

## Rank, Row-Reduced Form, and Solutions to $A \mathbf{x} = \mathbf{b}$

### Example

1. Consider the  $3 \times 4$  matrix  $A$  given by

$$A = \begin{pmatrix} 1 & -2 & 3 & 9 \\ -1 & 3 & 0 & -4 \\ 2 & -5 & 5 & 17 \end{pmatrix}.$$

Using the three elementary row operations we may rewrite  $A$  in an *echelon form* as

$$U = \begin{pmatrix} 1 & -2 & 3 & 9 \\ 0 & 1 & 3 & 5 \\ 0 & 0 & 2 & 4 \end{pmatrix}$$

or, continuing with additional row operations, in the *reduced row-echelon form*

$$R = \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & -1 \\ 0 & 0 & 1 & 2 \end{pmatrix}.$$

From the above, the homogeneous system  $A \mathbf{x} = \mathbf{0}$  has a solution that can be read as

$$\begin{aligned} x &+ w = 0 \\ y &- w = 0 \\ z &+ 2w = 0 \end{aligned}$$

or in vector form as

$$\begin{pmatrix} x \\ y \\ z \\ w \end{pmatrix} = \begin{pmatrix} -w \\ w \\ -2w \\ w \end{pmatrix} = w \begin{pmatrix} -1 \\ 1 \\ -2 \\ 1 \end{pmatrix}.$$

In the above, recall that  $w$  is a *free variable* while  $x$ ,  $y$ , and  $z$  are the three *pivot variables*. The solution of the homogeneous system  $A \mathbf{x} = \mathbf{0}$  (i.e., the *null space* of  $A$ ) consists of all scalar

multiples of the vector

$$\begin{pmatrix} -1 \\ 1 \\ -2 \\ 1 \end{pmatrix}$$

and, hence, has *dimension* 1 (the number of free variables).

2. Consider the  $4 \times 5$  matrix given by

$$\begin{pmatrix} 1 & 0 & -2 & 1 & 0 \\ 0 & -1 & -3 & 1 & 3 \\ -2 & -1 & 1 & -1 & 3 \\ 0 & 3 & 9 & 0 & -12 \end{pmatrix}.$$

The associated *row reduced echelon form* is given by

$$\begin{pmatrix} 1 & 0 & -2 & 0 & 1 \\ 0 & 1 & 3 & 0 & -4 \\ 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

For the homogeneous system

$$\begin{pmatrix} 1 & 0 & -2 & 1 & 0 \\ 0 & -1 & -3 & 1 & 3 \\ -2 & -1 & 1 & -1 & 3 \\ 0 & 3 & 9 & 0 & -12 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} = \mathbf{0}_{5 \times 1}$$

or, equivalently,

$$\begin{pmatrix} 1 & 0 & -2 & 0 & 1 \\ 0 & 1 & 3 & 0 & -4 \\ 0 & 0 & 0 & 1 & -1 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \end{pmatrix} = \mathbf{0}_{5 \times 1}$$

we see that  $x_1$ ,  $x_2$ , &  $x_4$  are the three *pivot variables* while  $x_3$  &  $x_5$  are the two *free variables*.

Here the *null space* of the given coefficient matrix is

$$\left\langle \begin{pmatrix} 2 \\ -3 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 4 \\ 0 \\ 1 \\ 1 \end{pmatrix} \right\rangle$$

and has *dimension 2* (the number of free variables).

### **Definition**

Suppose  $A$  is an  $m \times n$  matrix.

1. We call the number of free variables of  $Ax = b$  the ***nullity of A*** and we denote it by ***Nullity(A)***.
2. We call the number of pivots of  $A$  the ***rank of A*** and we denoted it by ***Rank(A)***.

Procedure for computing the rank of a matrix  $A$ :

1. Use elementary row operations to transform  $A$  to a matrix  $R$  in reduced row echelon form.
2. ***Rank(A)*** is the number of nonzero rows in  $R$ . (Why?)

### Example

1. For  $A = \begin{pmatrix} 1 & -2 & 3 & 9 \\ -1 & 3 & 0 & -4 \\ 2 & -5 & 5 & 17 \end{pmatrix}$  we see that  $\text{Rank}(A) = 3$  and  $\text{Nullity}(A) = 1$ .

2. The *nullity* of  $\begin{pmatrix} 1 & 0 & -2 & 1 & 0 \\ 0 & -1 & -3 & 1 & 3 \\ -2 & -1 & 1 & -1 & 3 \\ 0 & 3 & 9 & 0 & -12 \end{pmatrix}$  is 2 while its *rank* is 3.

Observe that in the above two examples that

$$\text{Rank}(A) + \text{Nullity}(A) = (\text{number of columns in the coefficient matrix } A).$$

This is no accident as the  $\text{Rank}(A)$  counts the pivot variables, the  $\text{Nullity}(A)$  counts the free variables, and the number of columns corresponds to the total number of variables for the coefficient matrix  $A$ .

