\ -\beta_{2,n} \\ \vdots \\ -\beta_{k,n} \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

$$x_1 = -\beta_{1,k+1} x_{k+1} - \beta_{1,k+2} x_{k+2} \dots \rightarrow \beta_{1,n} x_n$$

$$x_2 = -\beta_{2,k+1} x_{k+1} - \beta_{2,k+2} x_{k+2} \dots - \beta_{2,n} x_n$$

:

$$x_k = -\beta_{k,k+1} x_{k+1} - \beta_{k,k+2} x_{k+2} - \dots - \beta_{k,n} x_n$$

$$x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad \text{s.t.} \quad Ax = 0$$

$$\begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \boxed{?} U_{k+1} + \boxed{?} U_{k+2} + \dots + \boxed{!} U$$

$$U_{k+1} = \begin{bmatrix} -\beta_{1,k+1} \\ -\beta_{2,k+1} \\ \vdots \\ -\beta_{k,k+1} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad U_{k+2} = \begin{bmatrix} -\beta_{1,k+2} \\ -\beta_{2,k+2} \\ \vdots \\ -\beta_{k,k+2} \\ 0 \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \quad \dots \quad U_n = \begin{bmatrix} -\beta_{1,n} \\ -\beta_{2,n} \\ \vdots \\ -\beta_{k,n} \\ 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

Claim 3: Every solution to $Ax=0$ can be written as a linear combination of U_{k+1}, \dots, U_n .

Let x be a solution to $Ax=0$. It satisfies:

$$x_1 = -\beta_{1,k+1} x_{k+1} - \beta_{1,k+2} x_{k+2} \dots - \beta_{1,n} x_n$$

$$x_2 = -\beta_{2,k+1} x_{k+1} - \beta_{2,k+2} x_{k+2} \dots - \beta_{2,n} x_n$$

⋮

$$x_k = -\beta_{k,k+1} x_{k+1} - \beta_{k,k+2} x_{k+2} \dots - \beta_{k,n} x_n$$

Hence $x =$

$$\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_k \\ x_{k+1} \\ \vdots \\ x_n \end{bmatrix} = x_{k+1} U_{k+1} + x_{k+2} U_{k+2} + \dots + x_n U_n$$

Claims 1, 2, 3 show that U_{k+1}, \dots, U_n forms a basis for $\{x : Ax=0\}$

- Hence $\dim(\{x : Ax=0\}) = n-k$.

Summary of Part 1:

- $Ax=0$ rewritten as $x_1 A^1 + x_2 A^2 + \dots + x_n A^n = 0$
- A^1, A^2, \dots, A^k are linearly independent. Write A^{k+1}, \dots, A^n in terms of A^1, \dots, A^k .

$$\begin{aligned} A^{k+1} &= \beta_{1,k+1} A^1 + \beta_{2,k+1} A^2 + \dots + \beta_{k,k+1} A^k \\ &\vdots \\ A^n &= \beta_{1,n} A^1 + \beta_{2,n} A^2 + \dots + \beta_{k,n} A^k \end{aligned}$$

- From this, show that x is a soln. to $Ax=0$ iff.

$$\begin{aligned} x_1 &= -\beta_{1,k+1} x_{k+1} - \beta_{1,k+2} x_{k+2} - \dots - \beta_{1,n} x_n \\ x_2 &= -\beta_{2,k+1} x_{k+1} - \beta_{2,k+2} x_{k+2} - \dots - \beta_{2,n} x_n \\ &\vdots \\ x_k &= -\beta_{k,k+1} x_{k+1} - \beta_{k,k+2} x_{k+2} - \dots - \beta_{k,n} x_n \end{aligned}$$

- Use this to exhibit a basis U_{k+1}, \dots, U_n .

Next goal:

$$\dim \{x \mid Ax = 0\} = n - r \quad (r: \text{no. of linearly independent columns of } A)$$

$$= n - r \quad (r: \text{no. of linearly independent rows of } A)$$

To show: $\dim \{x \mid Ax = 0\} = n - r$

A preprocessing step:

We have:

$$\begin{array}{l} a_{11} x_1 + a_{12} x_2 + \dots + a_{1n} x_n = 0 \\ \vdots \\ a_{m1} x_1 + a_{m2} x_2 + \dots + a_{mn} x_n = 0 \end{array}$$

Define $A_1 = [a_{11} \ a_{12} \ a_{13} \ \dots \ a_{1n}]^T$

$A_2 = [a_{21} \ a_{22} \ \dots \ a_{2n}]^T$

\vdots
 $A_m = [a_{m1} \ a_{m2} \ \dots \ a_{mn}]^T$

Hypothesis: Rows A_1, A_2, \dots, A_r are linearly independent.
- Other rows A_{r+1}, \dots, A_n can be written as a combination of A_1, \dots, A_r .

