Slide 4-Matrices Algebra I MAT2040 Linear Algebra

Definition 4.1 (Set of Matrices)

$$\mathbb{R}^{m \times n} = \{m \times n \text{ Matrix} | \text{ entries } \in \mathbb{R}\}$$

$$\mathbb{C}^{m\times n} = \{m \times n \text{ Matrix} | \text{ entries} \in \mathbb{C}\}\$$

Given a $n \times n$ Matrix A, A is called a **square matrix**.

$$\mathbb{R}^{n \times n} = \{ n \times n \text{ Matrix} | \text{ entries } \in \mathbb{R} \}$$

$$\mathbb{C}^{n\times n} = \{n \times n \text{ Matrix} | \text{ entries } \in \mathbb{C}\}$$

Definition 4.2 (Set of Column Vectors)

$$\mathbb{R}^n = \mathbb{R}^{n \times 1} = \{ n \times 1 \text{ Matrix} | \text{ entries } \in \mathbb{R} \}$$

$$\mathbb{C}^n = \mathbb{C}^{n \times 1} = \{ n \times 1 \text{ Matrix} | \text{ entries } \in \mathbb{C} \}$$

Matrix Operation Definition

Definition 4.3 (Matrix Equality) Let A and B be two $m \times n$ -matrices. A and B are **equal** (written as "A = B") if $a_{ij} = b_{ij}$, for every $i = 1, \dots, m, j = 1 \dots, n$.

Definition 4.4 (Matrix addition) Let A and B be two $m \times n$ -matrices. The **sum of** A **and** B (written as "A + B") is defined to be the $m \times n$ -matrix C with entries $c_{ij} = a_{ij} + b_{ij}$, for every $i = 1, 2, \ldots, m$, $j = 1, 2, \ldots, n$.

Definition 4.5 (Scalar Multiplication) Let A be an $m \times n$ -matrix and α be any real number (α in \mathbb{R}). The **multiplication** of α and A (written as " αA ") is defined to be the $m \times n$ -matrix D with entries $d_{ij} = \alpha a_{ij}$, for every $i = 1, 2, \ldots, m, j = 1, 2, \ldots, n$.

Example 4.6

(1)

$$A = \begin{bmatrix} 1 & -2 & 3 \\ -5 & 4 & -1 \end{bmatrix}, B = \begin{bmatrix} 1 & -5 \\ -2 & 4 \\ 3 & -1 \end{bmatrix}, C = \begin{bmatrix} 1 & x & 3 \\ -5 & 4 & y \end{bmatrix}$$

Then $A \neq B$ (A and B have different sizes).

$$A = C \Rightarrow x = -2, y = -1.$$

(2)

$$A = \begin{bmatrix} 2 & -3 & 4 \\ 1 & 0 & -7 \end{bmatrix}, \quad B = \begin{bmatrix} 6 & 2 & -4 \\ 3 & 5 & 2 \end{bmatrix}.$$

Then

$$A + (-1)B = \begin{bmatrix} 2 & -3 & 4 \\ 1 & 0 & -7 \end{bmatrix} + (-1) \begin{bmatrix} 6 & 2 & -4 \\ 3 & 5 & 2 \end{bmatrix}$$
$$= \begin{bmatrix} 2 & -3 & 4 \\ 1 & 0 & -7 \end{bmatrix} + \begin{bmatrix} -6 & -2 & 4 \\ -3 & -5 & -2 \end{bmatrix}$$
$$= \begin{bmatrix} -4 & -5 & 8 \\ -2 & -5 & -9 \end{bmatrix}$$

$$A - B = \begin{bmatrix} 2 & -3 & 4 \\ 1 & 0 & -7 \end{bmatrix} - \begin{bmatrix} 6 & 2 & -4 \\ 3 & 5 & 2 \end{bmatrix} = \begin{bmatrix} -4 & -5 & 8 \\ -2 & -5 & -9 \end{bmatrix}$$

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Definition 4.7 (Zero Matrix) Let $A=(a_{ij})_{m\times n}$ such that $a_{ij}=0, \forall i=1,\cdots,m, j=1,\cdots,n$, then A is a **zero matrix**, denoted by $O=O_{m\times n}$.