### Theorem

Suppose  $A$  is an  $m \times n$  matrix. Then

$$n = \text{Rank}(A) + \text{Nullity}(A).$$

### Example

Consider the matrix

$$B = \begin{pmatrix} 1 & 1 & 4 & 1 & 2 \\ 0 & 1 & 2 & 1 & 1 \\ 0 & 0 & 0 & 1 & 2 \\ 1 & -1 & 0 & 0 & 2 \\ 2 & 1 & 6 & 0 & 1 \end{pmatrix}.$$

The row reduced echelon form of the above is

$$\mathbf{B} = \begin{pmatrix} 1 & 0 & 2 & 0 & 1 \\ 0 & 1 & 2 & 0 & -1 \\ 0 & 0 & 0 & 1 & 2 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

Since the number of nonzero rows is 3 (that is, there are three pivots), we see that

$$\mathit{Rank}(\mathbf{B}) = 3$$

and that the

$$\mathit{Nullity}(\mathbf{B}) = 5 - 3 = 2.$$

### **Problem**

Is the above  $5 \times 5$  matrix  $\mathbf{B}$  invertible (a.k.a. nonsingular)? So?!??

### **Theorem**

Suppose that  $\mathbf{A}$  is an  $n \times n$  matrix.  $\mathbf{A}$  is an invertible matrix if and only if  $\mathit{Rank}(\mathbf{A}) = n$ .

Why?

### **Problem**

Find replacements for  $a, b, c, d$  so that the  $3 \times 3$  matrix

$$\begin{pmatrix} 1 & 4 & 3 \\ 5 & a & b \\ c & d & 6 \end{pmatrix}$$

has rank of 1.

We now consider the nonhomogenous linear system  $\mathbf{A} \mathbf{x} = \mathbf{b}$ . We first observe that the solutions to the associated homogenous system  $\mathbf{A} \mathbf{x} = \mathbf{0}$  form a vector space (called the *null space*).

### Problem

Can the same be said for the set of solutions to  $\mathbf{A} \mathbf{x} = \mathbf{b}$ ?

Fortunately, there is a relationship between the solution sets to  $\mathbf{A} \mathbf{x} = \mathbf{0}$  and  $\mathbf{A} \mathbf{x} = \mathbf{b}$ .

### Theorem

If  $\mathbf{x}_p$  is a particular solution of  $\mathbf{A} \mathbf{x} = \mathbf{b}$  (that is,  $\mathbf{A} \mathbf{x}_p = \mathbf{b}$ ), then every solution of  $\mathbf{A} \mathbf{x} = \mathbf{b}$  can be written as

$$\mathbf{x} = \mathbf{x}_h + \mathbf{x}_p$$

where  $\mathbf{x}_h$  represents a solution of  $\mathbf{A} \mathbf{x} = \mathbf{0}$ .

### Proof

Let  $\mathbf{x}$  be any solution of  $\mathbf{A} \mathbf{x} = \mathbf{b}$ . Then

$$\mathbf{A} (\mathbf{x} - \mathbf{x}_p) = \mathbf{A} \mathbf{x} - \mathbf{A} \mathbf{x}_p - \mathbf{b} - \mathbf{b} = \mathbf{0}.$$

Thus,  $\mathbf{x} - \mathbf{x}_p$  is a solution of  $\mathbf{A} \mathbf{x} = \mathbf{0}$ , denote this solution by  $\mathbf{x}_h$ . Then

$$\mathbf{x} = \mathbf{x}_h + \mathbf{x}_p$$

where  $\mathbf{x}_p$  is a particular solution of  $\mathbf{A} \mathbf{x} = \mathbf{b}$  and  $\mathbf{x}_h$  represents a solution of  $\mathbf{A} \mathbf{x} = \mathbf{0}$ . >

**Example**

Solve  $\mathbf{A} \mathbf{x} = \mathbf{b}$  where

$$\mathbf{A} = \begin{pmatrix} 1 & 2 & 0 & -1 & 1 \\ 1 & 3 & 1 & 1 & -1 \\ 2 & 5 & 1 & 0 & 0 \\ 3 & 6 & 0 & 0 & -6 \\ 1 & 5 & 3 & 5 & -5 \end{pmatrix}$$

and

$$\mathbf{b} = \begin{pmatrix} -10 \\ -9 \\ -19 \\ -27 \\ -7 \end{pmatrix}.$$

(Illustrate the above theorem in a particular setting.)

