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A First Course in LINEAR ALGEBRA

Lecture Notes
by Karen Seyffarth

 $\mathbb{R}^n$ : Row, Column, and Null Space

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# A First Course in Linear Algebra

#### Lecture Notes

Current Lecture Notes Revision: Version 2017 - Revision A

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- Ilijas Farah, York University
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- Asia Weiss, York University

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# **Definitions**

Let A be an  $m \times n$  matrix. The **column space** of A, written col(A), is the span of the columns. The **row space** of A, written row(A), is the span of the rows.

### **Notation**

Let A and B be  $m \times n$  matrices. We write  $A \to B$  if B can be obtained from A by a sequence of elementary row operations. Note that  $A \to B$  if and only if  $B \to A$ .





#### Lemma

Let A and B be  $m \times n$  matrices such that A can be carried to B by elementary row [column] operations. Then row(A) = row(B) [col(A) = col(B)].

### Proof.

We will prove that the above is true for row operations, which can be easily applied to column operations. Let  $\vec{r_1}, \vec{r_2}, \dots, \vec{r_m}$  denote the rows of A.

- If B is obtained from A by a interchanging two rows of A, then A and B have exactly the same rows, so row(B) = row(A).
- Suppose  $p \neq 0$ , and for some j,  $1 \leq j \leq m$ , B is obtained from A by multiplying row j by p. Then

$$row(B) = span\{\vec{r}_1, \dots, p\vec{r}_j, \dots, \vec{r}_m\}.$$

Since  $\{\vec{r_1},\ldots,p\vec{r_j},\ldots,\vec{r_m}\}\subseteq \operatorname{row}(A)$ , it follows that  $\operatorname{row}(B)\subseteq \operatorname{row}(A)$ . Conversely, since  $\{\vec{r_1},\ldots,\vec{r_m}\}\subseteq \operatorname{row}(B)$ , it follows that  $\operatorname{row}(A)\subseteq \operatorname{row}(B)$ . Therefore,  $\operatorname{row}(B)=\operatorname{row}(A)$ .





# Proof (continued).

• Suppose  $p \neq 0$ , and suppose that for some i and j,  $1 \leq i, j \leq m$ , B is obtained from A by adding p time row j to row i. Without loss of generality, we may assume i < j.

Then

$$row(B) = span\{\vec{r}_1, \ldots, \vec{r}_{i-1}, \vec{r}_i + p\vec{r}_j, \ldots, \vec{r}_j, \ldots, \vec{r}_m\}.$$

Since

$$\{\vec{r}_1,\ldots,\vec{r}_{i-1},\vec{r}_i+p\vec{r}_j,\ldots,\vec{r}_m\}\subseteq \text{row}(A),$$

it follows that  $row(B) \subseteq row(A)$ .

Conversely, since

$$\{\vec{r}_1,\ldots,\vec{r}_m\}\subseteq \text{row}(B),$$

it follows that  $row(A) \subseteq row(B)$ . Therefore, row(B) = row(A).





# Corollary

Let A be an  $m \times n$  matrix, U an invertible  $m \times m$  matrix, and V an invertible  $n \times n$  matrix. Then row(UA) = row(A) and col(AV) = col(A).

# Proof.

Since U is invertible, U is a product of elementary matrices, implying that  $A \to UA$  by a sequence of elementary row operations. Then row(UA) = row(A).

Now consider AV:  $col(AV) = row((AV)^T) = row(V^TA^T)$  and  $V^T$  is invertible (a matrix is invertible if and only if its transpose is invertible). It follows from the first part of this Corollary that

$$row(V^TA^T) = row(A^T).$$

But  $row(A^T) = col(A)$ , and therefore col(AV) = col(A).





#### Lemma

Let A be an  $m \times n$  matrix and let R be its row-echelon form . Then the nonzero rows of R form a basis of row(R), and consequently of row(A).

A variation of this lemma provides a basis of col(A). Suppose A is row reduced to its row-echelon form R. Identify the pivot columns of R (columns which have leading ones), and take the corresponding columns of A. It turns out that this forms a basis of col(A).

# Example

Let

$$R = \left[ \begin{array}{ccccccc} 1 & 2 & 2 & -2 & 0 & 0 \\ 0 & 1 & 3 & 1 & -1 & 2 \\ 0 & 0 & 0 & 1 & -2 & 5 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{array} \right].$$

- **①** Since the nonzero rows of R are linearly independent, they form a basis of row(R).
- ② Let  $B = \{\vec{e_1}, \vec{e_2}, \vec{e_3}, \vec{e_4}\} \subseteq \mathbb{R}^5$ . Then B is linearly independent and spans  $\operatorname{col}(R)$ , and thus is a basis of  $\operatorname{col}(R)$ . Now let X denote the set of columns of R that contain the leading ones. Then X is a linearly independent subset of  $\operatorname{col}(R)$  with  $A = \operatorname{dim}(\operatorname{col}(R))$  vectors. It follows that X spans  $\operatorname{col}(R)$ , and therefore is a basis of  $\operatorname{col}(R)$ .