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Theorem 4.8 (Properties of Matrices Operations)

Let $A, B, C \in \mathbb{R}^{m \times n}$, $\alpha, \beta \in \mathbb{R}$.

(1)
$$A + B = B + A$$
.

$$(2) A + (B + C) = (A + B) + C.$$

(We can therefore use the notation A + B + C.)

(3)
$$\alpha(\beta A) = (\alpha \beta) A$$
.

(4)
$$\alpha(A+B) = \alpha A + \alpha B$$
.

(5)
$$(\alpha + \beta)A = \alpha A + \beta A$$
.

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Proof of (4) $\alpha(A+B) = \alpha A + \alpha B$ (others are left as exercises). Fix a row number i, and column number j.

$$(\alpha(A+B))_{ij} = \alpha(A+B)_{ij}$$
 (definition of scalar multiplication)
= $\alpha(a_{ij} + b_{ij})$ (definition of matrix addition).

Similarly,

$$(\alpha A + \alpha B)_{ij} = (\alpha A)_{ij} + (\alpha B)_{ij}$$
 (definition of matrix addition)
= $\alpha a_{ij} + \alpha b_{ij}$ (definition of scalar multiplication).

Now we are just comparing real (or complex) numbers, and you know the rules for comparing these:

$$\alpha(a_{ij} + b_{ij}) = \alpha a_{ij} + \alpha b_{ij}$$
 (distributive property).

We see that the (i,j)th entry of $\alpha(A+B)$ equals the (i,j)th entry of $\alpha A + \alpha B$ for all i and j, which is exactly the definition of two matrices being equal.

(Vectors from Matrix) Let $A = (a_{ij})_{m \times n}$ be a matrix, then the **ith row vector** is given by

$$\vec{\mathbf{a}}_i = (\mathbf{a}_{i1}, \mathbf{a}_{i2}, \cdots, \mathbf{a}_{in}), \quad i = 1, \cdots, m$$

And the **jth column vector** is given by

$$\mathbf{a}_j = \begin{bmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{mj} \end{bmatrix}, \quad j = 1, \cdots, n$$

Then

$$A = [\mathbf{a}_1, \cdots, \mathbf{a}_n] = \begin{bmatrix} \vec{\mathbf{a}}_1 \\ \vec{\mathbf{a}}_2 \\ \vdots \\ \vec{\mathbf{a}}_m \end{bmatrix}$$

If $A, B \in \mathbb{R}^{m \times n}, \alpha \in \mathbb{R}$,

$$A = [\mathbf{a}_1, \cdots, \mathbf{a}_n] = \begin{bmatrix} \vec{\mathbf{a}}_1 \\ \vec{\mathbf{a}}_2 \\ \vdots \\ \vec{\mathbf{a}}_m \end{bmatrix}, B = [\mathbf{b}_1, \cdots, \mathbf{b}_n] = \begin{bmatrix} \vec{\mathbf{b}}_1 \\ \vec{\mathbf{b}}_2 \\ \vdots \\ \vec{\mathbf{b}}_m \end{bmatrix}$$

Then

$$A + B = [\mathbf{a}_1 + \mathbf{b}_1, \cdots, \mathbf{a}_n + \mathbf{b}_n] = \begin{bmatrix} \vec{\mathbf{a}}_1 + \mathbf{b}_1 \\ \vec{\mathbf{a}}_2 + \vec{\mathbf{b}}_2 \\ \vdots \\ \vec{\mathbf{a}}_m + \vec{\mathbf{b}}_m \end{bmatrix}$$

$$\alpha A = [\alpha \mathbf{a}_1, \cdots, \alpha \mathbf{a}_n] = \begin{bmatrix} \alpha \vec{\mathbf{a}}_1 \\ \alpha \vec{\mathbf{a}}_2 \\ \vdots \\ \alpha \vec{\mathbf{a}}_m \end{bmatrix}$$

Matrix-Vector Multiplication

Definition I (From Steven's Book)

