

# 1 Section 1.4. Matrix Equations

## 1.1 Products of matrices and vectors

**Definition 1** Let  $A = [\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n]$  be a  $m \times n$  matrix whose  $i$ th column is  $\vec{a}_i \in R^m$ ,  $i = 1, 2, \dots, n$ , and  $\vec{x}$  a (unknown) vector in  $R^n$  :

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}_{n \times 1}. \quad \text{Notice that the number of columns in } A = \text{the number of rows in } \vec{x}.$$

The product  $A\vec{x}$  is defined as the linear combination of  $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n$  with the weight  $x_1, x_2, \dots, x_n$ , i.e.,

$$A\vec{x} = [\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \sum_{i=1}^n x_i \vec{a}_i = x_1 \vec{a}_1 + x_2 \vec{a}_2 + \dots + x_n \vec{a}_n.$$

If  $A$  is  $1 \times n$  matrix, i.e.,  $A = [a_1, a_2, \dots, a_n]$ , then

$$\begin{aligned} A\vec{x} &= [a_1 \ a_2 \ \dots \ a_n] \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1 a_1 + x_2 a_2 + \dots + x_n a_n \\ &= \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad (\text{dot product}) \end{aligned}$$

**Example 2** (a)

$$\begin{aligned} \begin{bmatrix} 1 & 2 & -1 \\ 2 & 5 & 3 \end{bmatrix} \begin{bmatrix} 4 \\ 3 \\ 7 \end{bmatrix} &= 4 \begin{bmatrix} 1 \\ 2 \end{bmatrix} + 3 \begin{bmatrix} 2 \\ 5 \end{bmatrix} + 7 \begin{bmatrix} -1 \\ 3 \end{bmatrix} \\ &= \begin{bmatrix} 1 \cdot 4 + 2 \cdot 3 + (-1) \cdot 7 \\ 2 \cdot 4 + 5 \cdot 3 + 3 \cdot 7 \end{bmatrix} = \begin{bmatrix} 3 \\ 44 \end{bmatrix} \end{aligned}$$

(b)

$$\begin{bmatrix} 2 & -2 & 3 \\ 8 & 0 & 1 \\ -5 & 2 & 0 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} = 1 \begin{bmatrix} 2 \\ 8 \\ -5 \\ 0 \end{bmatrix} + 3 \begin{bmatrix} -2 \\ 0 \\ 2 \\ 1 \end{bmatrix} + 2 \begin{bmatrix} 3 \\ 1 \\ 0 \\ 2 \end{bmatrix} = \begin{bmatrix} 2 \\ 10 \\ 1 \\ 7 \end{bmatrix}$$

Note that for product (b), one can also redo it using "dot product" as

$$\begin{bmatrix} 2 & -2 & 3 \\ 8 & 0 & 1 \\ -5 & 2 & 0 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} = \begin{bmatrix} [2 \ -2 \ 3] \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \\ [8 \ 0 \ 1] \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \\ [-5 \ 2 \ 0] \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \\ [0 \ 1 \ 2] \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} \begin{bmatrix} 2 \\ -2 \\ 3 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \\ \begin{bmatrix} 8 \\ 0 \\ 1 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \\ \begin{bmatrix} -5 \\ 2 \\ 0 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \\ \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix} \cdot \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix} \end{bmatrix} = \begin{bmatrix} 2 \\ 10 \\ 1 \\ 7 \end{bmatrix}$$

The matrix-vector product has the following **PROPERTIES**:

$$\begin{aligned} A(\vec{x} + \vec{y}) &= A\vec{x} + A\vec{y}, & A(\lambda\vec{x}) &= \lambda A\vec{x} \\ (A + B)\vec{x} &= A\vec{x} + B\vec{x}, & (\lambda A)\vec{x} &= \lambda(A\vec{x}) \end{aligned}$$

These relations can be summarized so-called linearity principle:

$$(aA + bB)(\lambda\vec{x} + \delta\vec{y}) = a\lambda A\vec{x} + a\delta A\vec{y} + b\lambda B\vec{x} + b\delta B\vec{y}.$$

## 1.2 Matrix Equations

The system of linear equation, or its equivalent vector equation

$$x_1\vec{a}_1 + x_2\vec{a}_2 + \dots + x_n\vec{a}_n = \vec{b}$$

can now be rewritten as **Matrix equation**:

$$A\vec{x} = \vec{b}.$$

**Theorem 3** Let  $A = [\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n]$  be a  $m \times n$  matrix whose  $i$ th column is  $\vec{a}_i \in R^m$ ,  $i = 1, 2, \dots, n$ , and  $\vec{x}$  an unknown vector in  $R^n$ . The following statements are equivalent:

1. For each vector  $\vec{b} \in R^m$ , the vector equation  $A\vec{x} = \vec{b}$  has a solution.
2. Each vector  $\vec{b} \in R^m$  is a linear combination of  $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n$ , the columns of  $A$ .
3. Column vectors of  $A$  span  $R^m$ .
4. The matrix  $A$  has a pivot position in each row.