Removing a row that is a combination of other rows does not change the solution set.

Example:

$$2x_1 + 3x_2 + 5x_3 + 4x_4 = 0$$

$$3x_1 + 2x_2 + 7x_3 + x_4 = 0$$

$$5x_1 + 5x_2 + 12x_3 + 5x_4 = 0$$

Removing 3rd row preserves solution set.

The pre processing step:

$$A_1 x = 0$$

$$A_2 x = 0$$

:

$$A_r x = 0$$

$$\underline{A_m x = 0}$$

:

$$\underline{\underline{A_m x = 0}}$$

linearly independent

Remove the equations $A_{r+1} x = 0, \dots, A_m x = 0$.

- Now we perform Gaussian elimination on the remaining rows to simplify further.

Gaussian elimination:

Involves repeated application of two operations:

- Exchange rows R_i and R_j
- $R_i \leftarrow R_i + \alpha \cdot R_j$ - add a multiple of some row to another row.

Both these operations preserve the set of solutions.

Consider our given set of r equations:

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & & \\ a_{r1} & a_{r2} & \dots & a_{rn} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

- Gaussian elimination on

$$\begin{bmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & & & \\ a_{r1} & a_{r2} & \dots & a_{rn} \end{bmatrix}$$

Gaussian elimination:

$$\left[\begin{array}{cccc} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & \dots & \dots & a_{2n} \\ \vdots & & & \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{array} \right] \rightarrow \left[\begin{array}{cccc} a_{11} & a_{12} & \dots & a_{1n} \\ 0 & a'_{22} & \dots & a'_{2n} \\ 0 & & \ddots & \\ 0 & a'_{n2} & \dots & a'_{nn} \end{array} \right]$$

- Find the row from among 2 to n having the first non-zero coefficient when reading from left to right.
- Move this row to the second position.

$$\left[\begin{array}{cccc} a_{11} & a_{12} & \dots & a_{1n} \\ 0 & 0 & \dots & a'_{2t_2} \dots a'_{2n} \\ 0 & & \vdots & \\ 0 & & & \\ 0 & 0 & & a'_{rt_r} \dots a'_{rn} \end{array} \right]$$

all 0's

Perform Gaussian elimination on the submatrix
(the other 0's will be unchanged)

In the end: the matrix looks like this:

$$\left[\begin{array}{ccccc} a'_{11} & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & a'_{2t_2} & \dots \\ 0 & 0 & 0 & 0 & \dots \\ \vdots & & & & \\ 0 & \vdots & \dots & \dots & 0 & a'_{rt_r} & \dots \end{array} \right]$$

with a'_{it_i} being the first non-zero coefficient in row i .

By exchanging columns, appropriately renaming the variables and scaling the rows, we can assume the system to look like:

$$\left[\begin{array}{cccc|c} 1 & & & & 0 \\ 0 & 1 & & & \\ 0 & 0 & 1 & & \\ \vdots & & 0 & \ddots & \\ 0 & 0 & 0 & 0 & 1 & \cdots \end{array} \right] \left[\begin{array}{c} x_1 \\ x_2 \\ \vdots \\ x_n \end{array} \right] = \left[\begin{array}{c} 0 \\ \vdots \\ 0 \end{array} \right]$$

Remark: Since we started with linearly independent rows, none of the rows will become fully 0 during the elimination.

Remark 2: x is a solution to $Ax=0$ iff it satisfies the modified system shown above.

- Once values for x_{r+1}, \dots, x_n are fixed, the values of x_1, \dots, x_r are determined.
- This suggests that the required dimension is $n-r$.
- We will now exhibit a basis to this end.

$$\begin{bmatrix} 1 & \dots & \dots \\ 0 & 1 & \dots \\ 0 & 0 & 1 \\ \vdots & & \vdots \\ 0 & 0 & 0 & 0 & 1 & \dots \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

Define: $U_{r+1} = \begin{bmatrix} \alpha_{1,r+1} \\ \alpha_{2,r+1} \\ \vdots \\ \alpha_{r,r+1} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ where $\alpha_{1,r+1}, \dots, \alpha_{r,r+1}$ are obtained from above

$U_n = \begin{bmatrix} \alpha_{1,n} \\ \vdots \\ \alpha_{r,n} \\ 0 \\ \vdots \\ 1 \end{bmatrix}$ where $\alpha_{1,n}, \dots, \alpha_{r,n}$ are obtained from above system.