Let
$$A = (a_{ij})_{m \times n} = \begin{bmatrix} \vec{a}_1 \\ \vec{a}_2 \\ \vdots \\ \vec{a}_m \end{bmatrix} \in \mathbb{R}^{m \times n}$$
, where $\vec{a}_1, \vec{a}_2, \cdots, \vec{a}_m$ are row vectors

and $\mathbf{u} = (u_i)_{n \times 1}$ is a column vector, then

$$A\mathbf{u} = \begin{bmatrix} \vec{\mathbf{a}}_1 \mathbf{u} \\ \vec{\mathbf{a}}_2 \mathbf{u} \\ \vdots \\ \vec{\mathbf{a}}_m \mathbf{u} \end{bmatrix}$$

where

$$\vec{\mathbf{a}}_i\mathbf{u}=a_{i1}u_1+a_{i2}u_2+\cdots+a_{in}u_n$$

is the scalar product of \vec{a}_i and \mathbf{u} .

Matrix-Vector Multiplication: Second Definition

Definition II (From Beezer's Notes) Let $A=(a_{ij})_{m\times n}=[\mathbf{a}_1,\mathbf{a}_2,\cdots,\mathbf{a}_n]\in\mathbb{R}^{m\times n}$, where $\mathbf{a}_1,\mathbf{a}_2,\cdots,\mathbf{a}_n$ are column vectors and $\mathbf{u}=(u_i)_{n\times 1}$ is also a column vector, then the matrix-vector product $A\mathbf{u}$ is

$$u_1\mathbf{a}_1 + u_2\mathbf{a}_2 + \cdots + u_n\mathbf{a}_n$$

which is a **linear Combination** of column vectors $\mathbf{a}_1, \mathbf{a}_2, \cdots, \mathbf{a}_n$ with weights u_1, \cdots, u_n .

Remark: Two definitions produce the same results.

Example 4.9

$$A = \begin{bmatrix} 1 & 4 & 2 & 3 & 5 \\ -2 & 1 & 3 & 0 & -1 \\ 0 & 7 & -1 & -2 & 4 \end{bmatrix}, \ \mathbf{u} = \begin{bmatrix} 1 \\ -2 \\ 0 \\ 5 \\ -1 \end{bmatrix},$$

By definition 1:

$$A\mathbf{u} = \begin{bmatrix} \vec{\mathbf{a}}_1 \mathbf{u} \\ \vec{\mathbf{a}}_2 \mathbf{u} \\ \vec{\mathbf{a}}_3 \mathbf{u} \end{bmatrix} = \begin{bmatrix} 1*1+4*(-2)+2*0+3*5+5*(-1) \\ (-2)*1+1*(-2)+3*0+0*5+(-1)*(-1) \\ 0*1+7*(-2)+(-1)*0+(-2)*5+4*(-1) \end{bmatrix}$$
$$= \begin{bmatrix} 3 \\ -3 \\ -28 \end{bmatrix}$$

By definition 2:

$$A\mathbf{u} = u_1 \mathbf{a}_1 + u_2 \mathbf{a}_2 + u_3 \mathbf{a}_3 + u_4 \mathbf{a}_4 + u_5 \mathbf{a}_5$$

$$= 1 \cdot \begin{bmatrix} 1 \\ -2 \\ 0 \end{bmatrix} + (-2) \cdot \begin{bmatrix} 4 \\ 1 \\ 7 \end{bmatrix} + 0 \cdot \begin{bmatrix} 2 \\ 3 \\ -1 \end{bmatrix} + 5 \cdot \begin{bmatrix} 3 \\ 0 \\ -2 \end{bmatrix} + (-1) \cdot \begin{bmatrix} 5 \\ -1 \\ 4 \end{bmatrix}$$

$$= \begin{bmatrix} 3 \\ -3 \\ -28 \end{bmatrix}$$

Let

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}, \quad \mathbf{b} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

