

# MA 265, Lesson 18

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## Rank (4.9)

Recall:

Thm 4.18 The row rank and column rank of a matrix are equal.

Thm 4.19 If  $A$  is an  $m \times n$  matrix, then  $\text{rank } A + \text{nullity } A = n$ .

(You will be asked to verify these theorems.)

Now, for an  $n \times n$  matrix, rank is related to singularity (and other related things.)

Theorem 4.20 If  $A$  is an  $n \times n$  matrix, then  $\text{rank } A = n$  if and only if  $A$  is row equivalent to  $I_n$ .

Proof. Transform  $A$  into RREF. If  $\text{rank } A = n$ , then  $A$  has  $n$  nonzero rows. Since  $A$  has  $n$  rows in total,  $\text{RREF}(A) = I_n$ . Thus,  $A$  is row equivalent to  $I_n$ .

If  $A$  is row equivalent to  $I_n$ , then  $\text{RREF}(A) = I_n$ . Since  $\text{rank } A$  is the number of nonzero rows of  $\text{RREF}(A) = I_n$ ,  $\text{rank } A = n$ .  $\square$

Being row equivalent to  $I_n$  gives us a lot of information.

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Corollary 4.7 An  $n \times n$  matrix  $A$  is nonsingular if and only if  $\text{rank } A = n$ .

Corollary 4.8 If  $A$  is an  $n \times n$  matrix, then  $\text{rank } A = n$  if and only if  $\det(A) \neq 0$ .

Corollary 4.9 Let  $A$  be  $n \times n$ . The homogeneous system  $A\vec{x} = \vec{0}$  has a nontrivial solution if and only if  $\text{rank } A < n$ .

Corollary 4.10 Let  $A$  be an  $n \times n$  matrix. The linear system  $A\vec{x} = \vec{b}$  has a unique solution for every  $n \times 1$  matrix  $\vec{b}$  if and only if  $\text{rank } A = n$ .

These are proven using Thm 4.20 and already established equivalences.

Ex. 1. Show that  $\left\{ \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ -2 \\ 1 \end{bmatrix} \right\}$  spans  $\mathbb{R}^3$ .

Form the matrix:

$$\begin{bmatrix} 1 & 0 & 2 \\ 2 & 1 & 0 \\ 0 & 1 & -2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & -4 \\ 0 & 1 & -2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & -4 \\ 0 & 0 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & -4 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \text{rank } A = 3, \text{ so } \det(A) \neq 0, \text{ so}$$

columns are linearly independent,

so since  $\dim \mathbb{R}^3 = 3$ , also

Spanning.

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Theorem 4.2 The linear system  $A\vec{x} = \vec{b}$  has a solution if and only if  $\text{rank } A = \text{rank } [A | \vec{b}]$ .

Proof. Suppose  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$  is a solution to the system  $A\vec{x} = \vec{b}$ . Let  $\vec{x}_1, \dots, \vec{x}_n$  be the columns of  $A$ . Then  $\vec{b} = a_1\vec{x}_1 + \dots + a_n\vec{x}_n$ , so  $\vec{b}$  is a linear combination of the columns of  $A$ . As such, converting  $[A | \vec{b}]$  into RCEF, we get  $[C | \vec{0}]$  where  $C = \text{RCEF}(A)$ . Thus,  $\text{rank } A = \text{rank } C = \text{column rank } [C | \vec{0}] = \text{rank } [C | \vec{0}] = \text{rank } [A | \vec{b}]$ .

Reversing the argument gives the other direction.  $\square$

Ex 2. Use Thm 4.2) to show

$$\left[ \begin{array}{cccc} 1 & -2 & -3 & 4 \\ 4 & -1 & -5 & 6 \\ 2 & 3 & 1 & -2 \end{array} \right] \left[ \begin{array}{c} x_1 \\ x_2 \\ x_3 \\ x_4 \end{array} \right] = \left[ \begin{array}{c} 1 \\ 2 \\ 2 \end{array} \right] \text{ is inconsistent.}$$

