Matrix multiplication

Recall that if A is an mxn matrix and
$$\vec{x} = \begin{bmatrix} \vec{x}_1 \\ \vdots \\ x_n \end{bmatrix}$$

an n-vector, where $A = \begin{bmatrix} \vec{a}_1 & \dots & \vec{a}_n \end{bmatrix}$,
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 $A\vec{x} = x_1\vec{a}_1 + x_2\vec{a}_2 + \dots + x_n\vec{a}_n$, an m-vector.

In this section, we extend this to multiplication of matrices. First we show how it connects to transformations.

Composition of transformations

Suppose $T: \mathbb{R}^{k} \to \mathbb{R}^{n}$ and $S: \mathbb{R}^{n} \to \mathbb{R}^{m}$ are transformations. If we first apply T and then apply S, we get the composition, or composite, $S \circ T$. $\mathbb{R}^{k} \xrightarrow{T} \mathbb{R}^{n} \xrightarrow{S} \mathbb{R}^{m}$ $\int_{S \circ T}$

We define SoT as follows. If \vec{x} is a k-vector, then SoT $(\vec{x}) := S(T(\vec{x}))$.

Note that T(x) is an n-vector, so it makes sense as

input for S.

Note that SoT takes as input k-vectors, and outputs
m-vectors.
nxk
Now let B be the matrix associated to T and A
the matrix associated to S. That is,

$$T(\vec{x}) = B\vec{x}$$
 and $S(\vec{y}) = A\vec{y}$.
Suppose $B = [\vec{b}_1, \vec{b}_2, \dots, \vec{b}_k]$, where \vec{b}_1 is the ith column,
and $\vec{x} = \begin{bmatrix} \vec{x}_1 \\ \vdots \\ \vdots \\ \vdots \\ \end{bmatrix}$. Then
 $S \circ T(\vec{x}) = S(T(\vec{x})) = A(B\vec{x})$
 $= A(x_1\vec{b}_1 + x_2\vec{b}_2 + \dots + x_k\vec{b}_k)$
 $= x_1A\vec{b}_1 + x_2A\vec{b}_2 + \dots + x_k\vec{A}\vec{b}_k$
 $m-vector$
 $= [A\vec{b}_1, A\vec{b}_2, \dots, A\vec{b}_k]\vec{x}$
 $\int_{columns}$

This shows that SoT is the transformation induced by the m×k matrix

$$\begin{bmatrix} A \vec{b}_1 & A \vec{b}_2 & \cdots & A \vec{b}_k \end{bmatrix}$$
.

This matrix is the product of A and B:

Definition: let A be an mxn matrix and B an nxk matrix. Write $B = [\vec{b}_1 \cdots \vec{b}_k]$, where \vec{b}_i is the ite column of B. The product AB is the mxk matrix defined

$$AB = [A\overline{b}, A\overline{b}_{z} \cdots Ab_{k}].$$

i.e. the it column is the m-vector $A\vec{b}_i$.

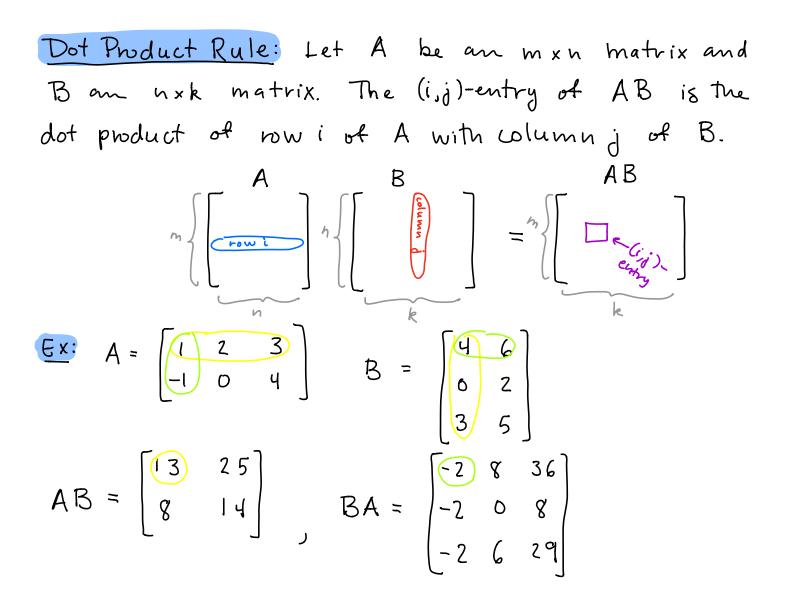
We saw above that if \vec{x} is a k-vector, we have: $A(\vec{B}\vec{x}) = (\vec{A}\vec{B})\vec{x}.$