We can write $A\mathbf{x}$ as the linear combination of the column vectors of A with weights x_1, x_2, \dots, x_n , so $A\mathbf{x} = \mathbf{b}$ is equivalent to

$$x_1 \begin{bmatrix} a_{11} \\ a_{21} \\ \vdots \\ a_{m1} \end{bmatrix} + x_2 \begin{bmatrix} a_{12} \\ a_{22} \\ \vdots \\ a_{m2} \end{bmatrix} + \cdots + x_n \begin{bmatrix} a_{1n} \\ a_{2n} \\ \vdots \\ a_{mn} \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

Now use the definition of scalar multiplication and the matrix addition, one has

$$\begin{bmatrix} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n \\ \vdots \\ a_{m1}x_1 + a_{m2}x_2 + a_{m3}x_3 + \dots + a_{mn}x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

Thus, $A\mathbf{x} = \mathbf{b}$ is equivalent to

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n = b_1,$$

 $a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n = b_2,$
 $a_{31}x_1 + a_{32}x_2 + a_{33}x_3 + \dots + a_{3n}x_n = b_3,$
 \vdots
 $a_{m1}x_1 + a_{m2}x_2 + a_{m3}x_3 + \dots + a_{mn}x_n = b_m.$

Theorem 4.10 (Equivalent Condition for a Consistent Linear **System**) The linear system $A\mathbf{x} = \mathbf{b}$ is consistent if and only if \mathbf{b} is a linear combination of the column vectors of A.

Proof. Suppose that
$$A = [\mathbf{a}_1, \dots, \mathbf{a}_n]$$
 and $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$.

By the definition of matrix-vector multiplication, $A\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$.

By the definition of matrix-vector multiplication, $A\mathbf{x} = \mathbf{b}$ is equivalent to

$$x_1\mathbf{a}_1+x_2\mathbf{a}_2+\cdots+x_n\mathbf{a}_n=\mathbf{b}.$$

This is also equivalent to that \mathbf{b} is a linear combination of column vectors of A.

Definition 4.11 (Matrix Product) Let $A \in \mathbb{R}^{m \times n}$ and $B = [\mathbf{b}_1, \mathbf{b}_2, \cdots, \mathbf{b}_l] \in \mathbb{R}^{n \times r}$, then the matrix product of A by B is a $m \times r$ matrix defined by

$$AB = [A\mathbf{b}_1, A\mathbf{b}_2, \cdots, A\mathbf{b}_r].$$

Remark

- 1. Matrix product is a natural generalization of the matrix-vector product.
- 2. AB exists only and if only the number of columns of A equal to the number of rows of B.

Theorem 4.12 (Matrix Product Alternative Definition) Let

$$A = (a_{ik})_{m imes n} = \begin{bmatrix} \vec{\mathbf{a}}_1 \\ \vec{\mathbf{a}}_2 \\ \vdots \\ \vec{\mathbf{a}}_m \end{bmatrix}$$
 and $B = (b_{kj})_{n imes l} = [\mathbf{b}_1, \mathbf{b}_2, \cdots, \mathbf{b}_r] \in \mathbb{R}^{n imes r}$, then $AB = [A\mathbf{b}_1, A\mathbf{b}_2, \cdots, A\mathbf{b}_l] = C \triangleq (c_{ij})_{m imes r}$

where
$$c_{ij} = \sum_{k=1}^{n} a_{ik} b_{kj} = \vec{\mathbf{a}}_{i} \mathbf{b}_{j}$$
.

Note that: \vec{a}_i is a $1 \times n$ matrix (row vector) while \mathbf{b}_j is a $n \times 1$ matrix (column vector), the product $\vec{a}_i \mathbf{b}_j$ will be a 1×1 matrix which is a scalar.

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Proof.

$$c_{ij} = (A\mathbf{b}_j)_i = \begin{bmatrix} \vec{\mathbf{a}}_1\mathbf{b}_j \\ \vdots \\ \vec{\mathbf{a}}_m\mathbf{b}_j \end{bmatrix}_i - - - - - - - - (\text{ith entry of } A\mathbf{b}_j)$$

$$c_{ij} = (A\mathbf{b}_j)_i$$

= $\vec{a}_i \mathbf{b}_j - ($ by using matrix – vector multiplication definition 2)
= $a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj}$

Remark. Most of the book uses the second definition for the matrix product.