# **Definition**

Previously, we defined rank(A) to be the number of leading entries in the row-echelon form of A. Using an understanding of dimension and row space, we can now define rank as follows:

$$rank(A) = dim(row(A))$$

# Theorem (Rank Theorem)

Let A be an  $m \times n$  matrix. Then  $\dim(\operatorname{col}(A))$ , the dimension of the column space, is equal to the dimension of the row space,  $\dim(\operatorname{row}(A))$ .



Find the rank of the following matrix and describe the column and row spaces.

$$A = \left[ \begin{array}{rrrrr} 1 & 2 & 1 & 3 & 2 \\ 1 & 3 & 6 & 0 & 2 \\ 3 & 7 & 8 & 6 & 6 \end{array} \right]$$

### Solution

The reduced row-echelon form of A is

$$\left[\begin{array}{ccccc}
1 & 0 & -9 & 9 & 2 \\
0 & 1 & 5 & -3 & 0 \\
0 & 0 & 0 & 0 & 0
\end{array}\right]$$

Therefore, the rank is 2.

# Solution continued

Notice that the first two columns of R are pivot columns. We find the corresponding columns of A to create a basis for col(A):

$$\left\{ \left[\begin{array}{c} 1\\1\\3 \end{array}\right], \left[\begin{array}{c} 2\\3\\7 \end{array}\right] \right\}$$

We know that the nonzero rows of R create a basis of row(A). For the above matrix, the row space equals

$$row(A) = span \{ \begin{bmatrix} 1 & 0 & -9 & 9 & 2 \end{bmatrix}, \begin{bmatrix} 0 & 1 & 5 & -3 & 0 \end{bmatrix} \}$$



# Corollary

Let A be a matrix. Then the following are true:

- 2 For A of size  $m \times n$ ,  $rank(A) \leq m$  and  $rank(A) \leq n$ .
- **3** For A of size  $n \times n$ , A is invertible if and only if rank(A) = n.
- For invertible matrices B and C of appropriate size, rank(A) = rank(BA) = rank(AC).



## **Theorem**

Let A be an  $m \times n$  matrix. The following are equivalent.

- $\bigcirc$  rank(A) = n.
- 2  $row(A) = \mathbb{R}^n$ , i.e., the rows of A span  $\mathbb{R}^n$ .
- **3** The columns of A are independent in  $\mathbb{R}^m$ .
- 4 The  $n \times n$  matrix  $A^T A$  is invertible.
- **5** There exists an  $n \times m$  matrix C so that  $CA = I_n$ .
- **6** If  $A\vec{x} = \vec{0}_m$  for some  $\vec{x} \in \mathbb{R}^n$ , then  $\vec{x} = \vec{0}_n$ .

### Theorem

Let A be an  $m \times n$  matrix. The following are equivalent.

- $\bigcirc$  rank(A) = m.
- 2  $\operatorname{col}(A) = \mathbb{R}^m$ , i.e., the columns of A span  $\mathbb{R}^m$ .
- **3** The rows of A are independent in  $\mathbb{R}^n$ .
- 4 The  $m \times m$  matrix  $AA^T$  is invertible.
- **5** There exists an  $n \times m$  matrix C so that  $AC = I_m$ .
- **6** The system  $A\vec{x} = \vec{b}$  is consistent for every  $\vec{b} \in \mathbb{R}^m$ .

# **Definitions**

Let A be an  $m \times n$  matrix. The null space of A is defined as

$$\mathsf{null}(A) = \{ \vec{x} \in \mathbb{R}^n \mid A\vec{x} = \vec{0}_m \},\$$

and the image space of A is defined as

$$\mathsf{im}(A) = \{A\vec{x} \mid \vec{x} \in \mathbb{R}^n\}.$$

Note. Since A is  $m \times n$  and  $\vec{x} \in \mathbb{R}^n$ ,  $A\vec{x} \in \mathbb{R}^m$ , so  $\text{im}(A) \subseteq \mathbb{R}^m$  while  $\text{null}(A) \subseteq \mathbb{R}^n$ .





Prove that if A is an  $m \times n$  matrix, then null(A) is a subspace of  $\mathbb{R}^n$ .

# Proof.

- Since  $A\vec{0}_n = \vec{0}_m$ ,  $\vec{0}_n \in \text{null}(A)$ .
- Let  $\vec{x}, \vec{y} \in \text{null}(A)$ . Then  $A\vec{x} = \vec{0}_m$  and  $A\vec{y} = \vec{0}_m$ , so  $A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y} = \vec{0}_m + \vec{0}_m = \vec{0}_m,$  and thus  $\vec{x} + \vec{y} \in \text{null}(A)$ .
- Let  $\vec{x} \in \text{null}(A)$  and  $k \in \mathbb{R}$ . Then  $A\vec{x} = \vec{0}_m$ , so  $A(k\vec{x}) = k(A\vec{x}) = k\vec{0}_m = \vec{0}_m,$  and thus  $k\vec{x} \in \text{null}(A)$ .

