CHAPTER 2 MATRICES AND LINEAR TRANSFORMATIONS 2.1 Matrix Multiplication

From matrix-vector products to matrix-matrix multiplication.

Let $\mathbf{v}, \mathbf{x}, \mathbf{y} \in \mathcal{R}^n$. Suppose A and B are $n \times n$ matrices.



Does there exist an *n* x *n* matrix *C* such that $\mathbf{y} = C\mathbf{v}$? (for all $\mathbf{v} \in \mathbf{R}^n$)

In general, given $A \in \mathcal{R}^{m \times n}$ and $B \in \mathcal{R}^{n \times p}$, we wish to find $C \in \mathcal{R}^{m \times p}$ such that $C\mathbf{v} = A(B\mathbf{v})$ for all $\mathbf{v} \in \mathcal{R}^p$.

Proposition

Let A be an $m \times n$ matrix and B be an $n \times p$ matrix. Then there is a unique $m \times p$ matrix C such that $C\mathbf{v} = A(B\mathbf{v})$ for every $p \times 1$ vector \mathbf{v} . Furthermore,

 $C = \begin{bmatrix} A\mathbf{b}_1 & A\mathbf{b}_2 & \cdots & A\mathbf{b}_p \end{bmatrix}$

Proof

Existence:

Uniqueness:

Proposition

Let A be an $m \times n$ matrix and B be an $n \times p$ matrix. Then there is a unique $m \times p$ matrix C such that $C\mathbf{v} = A(B\mathbf{v})$ for every $p \times 1$ vector \mathbf{v} . Furthermore,

 $C = \begin{bmatrix} A\mathbf{b}_1 & A\mathbf{b}_2 & \cdots & A\mathbf{b}_p \end{bmatrix}$

Proof

Existence:

$$A(B\mathbf{v}) = A(v_1\mathbf{b}_1 + v_2\mathbf{b}_2 + \dots + v_p\mathbf{b}_p) \quad \text{Def. of Matrix-Vector products}$$

$$= A(v_1\mathbf{b}_1) + A(v_2\mathbf{b}_2) + \dots + A(v_p\mathbf{b}_p) \text{ Theorem 1.3(a)}$$

$$= v_1A\mathbf{b}_1 + v_2A\mathbf{b}_2 + \dots + v_pA\mathbf{b}_p \quad \text{Theorem 1.3(b)}$$

$$= \begin{bmatrix} A\mathbf{b}_1 & A\mathbf{b}_2 & \dots & A\mathbf{b}_p \end{bmatrix} \mathbf{v}.$$
Def. of Matrix-Vector products

Uniqueness: Suppose there is another $Q \in \mathcal{R}^{m \times p}$ such that $Q\mathbf{v} = A(B\mathbf{v})$ for all $\mathbf{v} \in \mathcal{R}^p$. Then $\mathbf{q}_j = Q\mathbf{e}_j = A(B\mathbf{e}_j) = A\mathbf{b}_j = \mathbf{c}_j$, i.e., Q = C.

Definition

Let A be an $m \times n$ matrix and B be an $n \times p$ matrix. We define the (matrix) product AB to be the $m \times p$ matrix whose *j*th column is $A\mathbf{b}_j$. That is,

 $AB = \begin{bmatrix} A\mathbf{b}_1 & A\mathbf{b}_2 & \cdots & A\mathbf{b}_p \end{bmatrix}.$

Dimension requirement: $(m \times n)(n \times p) = (m \times p)$



The Row-Column Rule for the (i, j)-entry of a Matrix Product

$$\operatorname{row} i \text{ of } A \longrightarrow \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{i1} & a_{i2} & \cdots & a_{in} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \begin{bmatrix} b_{11} & \cdots & b_{1j} \\ b_{21} & \cdots & b_{2p} \\ \vdots \\ b_{n1} & \cdots & b_{nj} \end{bmatrix} \cdots & b_{np} \end{bmatrix}$$
$$\begin{bmatrix} a_{i1} & a_{i2} & \cdots & a_{in} \\ \vdots \\ b_{nj} \end{bmatrix} \begin{bmatrix} b_{1j} \\ b_{2j} \\ \vdots \\ b_{nj} \end{bmatrix} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj}$$



Questions:

1) Does there exist a matrix *K* such that KA = A for all $A \in \mathcal{M}_{m \times n}$?

2) If C = AB, then how can we express C^T in terms of A and B?

Theorem 2.1

Let A and B be $k \times m$ matrices, C be an $m \times n$ matrix, and P and Q be $n \times p$ matrices. Then the following statements are true:

(a) s(AC) = (sA)C = A(sC) for any scalar s.
(b) A(CP) = (AC)P. (associative law of matrix multiplication)
(c) (A + B)C = AC + BC. (right distributive law)
(d) C(P + Q) = CP + CQ. (left distributive law)
(e) I_kA = A = AI_m.
(f) The product of any matrix and a zero matrix is a zero matrix.
(g) (AC)^T = C^TA^T.