Claim 1: U_{r+1}, \dots, U_n are linearly independent

Claim 2: $AU_{r+1} = AU_{r+2} = \dots = AU_n = 0$

Claim 3: Every solution to $Ax=0$ can be written as a linear combination of U_{r+1}, \dots, U_n .

$$\begin{bmatrix} 1 & \dots & \dots & \dots \\ 0 & 1 & \dots & \dots \\ 0 & 0 & 1 & \dots \\ \vdots & \ddots & & \dots \\ 0 & 0 & 0 & 0 & 1 & \dots \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix}$$

Define: $U_{r+1} = \begin{bmatrix} \alpha_{1,n} \\ \alpha_{2,n} \\ \vdots \\ \alpha_{r,n} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ where $\alpha_{1,n}, \dots, \alpha_{r,n}$ are obtained from above

$U_n = \begin{bmatrix} \alpha_{1,n} \\ \vdots \\ \alpha_{r,n} \\ 0 \\ \vdots \\ 1 \end{bmatrix}$ where $\alpha_{1,n}, \dots, \alpha_{r,n}$ are obtained from above system.

Claim 3: Every solution to $Ax=0$ is a combination of U_{r+1}, \dots, U_n .

Let $x = [x_1, x_2, \dots, x_n]^T$ be a solution to $Ax=0$.

- Define $x' = x_{r+1} U_{r+1} + x_{r+2} U_{r+2} + \dots + x_n U_n$

We will show that $x = x'$.

- Define $x'' = x - x'$. To show: $x'' = \vec{0}$

Observation 1: $x''_{r+1} = x''_{r+2} = \dots = x''_n = 0$

Observation 2: $Ax'' = Ax - Ax' = 0 - \sum_{i=r+1}^n x_i A U_i = 0$

Hence x'' is a solution to $Ax=0$.

$$\begin{bmatrix} 1 & \dots & \dots & \dots \\ 0 & 1 & \dots & \dots \\ 0 & 0 & 1 & \dots \\ \vdots & & 0 & \dots \\ 0 & 0 & 0 & 0 & 1 & \dots \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \end{bmatrix} \quad \xrightarrow{\text{(*)}}$$

Define: $U_{t+1} = \begin{bmatrix} \alpha_{1,n} \\ \alpha_{2,n} \\ \vdots \\ \alpha_{t,n} \\ 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ where $\alpha_{1,n}, \dots, \alpha_{t,n}$ are obtained from above

$U_n = \begin{bmatrix} \alpha_{1,n} \\ \vdots \\ \alpha_{t,n} \\ 1 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$ where $\alpha_{1,n}, \dots, \alpha_{t,n}$ are obtained from above system.

Observation 1: $x''_{r+1} = x''_{r+2} = \dots = x''_n = 0$

Observation 2: $Ax'' = Ax - Ax' = 0 - \sum_{i=r+1}^n x_i A_{i1} = 0$

Hence x'' is a solution to $Ax = 0$.

Since x'' is a solution to $Ax=0$, it satisfies the system

(*) shown above.

Moreover $x''_{r+1} = \dots = x''_n = 0$.

From (*) this means $x''_1 = x''_2 = \dots = x''_r = 0$

This proves $x'' = x - x' = 0$

Hence the claim that x is a linear combination of U_{t+1}, \dots, U_n

Claims 1, 2, 3 show that U_{r+1}, \dots, U_n forms

a basis to $\{x : Ax=0\}$.

Hence $\dim \{x : Ax=0\} = n-r$.

Summary of Part 2:

- Restrict $Ax=0$ to the first r linearly independent rows.
- Perform Gaussian elimination to get a simplified system where x_1, \dots, x_r get determined by values of x_{r+1}, \dots, x_n .
- Exhibit a basis using this observation.

Summary:

$$\dim \{x \mid Ax = 0\} = n - r \quad (r: \text{no. of linearly independent columns of } A)$$

$$= n - r \quad (r: \text{no. of linearly independent rows of } A)$$

Corollary:

dimension of column space = dimension of row space of A

column rank = row rank

- This quantity is called rank of the matrix.