Therefore, null(A) is a subspace of  $\mathbb{R}^n$ .





Prove that if A is an  $m \times n$  matrix, then im(A) is a subspace of  $\mathbb{R}^m$ .

# Proof.

- Since  $\vec{0}_n \in \mathbb{R}^n$  and  $A\vec{0}_n = \vec{0}_m$ ,  $\vec{0}_m \in \text{im}(A)$ .
- Let  $\vec{x}, \vec{y} \in \text{im}(A)$ . Then  $\vec{x} = A\vec{u}$  and  $\vec{y} = A\vec{v}$  for some  $\vec{u}, \vec{v} \in \mathbb{R}^n$ , so  $\vec{x} + \vec{y} = A\vec{u} + A\vec{v} = A(\vec{u} + \vec{v})$ . Since  $\vec{u} + \vec{v} \in \mathbb{R}^n$ , it follows that  $\vec{x} + \vec{y} \in \text{im}(A)$ .
- Let  $\vec{x} \in \text{im}(A)$  and  $k \in \mathbb{R}$ . Then  $\vec{x} = A\vec{u}$  for some  $\vec{u} \in \mathbb{R}^n$ , and thus  $k\vec{x} = k(A\vec{u}) = A(k\vec{u})$ . Since  $k\vec{u} \in \mathbb{R}^n$ , it follows that  $k\vec{x} \in \text{im}(A)$ .

Therefore, im(A) is a subspace of  $\mathbb{R}^m$ .





### Theorem

Let A be an  $m \times n$  matrix such that  $\operatorname{rank}(A) = r$ . Then the system  $A\vec{x} = \vec{0}_m$  has n - r basic solutions, providing a basis of  $\operatorname{null}(A)$  with  $\operatorname{dim}(\operatorname{null}(A)) = n - r$ .

### Outline of Proof.

- We have already seen that null(A) is spanned by any set of basic solutions to  $A\vec{x} = \vec{0}_m$ , so it is enough to prove that  $\dim(\text{null}(A)) = n r$ .
- Suppose  $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$  is a basis of null(A) (show k = n r).
- Extend  $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k\}$  to a basis  $\{\vec{x}_1, \vec{x}_2, \dots, \vec{x}_k, \dots \vec{x}_n\}$  of  $\mathbb{R}^n$ .
- Consider the set  $\{A\vec{x}_1, A\vec{x}_2, \dots, A\vec{x}_k, \dots A\vec{x}_n\}$  (a subset of  $\mathbb{R}^m$ ).
- Then  $A\vec{x_j} = \vec{0}_m$  for  $1 \le j \le k$  since  $\vec{x_1}, \dots, \vec{x_k} \in \text{null}(A)$ .
- To complete the proof, show  $S = \{A\vec{x}_{k+1}, \dots A\vec{x}_n\}$  is a basis of im(A), by showing S is independent, and that S spans im(A).
- Since im(A) = col(A), dim(im(A)) = r, implying n k = r. Therefore, k = n r.





# Example

Let

$$A = \left[ \begin{array}{rrr} 1 & 2 & 1 \\ 0 & -1 & 1 \\ 2 & 3 & 3 \end{array} \right]$$

Find null(A) and im(A).

# Solution

T find null(A), we need solutions to:

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

The solution is given by

$$egin{array}{c|c} t & -3 & \ 1 & \ 1 & \ \end{array} : t \in \mathbb{R}$$

Therefore,

$$\mathsf{null}(A) = \mathsf{span} \left\{ \left[ \begin{array}{c} -3 \\ 1 \\ 1 \end{array} \right] \right\}$$

### Solution continued

Finally im (A) is just  $\{A\vec{x}: \vec{x} \in \mathbb{R}^n\}$ , that is im (A) = col(A).

Notice from the above calculation that that the first two columns of the reduced row-echelon form—are pivot columns. Therefore

$$\operatorname{im}(A) = \operatorname{col}(A) = \operatorname{span}\left\{ \begin{bmatrix} 1\\0\\2 \end{bmatrix}, \begin{bmatrix} 2\\-1\\3 \end{bmatrix} \right\}$$





Let A be a 5  $\times$  6 matrix. Can the columns of A be independent? Can the rows of A be independent? Justify your answer.

# Solution

The rank of the matrix is at most five; since there are six columns, the columns can not be independent. However, the rows could be independent: take a  $5 \times 6$  matrix whose first five columns are the columns of the  $5 \times 5$  identity matrix.





Let A be an  $5 \times 9$  matrix. Is it possible that  $\dim(\text{null}(A)) = 3$ ? Justify your answer.

## Solution

As a consequence of the Rank Theorem, we have rank  $(A) \leq 5$ , so  $\dim(\operatorname{im}(A)) \leq 5$ . Since dim(null(A)) = 9 - dim(im(A)), it follows that

$$\dim(\text{null}(A)) \ge 9 - 5 = 4.$$

Therefore, it is not possible that dim(null(A)) = 3.