Example 4.13

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

Then by using matrix-matrix multiplication definition 1,

$$AB = \begin{bmatrix} A \begin{bmatrix} 1 \\ -1 \\ 1 \\ 6 \\ 1 \end{bmatrix}, A \begin{bmatrix} 6 \\ 4 \\ 1 \\ 4 \\ -2 \end{bmatrix}, A \begin{bmatrix} 2 \\ 3 \\ 2 \\ -1 \\ 3 \end{bmatrix}, A \begin{bmatrix} 1 \\ 2 \\ 3 \\ 2 \\ 0 \end{bmatrix} = \begin{bmatrix} 28 & 17 & 20 & 10 \\ 20 & -13 & -3 & -1 \\ -18 & -44 & 12 & -3 \end{bmatrix}$$

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Alternatively by using matrix-matrix multiplication definition 2, let $AB = C = (c_{ij})_{3\times 4}$, all the entries of C can be figured out. For example, $c_{12} = \vec{a}_1 b_2 = 1*6+2*4+(-1)*1+4*4+6*(-2) = 17$, and other entries can also be calculated.

Note that BA does not exist, because the number of column of B is not equal to the number of rows of A.

Remark

- 1. AB exists if and only if the number of columns of A equals to the number of rows of B.
- 2. AB exists does not imply that BA exists.
- 3. Even if both AB and BA exists, they are generally not equal (A and B are generally not commutative).

Theorem 4.14 (Properties of Matrix-vector multiplication) Let $A \in \mathbb{R}^{m \times n}$, $\mathbf{x}, \mathbf{y} \in \mathbb{R}^n$, $\alpha \in \mathbb{R}$, then

$$(1) A(\mathbf{x} + \mathbf{y}) = A\mathbf{x} + A\mathbf{y},$$

(2)
$$A(\alpha \mathbf{x}) = (\alpha A)\mathbf{x} = \alpha(A\mathbf{x})$$
,

Proof. Only show (1). Let
$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$
, $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$, then

$$A(\mathbf{x} + \mathbf{y}) = (x_1 + y_1)\mathbf{a}_1 + (x_2 + y_2)\mathbf{a}_2 + \ldots + (x_n + y_n)\mathbf{a}_n$$

$$A\mathbf{x} + A\mathbf{y} = x_1\mathbf{a}_1 + x_2\mathbf{a}_2 + \dots + x_n\mathbf{a}_n + y_1\mathbf{a}_1 + y_2\mathbf{a}_2 + \dots + y_n\mathbf{a}_n$$

= $(x_1 + y_1)\mathbf{a}_1 + (x_2 + y_2)\mathbf{a}_2 + \dots + (x_n + y_n)\mathbf{a}_n$.

Corollary: If $A \in \mathbb{R}^{m \times n}$, $\mathbf{x}_i \in \mathbb{R}^n$, $\alpha_i \in \mathbb{R}$ $(i = 1, \dots, s)$, then

$$A(\alpha_1 \mathbf{x}_1 + \dots + \alpha_s \mathbf{x}_s) = \alpha_1 A \mathbf{x}_1 + \dots + \alpha_s A \mathbf{x}_s.$$

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Theorem 4.15 (Properties of Matrix Product I) Let $A \in \mathbb{R}^{m \times n}$,

 $B, C \in \mathbb{R}^{n \times l}$, $\alpha \in \mathbb{R}$, then

(1) A(B+C) = AB + AC

(2) $\alpha(AB) = (\alpha A)B = A(\alpha B)$.

If $A \in \mathbb{R}^{n \times I}$, $B, C \in \mathbb{R}^{m \times n}$, then (3) (B + C)A = BA + CA

Proof. Only show (1). Others are excises.