So if T_A and T_B are the associated transformations, then $(T_A \circ T_B)(\vec{x}) = T_A(T_B(\vec{x})) = A(B\vec{x}) = (AB)\vec{x} = T_{AB}(\vec{x}).$ Thus $T_A \circ T_B = T_{AB}$

$$\underbrace{ \begin{bmatrix} x \\ Let \\ A = \begin{bmatrix} 1 & 0 & 3 \\ 2 & 2 & -1 \\ 1 & 1 & 0 \end{bmatrix} }_{3}, B = \begin{bmatrix} 2 & 3 \\ 4 & 0 \\ -1 & -2 \end{bmatrix} \right\} 3$$

$$A \begin{bmatrix} 2\\ 4\\ -1 \end{bmatrix} = \begin{bmatrix} 2\\ 4\\ 2 \end{bmatrix} + \begin{bmatrix} 0\\ 8\\ 4 \end{bmatrix} + \begin{bmatrix} -3\\ 1\\ 0 \end{bmatrix} = \begin{bmatrix} -1\\ 13\\ 6 \end{bmatrix}$$
$$A \begin{bmatrix} 3\\ 0\\ -2 \end{bmatrix} = \begin{bmatrix} 3\\ 6\\ 3 \end{bmatrix} + \begin{bmatrix} 0\\ 0\\ 0 \end{bmatrix} + \begin{bmatrix} -6\\ 2\\ 0 \end{bmatrix} = \begin{bmatrix} -3\\ 8\\ 3 \end{bmatrix}$$
$$So \quad AB = \begin{bmatrix} -1\\ -3\\ 13\\ 8\\ 6\\ 3 \end{bmatrix}$$

There is another way to compute the product of two matrices that computes each entry rather than each column:



This example shows that AB ≠ BA. That is, matrix multiplication is not commutative. It's not even commutative if A and B are both square matrices of the same size.

EX:
$$A = \begin{bmatrix} -2 & 1 \\ 3 & 0 \end{bmatrix}$$
, $B = \begin{bmatrix} 5 & -3 \\ -4 & -2 \end{bmatrix}$
Then $AB = \begin{bmatrix} -14 & 4 \\ 15 & -9 \end{bmatrix}$ and $BA = \begin{bmatrix} -19 & 5 \\ 2 & -4 \end{bmatrix}$

However, some pairs of matrices do commute with each other:

$$E \times A = \begin{bmatrix} 1 & 2 \\ 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 3 & 1 \\ 0 & 3 \end{bmatrix}$$

$$A = \begin{bmatrix} 3 & 7 \\ 0 & 3 \end{bmatrix}, \quad BA = \begin{bmatrix} 3 & 7 \\ 0 & 3 \end{bmatrix}$$

$$E \times A = \begin{bmatrix} 1 & 2 & 0 \\ -3 & 0 & 5 \end{bmatrix}, \quad T_{2} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad T_{3} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

Then $AI_3 = A = I_2 A$.

This is true in general:

If A is an mxn matrix, and I_m and I_n are the mxm and hxn identity matrices, respectively, then $AI_n = A = I_m A$.

Some properties of matrix multiplication:

let a be a scalar, and A, B, C matrices of sizes such that the given products are defined. Then:

(1)
$$A(BC) = (AB)C$$
 (associativity)
(2) $A(B+C) = AB+AC$ and $(A+B)C = AC+BC$ (distributivity)
(3) $a(AB) = (aA)B = A(aB)$.
(4) $(AB)^{T} = B^{T}A^{T}$.

Block matrices

Sometimes it is convenient to write matrices whose entries are themselves matrices, called blocks. Such a matrix is partitioned into blocks.

For example, when we write a matrix in terms of its columns, it is a block partition.

$$A = [\vec{a}_1 \dots \vec{a}_n]$$
 is a block partition of A.

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 2 & 2 & 1 & 0 & 1 \\ 3 & -1 & -1 & 1 & 0 \end{bmatrix} = \begin{bmatrix} I_2 & O_{23} \\ P & Q \end{bmatrix}$$
$$B = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ \frac{1}{7} & 2 \\ -1 & 0 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} I_2 \\ Y \\ X \end{bmatrix}$$

Εx

The product can be computed in block form:

$$AB = \begin{bmatrix} I_{2} & O_{23} \\ P & Q \end{bmatrix} \begin{bmatrix} I_{2} \\ X \end{bmatrix} = \begin{bmatrix} I^{2} + OX \\ PI + QX \end{bmatrix}$$
$$QX = \begin{bmatrix} I & 0 & I \\ -1 & I & 0 \end{bmatrix} \begin{bmatrix} 7 & 2 \\ -1 & 0 \\ I & 0 \end{bmatrix} = \begin{bmatrix} 8 & 2 \\ -8 & -2 \end{bmatrix}$$
$$So \quad AB = \begin{bmatrix} I & 0 \\ 0 & I \\ -5 & -3 \end{bmatrix}.$$

Theorem: If matrices A and B are partitioned into blocks, AB can be computed using matrix multiplication using blocks as entries, as long as the corresponding blocks being multiplied are compatible.

Practice Problems: 2.3:26,3a,6b,7,8ac,9