**Proof.** (1)  $\implies$  (2) : (1) implies that for each  $\vec{b}$ , there exists a solution  $x_i = \lambda_i$ ,  $i = 1, \dots, n$  such that

$$\lambda_1 \vec{a}_1 + \lambda_2 \vec{a}_2 + \dots + \lambda_n \vec{a}_n = \vec{b}.$$

Therefore, each  $\vec{b}$  is a linear combination of  $\vec{a}_i$ .

(2)  $\implies$  (3) : Since each vector  $\vec{b}$  in  $R^m$  is a linear combination, by definition, each  $\vec{b}$  in  $R^m$  is in the subset spanned by  $\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n$ , i.e.,  $R^m \subset \text{Span}\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_n\} \subset R^m$ .

(3)  $\implies$  (4) : Suppose (3) is true but (4) is NOT true. then the reduced Echelon form of  $A$  has at least one zero row, i.e.,  $A$  is row equivalent to  $E$  by a series of elementary row reductions:

$$A \rightarrow E = \begin{bmatrix} * & * & \dots & * \\ \dots & \dots & \dots & \dots \\ * & * & \dots & * \\ 0 & 0 & \dots & 0 \end{bmatrix}$$

We now consider the augmented matrix

$$\left[ E : \vec{e}_m \right] = \begin{bmatrix} * & * & \dots & * & 0 \\ \dots & \dots & \dots & \dots & \dots \\ * & * & \dots & * & 0 \\ 0 & 0 & \dots & 0 & 1 \end{bmatrix}, \quad (1)$$

and reverse completely all the row reductions we did from  $A$  to  $E$ . This leads to

$$\left[ E : \vec{e}_m \right] \rightarrow \left[ A : \vec{b} \right],$$

where  $\vec{b}$  is a vector in  $R^m$ . Since the system corresponding to (1) is inconsistent,  $A\vec{x} = \vec{b}$  for this particular vector  $\vec{b}$  is inconsistent. In other words,  $\vec{b}$  cannot be possibly a linear combination of  $\vec{a}_i$ . This contradicts to (3).

For instance,

$$A = \begin{bmatrix} 1 & 2 \\ -1 & -2 \end{bmatrix} \xrightarrow{R_1 + R_2} \tilde{R}_2 \begin{bmatrix} 1 & 2 \\ 0 & 0 \end{bmatrix} = E$$

Then (with  $m = 2$ )

$$\left[ E : \vec{e}_m \right] = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix} \xrightarrow{\tilde{R}_2 - R_1} R_2 \begin{bmatrix} 1 & 2 & 0 \\ -1 & -2 & 1 \end{bmatrix}.$$

Thus

$$A\vec{x} = \vec{b}, \quad \vec{b} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \text{ is inconsistent.}$$

(4)  $\implies$  (1) : Suppose for simplicity,  $A$  has the reduced Echelon form

$$A \rightarrow E = \begin{bmatrix} 1 & * & \dots & * & \dots & * \\ 0 & 1 & \dots & * & \dots & \dots \\ \dots & \dots & \dots & \dots & \dots & * \\ 0 & 0 & \dots & 1 & \dots & * \end{bmatrix}.$$

Then, for each  $\vec{b}$ ,  $\left[ A : \vec{b} \right]$  must have the reduced Echelon form

$$\left[ A : \vec{b} \right] \rightarrow \begin{bmatrix} 1 & * & \dots & * & \dots & \vdots & * \\ 0 & 1 & \dots & * & \dots & \vdots & * \\ \dots & \dots & \dots & \dots & \dots & \vdots & * \\ 0 & 0 & \dots & 1 & \dots & \vdots & * \end{bmatrix}.$$

The augmented matrix on the right is consistent. Therefore,  $A\vec{x} = \vec{b}$  is consistent. ■

**Example 4** Let

$$A = \begin{bmatrix} 1 & 3 & 4 \\ -4 & 2 & -6 \\ -3 & -2 & -7 \end{bmatrix}.$$

1. Determine if  $A\vec{x} = \vec{b}$  has a solution for ALL  $\vec{b} \in R^3$ .
2. Describe  $\text{Span}\{\vec{a}_1, \vec{a}_2, \vec{a}_3\}$ , where  $A = [\vec{a}_1, \vec{a}_2, \vec{a}_3]$ .