Method 1: Suppose

$$B = [\mathbf{b}_1, \cdots, \mathbf{b}_l], C = [\mathbf{c}_1, \cdots, \mathbf{c}_l],$$

then

$$B + C = [(\mathbf{b}_1 + \mathbf{c}_1), (\mathbf{b}_2 + \mathbf{c}_2), \dots, (\mathbf{b}_l + \mathbf{c}_l)].$$

Therefore

$$A(B+C) = [A(\mathbf{b}_1 + \mathbf{c}_1), A(\mathbf{b}_2 + \mathbf{c}_2), \dots, A(\mathbf{b}_l + \mathbf{c}_l)].$$

On the other hand,

$$AB = [A\mathbf{b}_1, A\mathbf{b}_2, \cdots, A\mathbf{b}_l], \quad AC = [A\mathbf{c}_1, A\mathbf{c}_2, \cdots, A\mathbf{c}_l].$$

Thus

$$\begin{aligned} AB + AC &= [A\mathbf{b}_1, \ A\mathbf{b}_2, \ \cdots, \ A\mathbf{b}_I] + [A\mathbf{c}_1, \ A\mathbf{c}_2, \ \cdots, \ A\mathbf{c}_I] \\ &= [A(\mathbf{b}_1 + \mathbf{c}_1), \ A(\mathbf{b}_2 + \mathbf{c}_2), \ \ldots, \ A(\mathbf{b}_I + \mathbf{c}_I)] \\ &= A(B + C) \end{aligned}$$

Method 2: The (i, k)-entry of A is a_{ik} , the (k, j)-entry of B + C is $b_{kj} + c_{kj}$, thus the (i, j)-entry of A(B + C) is

$$\sum_{k=1}^{n} a_{ik} (b_{kj} + c_{kj}) = \sum_{k=1}^{n} a_{ik} b_{kj} + \sum_{k=1}^{n} a_{ik} c_{kj}.$$

On the other hand, the (i, j)-entry of AB + AC is

$$\sum_{k=1}^{n} a_{ik} b_{kj} + \sum_{k=1}^{n} a_{ik} c_{kj}.$$

Thus,

$$A(B+C)=AB+AC.$$

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Lemma 4.16 Let $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{n \times p}$, $\mathbf{x} \in \mathbb{R}^{p}$, then

$$(AB)\mathbf{x} = A(B\mathbf{x}). \quad (Assocativity)$$
Proof. Let $B = [\mathbf{b}_1, \mathbf{b}_2, \ \dots, \mathbf{b}_p], \ \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_p \end{bmatrix}$ Then,

$$A(B\mathbf{x}) = A(x_1\mathbf{b}_1 + x_2\mathbf{b}_2 + \dots + x_p\mathbf{b}_p)$$

$$= x_1A\mathbf{b}_1 + x_2A\mathbf{b}_2 + \dots + x_pA\mathbf{b}_p$$

$$= [A\mathbf{b}_1, A\mathbf{b}_2, \dots, A\mathbf{b}_p]\mathbf{x}$$

$$= (AB)\mathbf{x}$$

Where the corollary of Theorem 4.14 (linearity of Ax) is used.

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Theorem 4.17 (Property of Matrix Product II) Let $A \in \mathbb{R}^{m \times n}$, $B \in \mathbb{R}^{n \times p}$, $C \in \mathbb{R}^{p \times \ell}$. then

$$(AB)C = A(BC)$$
. (Assocativity)

Proof.

Method 1:

Let $C = [\mathbf{c}_1, \mathbf{c}_2, \dots, \mathbf{c}_{\ell}]$, then

$$(AB)C = [(AB)\mathbf{c}_1, (AB)\mathbf{c}_2, \ldots, (AB)\mathbf{c}_\ell]$$

and because $BC = [B\mathbf{c}_1, B\mathbf{c}_2, \dots, B\mathbf{c}_\ell]$, we have

$$A(BC) = [A(B\mathbf{c}_1), \ A(B\mathbf{c}_2), \ \ldots, \ A(B\mathbf{c}_\ell)].$$

Since $(AB)\mathbf{c}_i = A(B\mathbf{c}_i)$ $(i = 1, \dots, \ell)$ by using **Lemma 4.16**, thus (AB)C = A(BC).

Another proof is put into the appendix.

Therefore, we can write (AB)C = A(BC) = ABC.

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