**Solution**

Using the elementary row operations (via technology!) on the augmented matrix  $(\mathbf{A} | \mathbf{b})$  we find that

$$(\mathbf{A} | \mathbf{b}) = \begin{pmatrix} 1 & 2 & 0 & -1 & 1 & -10 \\ 1 & 3 & 1 & 1 & -1 & -9 \\ 2 & 5 & 1 & 0 & 0 & -19 \\ 3 & 6 & 0 & 0 & -6 & -27 \\ 1 & 5 & 3 & 5 & -5 & -7 \end{pmatrix}$$

is row equivalent to the *reduced row echelon form matrix*

$$\begin{pmatrix} 1 & 0 & -2 & 0 & -10 & -7 \\ 0 & 1 & 1 & 0 & 4 & -1 \\ 0 & 0 & 0 & 1 & -3 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

We see that  $\text{Rank}(A) = 3$  (with free variables of  $x_3$  &  $x_5$ ). In vector form we may write the solution to  $A \mathbf{x} = \mathbf{b}$  as

$$\mathbf{x} = \begin{pmatrix} 2x_3 + 10x_5 - 7 \\ -x_3 - 4x_5 - 1 \\ x_3 \\ 3x_5 + 1 \\ x_5 \end{pmatrix} = x_3 \begin{pmatrix} 2 \\ -1 \\ 1 \\ 0 \\ 0 \end{pmatrix} + x_5 \begin{pmatrix} 10 \\ -4 \\ 0 \\ 3 \\ 1 \end{pmatrix} + \begin{pmatrix} -7 \\ -1 \\ 0 \\ 1 \\ 0 \end{pmatrix}.$$

The null space for  $A$  is given by  $\left\langle \begin{pmatrix} 2 \\ -1 \\ 1 \\ 0 \\ 0 \end{pmatrix}, \begin{pmatrix} 10 \\ -4 \\ 0 \\ 3 \\ 1 \end{pmatrix} \right\rangle$  and  $\mathbf{x}_p = \begin{pmatrix} -7 \\ -1 \\ 0 \\ 1 \\ 0 \end{pmatrix}$ .

### **Definition**

Suppose that  $A$  is an  $m \times n$  matrix. Then we define the *row space of  $A$*  as

$$R_A = \langle r_1, r_2, r_3, \dots, r_m \rangle$$

where  $r_k$  is the  $k^{\text{th}}$  row of  $A$  and the *column space of  $A$*  as

$$C_A = \langle c_1, c_2, c_3, \dots, c_n \rangle$$

where  $c_k$  is the  $k^{\text{th}}$  column of  $A$ .



### Theorem

If an  $m \times n$  matrix  $A$  is row equivalent to an  $m \times n$  matrix  $B$ , the row space of  $A$  is equal to the row space of  $B$ .

Note: The above says that the elementary row operations do not change the row space of a matrix. However, the elementary row operations may change the column space.

### Example

Since

$$A = \begin{pmatrix} 1 & 2 & 0 & -1 & 1 \\ 1 & 3 & 1 & 1 & -1 \\ 2 & 5 & 1 & 0 & 0 \\ 3 & 6 & 0 & 0 & -6 \\ 1 & 5 & 3 & 5 & -5 \end{pmatrix} \sim B = \begin{pmatrix} 1 & 0 & -2 & 0 & -10 \\ 0 & 1 & 1 & 0 & 4 \\ 0 & 0 & 0 & 1 & -3 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix},$$

the above theorem implies that  $R_A = R_B$ . In particular,

$$R_A = \langle (1 \ 0 \ -2 \ 0 \ 10), (0 \ 1 \ 1 \ 0 \ 4), (0 \ 0 \ 0 \ 1 \ -3) \rangle = R_B.$$

We note that only three row vectors (and not five!) were required to span  $R_A$ . Now, by forming

$$C = \text{Transpose}(\text{rref}(\text{Transpose}(A))) = (\text{rref}(A^T))^T$$

we find that

$$C = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ -2 & 3 & 0 & 0 & 0 \end{pmatrix}$$

and so it follows by the above theorem that

$$C_A = \left\langle \left( \begin{array}{c} 1 \\ 0 \\ 1 \\ 0 \\ -2 \end{array} \right), \left( \begin{array}{c} 0 \\ 1 \\ 1 \\ 0 \\ 3 \end{array} \right), \left( \begin{array}{c} 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{array} \right) \right\rangle$$

As with  $R_A$ , only three (column) vectors are required to span  $C_A$ . We observe that

$$(-10) \begin{pmatrix} 1 \\ 0 \\ 1 \\ 0 \\ -2 \end{pmatrix} + (-9) \begin{pmatrix} 0 \\ 1 \\ 1 \\ 0 \\ 3 \end{pmatrix} + (-27) \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} -10 \\ -9 \\ -19 \\ -27 \\ -7 \end{pmatrix}$$

and so  $\mathbf{b} = \begin{pmatrix} -10 \\ -9 \\ -19 \\ -27 \\ -7 \end{pmatrix} \in C_A$ .

### Problem

Explain why in the above example that  $C_A \neq C_B$ . (Hint: Consider the rows of zeros in  $B$ .)

The last example makes at least one direction of the following theorem plausible.

### Theorem

The linear system  $A\mathbf{x} = \mathbf{b}$  is consistent if and only if  $\mathbf{b} \in C_A$ .

We close this unit by relating the rank of a coefficient matrix to the existence of at least one solution to  $\mathbf{A} \mathbf{x} = \mathbf{b}$ .

**Theorem**

The linear system  $\mathbf{A} \mathbf{x} = \mathbf{b}$  is consistent if and only if  $\text{Rank}(\mathbf{A}) = \text{Rank}(\mathbf{A} | \mathbf{b})$ .