

Four different ways of computing the matrix product AB :

- 1 The entry in the i th row and j th column of AB , denoted $(AB)_{ij}$, is the dot product of the i th row of A with the j th column of B .
- 2 $A\vec{x} = x_1\vec{a}_1 + x_2\vec{a}_2 + \cdots + x_n\vec{a}_n$, where the vectors $\vec{a}_1, \dots, \vec{a}_n$ are the columns of A .
- 3 $AB = [A\vec{b}_1 A\vec{b}_2 \cdots A\vec{b}_n]$, where the vectors $\vec{b}_1, \dots, \vec{b}_n$ are the columns of B .
- 4 $AB = \begin{bmatrix} \vec{r}_1 B \\ \vec{r}_2 B \\ \vdots \\ \vec{r}_m B \end{bmatrix}$ where the vectors $\vec{r}_1, \dots, \vec{r}_m$ are the rows of A .

Some properties of matrix products

- 1 To multiply AB , we need the number of columns of A equal to the number of rows of B .
- 2 Most of the time $AB \neq BA$. (Matrix multiplication is not commutative.)
- 3 But $(AB)C = A(BC)$ as long as you keep them in the same order. (Matrix multiplication is associative.)

$$4 \quad \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \cdot \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix} = \begin{bmatrix} x_1 & x_2 & x_3 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$$

- 5 $A\vec{x}$ works as a matrix product if \vec{x} is a column vector.
- 6 Multiplying by a scalar is commutative: $A(c\vec{x}) = c(A\vec{x}) = (cA)\vec{x}$.
- 7 $A(\vec{x} + \vec{y}) = A\vec{x} + A\vec{y}$ and $A(B + C) = AB + AC$. (Matrix multiplication is distributive over matrix addition.)

True or False?

- 1 If AB and BA are both defined, then A and B are both square matrices.
- 2 If B has a column of zeros, then so does AB .
- 3 If A has a column of zeros, then so does AB .
- 4 If A has two rows repeated, then so does AB .
- 5 If A has two columns repeated, then so does AB .
- 6 If $AB = 0$ then $A = 0$ or $B = 0$.
- 7 If $AC = BC$ then $A = B$.

Special matrices and other matrix operations:

- A **zero matrix**
$$\begin{bmatrix} 0 & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 0 \end{bmatrix}$$

- An **identity matrix** $I_n = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$

- The **transpose** of A , A^T is formed by swapping rows and columns of A , equivalently, reflecting A across its main diagonal, equivalently, $(A^T)_{ij} = (A)_{ji}$.
- The **trace** of A , $\text{tr}(A)$ is the sum of the entries on the main diagonal of A .
- An $n \times n$ matrix A is **invertible** or **nonsingular** if there is a matrix A^{-1} (the **inverse** of A) such that $AA^{-1} = I_n$ and $A^{-1}A = I_n$.

Properties of Transposes and Inverses

- 1 $(AB)^T = B^T A^T$.
- 2 If A is invertible, then A^T is invertible, and $(A^T)^{-1} = (A^{-1})^T$.
- 3 If A is an invertible matrix, then A^{-1} is invertible, and $(A^{-1})^{-1} = A$.
- 4 If A and B are invertible $n \times n$ matrices, then AB is invertible, and

$$(AB)^{-1} = B^{-1}A^{-1}.$$

- 5 If A is an invertible $n \times n$ matrix and k is a positive integer, then A^k is invertible, and

$$(A^k)^{-1} = (A^{-1})^k.$$

- 6 If A is an invertible matrix, then the equation $A\vec{x} = \vec{b}$ has the unique solution $\vec{x} = A^{-1}\vec{b}$.