Proof You show (a), (d), (e), and (f). (b) dimension check (AC)P: $[(k \times m)(m \times n)](n \times p) = (k \times p)$ A(CP): $(k \times m)[(m \times n)(n \times p)] = (k \times p)$; Let $\mathbf{u}_j \equiv \text{column } j$ of $CP = C\mathbf{p}_j \Rightarrow \text{column } j$ of $A(CP) = A\mathbf{u}_j = A(C\mathbf{p}_j)$, also, column j of $(AC)P = (AC)\mathbf{p}_j = A(C\mathbf{p}_j)$ by definition.

(c) you do the dimension check; column j of $(A + B)C = (A + B)\mathbf{c}_j = A\mathbf{c}_j + B\mathbf{c}_j$ = column j of AC + BC (g) you do the dimension check; the (i,j)-entry of $(AC)^T$ is the (j,i)-entry of AC, which is $\begin{bmatrix} a_{j1} & a_{j2} & \cdots & a_{jm} \\ \hline \mathbf{row} \, \mathbf{j} \, \mathbf{of} \, A \end{bmatrix} \begin{bmatrix} c_{1i} \\ c_{2i} \\ \vdots \\ c_{mi} \end{bmatrix} = a_{j1}c_{1i} + a_{j2}c_{2i} + \cdots + a_{jm}c_{mi}$ the (i,j)-entry of $C^T A^T$ is the product of row *i* of C^T and column *j* of A^T , which is $\begin{bmatrix} c_{1i} & c_{2i} & \cdots & c_{mi} \\ (\text{column } i \text{ of } C)^T \end{bmatrix} \begin{bmatrix} a_{j1} \\ a_{j2} \\ \vdots \\ a_{im} \end{bmatrix} = c_{1i}a_{j1} + c_{2i}a_{j2} + \cdots + c_{mi}a_{jm}$

The associative law allows using *ABC* to represent (*AB*)*C* or *A*(*BC*). Power of square matrices: $A \in \mathcal{M}_{n \times n}$, $A^k = A A \cdots A$ (*k* times), and by convention, $A^1 = A$, $A^0 = I_n$. Augmented matrix: If A and B are matrices with the same number of rows, then the augmented matrix [A B] is the matrix whose columns are the columns of A followed by the columns of B.

Property: For any $P \in \mathbb{R}^{m \times n}$ and matrices A and B with n rows, P[A B] = [PA PB].

Example:
$$A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$
, $B = \begin{bmatrix} 2 & 0 & 1 \\ -1 & 3 & 1 \end{bmatrix}$, and $P = \begin{bmatrix} 1 & 2 \\ 2 & -1 \\ 0 & 1 \end{bmatrix}$.
Then $[A B] = \begin{bmatrix} 1 & 0 & 2 & 0 & 1 \\ 0 & 1 & -1 & 3 & 1 \end{bmatrix}$, $PA = P$, and $PB = \begin{bmatrix} 0 & 6 & 3 \\ 5 & -3 & 1 \\ -1 & 3 & 1 \end{bmatrix}$.
Also, $P[A B] = P \begin{bmatrix} 1 & 0 & 2 & 0 & 1 \\ 0 & 1 & -1 & 3 & 1 \end{bmatrix} = [PA PB]$ (you check).

Diagonal matrices: $A = [a_{ij}] \in \mathcal{M}_{n \times n}$ and $a_{ij} = 0$ for $i \neq j$, sometimes denoted by $A = \text{diag}[a_{11} \cdots a_{nn}]$.

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix} \quad B = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 2 \end{bmatrix} \quad AB = \begin{bmatrix} 3 & 0 & 0 \\ 0 & -2 & 0 \\ 0 & 0 & 6 \end{bmatrix}$$

Property:
$$A = \text{diag}[a_{11} \cdots a_{nn}], B = \text{diag}[b_{11} \cdots b_{nn}] \Rightarrow$$

 $AB = \text{diag}[a_{11}b_{11} \cdots a_{nn}b_{nn}].$

Symmetric matrices: $A = [a_{ij}] \in \mathcal{M}_{n \times n}$ and $a_{ij} = a_{ji}$, or $A = A^T$.

$$A = \begin{bmatrix} 1 & 2 & 4 \\ 2 & 3 & -1 \\ 4 & -1 & 5 \end{bmatrix} = A^T \qquad B = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \neq B^T$$

Property: For any $A \in \mathcal{M}_{n \times n}$, AA^T and A^TA are square and symmetric: $(AA^T)^T = A^{TT}A^T = AA^T$ and $(A^TA)^T = A^TA^{TT} = A^TA$. **Homework Set for 2.1**

```
Section 2.1: Problems 1, 3, 7, 9, 13, 19, 21, 23, 25, 27, 29, 31, 33, 35, 37, 41, 47, 49
```