**Solution.** (1) By row operation, we have

$$\begin{bmatrix} 1 & 3 & 4 \\ -4 & 2 & -6 \\ -3 & -2 & -7 \end{bmatrix} \xrightarrow{\substack{R_2 + 4R_1 \rightarrow R_2 \\ R_3 + 3R_1 \rightarrow R_3}} \begin{bmatrix} 1 & 3 & 4 \\ 0 & 14 & 10 \\ 0 & 7 & 5 \end{bmatrix} \xrightarrow{R_3 - R_2/2 \rightarrow R_3} \begin{bmatrix} 1 & 3 & 4 \\ 0 & 14 & 10 \\ 0 & 0 & 0 \end{bmatrix}.$$

By (4) in Theorem,  $A\vec{x} = \vec{b}$  doesn't has a solution for all  $\vec{b}$ .

**Solution.** (2) We need to find for what  $\vec{b}$ ,

$$\vec{b} \in \text{Span}\{\vec{a}_1, \vec{a}_2, \vec{a}_3\}?$$

In other words, we need to describe all  $\vec{b}$  such that  $A\vec{x} = \vec{b}$  has a solution. In other words, we need to find conditions on  $\vec{b}$  under which

$$A\vec{x} = \vec{b} \text{ is consistent.}$$

To this end, we perform exactly the same row operations as above for the augmented matrix

$$\left[ A : \vec{b} \right] :$$

$$\begin{aligned} \left[ A : \vec{b} \right] &= \begin{bmatrix} 1 & 3 & 4 & b_1 \\ -4 & 2 & -6 & b_2 \\ -3 & -2 & -7 & b_3 \end{bmatrix} \xrightarrow{\substack{R_2 + 4R_1 \rightarrow R_2 \\ R_3 + 3R_1 \rightarrow R_3}} \begin{bmatrix} 1 & 3 & 4 & b_1 \\ 0 & 14 & 10 & b_2 + 4b_1 \\ 0 & 7 & 5 & b_3 + 3b_1 \end{bmatrix} \\ &\xrightarrow{R_3 - R_2/2 \rightarrow R_3} \begin{bmatrix} 1 & 3 & 4 & b_1 \\ 0 & 14 & 10 & b_2 + 4b_1 \\ 0 & 0 & 0 & b_3 + 3b_1 - b_2/2 - 2b_1 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 3 & 4 & b_1 \\ 0 & 14 & 10 & 4b_1 + b_2 \\ 0 & 0 & 0 & b_1 - b_2/2 + b_3 \end{bmatrix}. \end{aligned}$$

We know that  $A\vec{x} = \vec{b}$  has a solution iff the last column is not pivot, i.e.,

$$b_1 - b_2/2 + b_3 = 0, \text{ or } b_3 = -b_1 + b_2/2.$$

We conclude that  $\vec{b} \in \text{Span}\{\vec{a}_1, \vec{a}_2, \vec{a}_3\}$  iff  $b_3 = -b_1 + b_2/2$ , and

$$\begin{aligned} \vec{b} &= \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ -b_1 + b_2/2 \end{bmatrix} = \begin{bmatrix} b_1 + 0b_2 \\ 0b_1 + b_2 \\ -b_1 + b_2/2 \end{bmatrix} \\ &= \begin{bmatrix} b_1 \\ 0b_1 \\ -b_1 \end{bmatrix} + \begin{bmatrix} 0b_2 \\ b_2 \\ b_2/2 \end{bmatrix} = b_1 \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix} + b_2 \begin{bmatrix} 0 \\ 1 \\ 1/2 \end{bmatrix} \end{aligned}$$

for any  $b_1$  and  $b_2$ . Later, we call it "parametric representation". Therefore,

$$\text{Span}\{\vec{a}_1, \vec{a}_2, \vec{a}_3\} = \text{Span}\left\{ \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 1/2 \end{bmatrix} \right\}.$$

## Homework 1.4:

#5,7,9,11,13,17,21,23