

4.6 Rank of a matrix (cont'd)

ex: Find bases for the row space and the
#8) column space of

$$A = \begin{bmatrix} 2 & 5 \\ -2 & -5 \\ -6 & -15 \end{bmatrix} \quad \xrightarrow{\text{rref}}$$

$$B = \begin{bmatrix} 1 & 2.5 \\ 0 & 0 \\ 0 & 0 \end{bmatrix}$$

Basis for the row space of A = $\{(1, 2.5)\} = \{(1, \frac{5}{2})\}$
So the row rank of A is 1.

Basis for the column space of A (method 2)
= $\{(2, -2, -6)\}$

Defn: The column rank of a matrix $A = \dim(\text{column space of } A)$.
= number of leading 1s in the
reduced row echelon form of A.

Theorem 4.15: The row rank of A = the column rank of A.
That is, the dimension of the row space of A
= the dimension of the column space of A.

proof: Both numbers equal the number of leading 1s in
the reduced row echelon form of A.

Defn: The rank of a matrix is the dimension of the row space (or column space) of A .

ex 24) Find a basis for the column space of A ,

where $A = \begin{bmatrix} 4 & 20 & 31 \\ 6 & -5 & -6 \\ 2 & -11 & -16 \end{bmatrix} \xrightarrow{\text{ref}}$

$$B = \begin{bmatrix} 1 & 0 & \frac{1}{4} \\ 0 & 1 & \frac{3}{2} \\ 0 & 0 & 0 \end{bmatrix}$$

[Method 2] Basis for the column space of $A = \{ (4, 6, 2), (20, -5, -11) \}$

[Another correct answer by method 1]

$$A^T = \begin{bmatrix} 4 & 6 & 2 \\ 20 & -5 & -11 \\ 31 & -6 & -16 \end{bmatrix} \xrightarrow{\text{ref}} \begin{bmatrix} 1 & 0 & -\frac{2}{5} \\ 0 & 1 & \frac{3}{5} \\ 0 & 0 & 0 \end{bmatrix}$$

Basis for column space of $A = \{ (1, 0, -\frac{2}{5}), (0, 1, \frac{3}{5}) \}$

Remark: In effect the matrix

$$C = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -\frac{2}{5} & \frac{3}{5} & 0 \end{bmatrix}$$

is column equivalent to A so column space of C = column space of A .

Take the nonzero columns of C as the basis for the column space of A .

This is what method 1 is.

Defn : If A is an $m \times n$ matrix,
the set of all solutions of the
homogeneous system of equations

$$\begin{matrix} m \text{ equations} \rightarrow \\ n \text{ variables} \end{matrix} \quad \begin{matrix} m \times n & n \times 1 & & m \times 1 \\ A \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \end{matrix} \left. \vphantom{\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}} \right\} m \text{ times}$$

is called the nullspace (or solution space) of A .

^{more} Compact notation: $A \vec{x} = \vec{0}$

We denote the nullspace of A as $N(A)$.

Thm 4.16: The nullspace of A is a subspace
of $\mathbb{R}^n$. We call $\dim(N(A)) = \underline{\text{nullity of } A}$.

proof: (easy). The nullspace of A is nonempty since $\vec{0} \in N(A)$:

$$A \vec{0} = \vec{0}$$

$N(A)$ is closed under addition: Suppose that $\vec{x}_1 \in N(A)$ and $\vec{x}_2 \in N(A)$.

That is, $A \vec{x}_1 = \vec{0}$ and $A \vec{x}_2 = \vec{0}$. Then

$$A(\vec{x}_1 + \vec{x}_2) = A \vec{x}_1 + A \vec{x}_2 = \vec{0} + \vec{0} = \vec{0}$$

so that $\vec{x}_1 + \vec{x}_2 \in N(A)$.

$N(A)$ is closed under scalar multiplication: Let $\vec{x} \in N(A)$ and

let c be a scalar. Then $A(c\vec{x}) = cA\vec{x} = c\vec{0} = \vec{0}$

so that $c\vec{x} \in N(A)$.

Since the nullspace of A is a nonempty subset of $\mathbb{R}^n$
which is closed under addition and scalar multiplication,
it is a subspace of $\mathbb{R}^n$.