$$\left[ \begin{array}{cccc|c} 1 & -2 & -3 & 4 & 1 \\ 4 & -1 & -5 & 6 & 2 \\ 2 & 3 & 1 & -2 & 2 \end{array} \right] \rightarrow \left[ \begin{array}{cccc|c} 1 & 0 & -13 & 8/7 & 0 \\ 0 & 1 & 1 & -10/7 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{array} \right] \text{ rank 3} \neq$$

$$\left[ \begin{array}{cccc|c} 1 & -2 & -3 & 4 & 1 \\ 4 & -1 & -5 & 6 & 0 \\ 2 & 3 & 1 & -2 & 0 \end{array} \right] \rightarrow \left[ \begin{array}{cccc|c} 1 & 0 & -1 & 8/7 & 1 \\ 0 & 1 & 1 & -10/7 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{array} \right] \text{ rank 2}$$

Recall  $\text{rank } A \leq \min\{m, n\}$  when  $A$  is  $m \times n$ .  
 (since  $\text{row rank } A \leq m$  and  $\text{column rank } A \leq n$ )

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Some things you can show with rank:

Ex 3. Let  $A$  be  $m \times n$ . Show that  $\vec{A}\vec{x} = \vec{b}$  has a solution for every  $m \times 1$  matrix  $\vec{b}$  if and only if  $\text{rank } A = m$ .

Proof. If  $\vec{A}\vec{x} = \vec{b}$  has a solution for every  $\vec{b}$ , by Thm 4.20,  $[A | \vec{b}]$  has the same rank as  $A$  for every  $\vec{b}$ . Thus, every vector in  $\mathbb{R}^n$  is a linear combo of the columns of  $A$ , so the columns of  $A$  span  $\mathbb{R}^n$ . This is not possible if  $\text{rank } A \neq m$ . Conversely, if  $\text{rank } A = m$ , then every row is nonzero for  $A$  in RREF, so we get a solution.

Ex 3. Let  $A$  be  $m \times n$  with  $m \neq n$ . Show that either the rows or columns of  $A$  are linearly dependent.

Proof.  $\text{rank } A \leq \min\{m, n\}$ .

case 1. Suppose  $m < n$ . Then  $\text{rank } A \leq m < n$ . Putting  $A$  in RCEF, we get  $\leq m$  nonzero rows, thus  $\dim \text{column space } A \leq m < n$ , so the columns of  $A$  are linearly dependent (more than in a basis).

case 2. Suppose  $n < m$ . Then  $\text{rank } A \leq n < m$ .

Putting  $A$  in RREF, we get  $\leq n$  nonzero rows, thus  $\dim \text{row space } A \leq n < m$ , so the rows of  $A$  are linearly dependent.  $\square$

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Ex 4. Show  $\text{rank } A = \text{rank } A^T$ . Does  $\text{nullity } A = \text{nullity } A^T$ ?

Proof.  $\text{rank } A = \text{row rank } A = \text{column rank } A^T = \text{rank } A^T$ .  $\square$

Now,  $\text{nullity } A \neq \text{nullity } A^T$  in general.

If  $A$  is  $4 \times 7$  and  $\text{rank } A = 4$ , then  $\text{nullity } A = 3$ .

Now,  $A^T$  is  $7 \times 4$  and  $\text{rank } A^T = 4$ , so  $\text{nullity } A^T = 0$ .

Ex 5. Suppose  $A\vec{x} = \vec{b}$  is consistent ( $A$  is  $m \times n$ ).

Prove the solution is unique if and only if  $\text{rank } A = n$ .

Proof. Suppose the solution is unique. Let  $\vec{x}_p$  be the particular solution. Since  $\vec{x}_p + \vec{x}_h$  is a solution for any homogeneous solution  $\vec{x}_h$ ,  $\vec{x}_h = \vec{0}$  is the only homogeneous solution. Thus  $\text{nullity } A = 0$ .

By Rank-Nullity,  $\text{rank } A = n - 0 = n$ .

Suppose  $\text{rank } A = n$ . Then by rank-nullity,  $\text{nullity } A = 0$ . So  $\vec{0}$  is the only homogeneous solution.

Since all solutions to  $A\vec{x} = \vec{b}$  are of the form

$\vec{x}_p + \vec{x}_h$ ,  $\vec{x}_p + \vec{0} = \vec{x}_p$  is the only solution, so it is unique.  $\square$