

Sec. 1.6 Algebraic properties of matrix operations

① Properties of addition

Same as for scalars. See Thm. 7.

② Properties of multiplication

Most (but not all — see topic ③) are similar to those for scalars.

Associative property holds:

$$(AB)C = A(BC) \quad \text{Thm. 8}$$

MUST READ

Distributive properties hold: see Thm. 9.

③ Differences from scalar multiplication.

(D-1) $AB \neq BA$ in general (see Ex. 4 in sec. 1.5 (notes)).

(D-2) $(AB = AC) \not\Rightarrow B = C$,
in general

even if $A \neq \mathcal{O}$ ← zero matrix.

Ex. 1 $A = \begin{pmatrix} 4 & 2 \\ 2 & 1 \end{pmatrix}$, $B = \begin{pmatrix} 12 \\ 34 \end{pmatrix}$, $C = \begin{pmatrix} 3 & 1 \\ -1 & 6 \end{pmatrix}$.

$$AB = \begin{pmatrix} 10 & 16 \\ 5 & 8 \end{pmatrix} = AC.$$

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Corollary: in general

$$(P \text{ and } Q = \mathcal{O}) \not\Rightarrow (P \text{ or } Q = \mathcal{O}).$$

Proof: starting with (D-2):

$$(AB = AC) \not\Rightarrow (B = C)$$

↓ Thm. 7

$$(AB - AC = \mathcal{O}) \not\Rightarrow (B - C = \mathcal{O})$$

↓ Thm. 9

$$\underbrace{(A)}_P \underbrace{(B-C)}_Q = \mathcal{O} \not\Rightarrow \underbrace{(B-C)}_Q = \mathcal{O}$$

and also $\underbrace{A}_P \neq \mathcal{O}$.

For an example with numbers, take $P = A$ and $Q = B - C$, with A, B, C in Ex. 1.

4) Scalar product and transpose of a matrix

Recall the dot product in Calculus:

$$\vec{a} = \langle a_1, a_2 \rangle, \vec{b} = \langle b_1, b_2 \rangle, \Rightarrow \vec{a} \cdot \vec{b} = a_1 b_1 + a_2 b_2.$$

Our notations: $\underline{a} = \begin{pmatrix} a_1 \\ a_2 \end{pmatrix}, \underline{b} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$.

Define the scalar product of \underline{a} and \underline{b} :

$$\underline{a}^T \underline{b} \equiv (a_1, a_2) \cdot \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} = a_1 b_1 + a_2 b_2$$

"transpose": make a column a row (and vice versa; see below).

In general, the scalar product of

$$\underline{x} = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \text{ and } \underline{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix} \text{ is:}$$

$$\underline{x}^T \underline{y} \equiv (x_1, x_2, \dots, x_n) \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix} = x_1 y_1 + x_2 y_2 + \dots + x_n y_n$$

" \underline{x} -transpose".

Transpose of a matrix A : A^T .

To get A^T , make each column of A a row (and vice versa - this occurs automatically).

See p. 63 in book for a rigorous definition.

Ex 2. Find the transpose of:

$$(a) A = \begin{pmatrix} \boxed{1} & \boxed{2} & \boxed{3} \\ \boxed{4} & \boxed{5} & \boxed{0} \end{pmatrix} \Rightarrow A^T = \begin{pmatrix} \boxed{1} & \boxed{4} \\ \boxed{2} & \boxed{5} \\ \boxed{3} & \boxed{0} \end{pmatrix}$$

$\boxed{2 \times 3}$ $\boxed{3 \times 2}$

$$(b) A = \begin{pmatrix} \boxed{1} & \boxed{2} \\ \boxed{3} & \boxed{4} \end{pmatrix} \Rightarrow A^T = \begin{pmatrix} \boxed{1} & \boxed{3} \\ \boxed{2} & \boxed{4} \end{pmatrix}$$

$$(c) A = \begin{pmatrix} \boxed{1} & \boxed{2} \\ \boxed{2} & \boxed{4} \end{pmatrix} \Rightarrow A^T = \begin{pmatrix} \boxed{1} & \boxed{2} \\ \boxed{2} & \boxed{4} \end{pmatrix} = A!$$

Def: Matrices such that $A^T = A$ are called symmetric.

Ex. 3 of a
symmetric
3x3 matrix

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 4 & 5 \\ 3 & 5 & 6 \end{pmatrix}$$

Thm. 10 Properties of the transpose.

1) Let A & B are $m \times n$. Then $(A+B)^T = A^T + B^T$.

2) Let A be $m \times n$ and B be $n \times p$. Then

$(AB)^T = B^T A^T$ ← **NOTE THE ORDER!**

Why does one need to "flip" the order of A & B? Check the dimensions:

$A \ B \Rightarrow (AB)^T$ is $p \times m$.
 $m \times n = n \times p$
 $m \times p$

$B^T \ A^T \Rightarrow B^T A^T$ is also $p \times m$.
 $(n \times p)^T \ (m \times n)^T$
 $p \times n = n \times m$
 $p \times m$

But $(A^T) \ (B^T)$ may be undefined!
 $n \times m \quad p \times n$

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• Special case: if r is a scalar, then $(rA)^T = rA^T$.

$$3) (A^T)^T = A.$$

Proofs are in the book.

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⑤ On your own: ← **MUST READ** !
pp. 65-67: about the identity matrix
(analogue of "1" for scalars)

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pp 67, 68: about vector norm (= "length").

⑥ Some examples of proofs

Ex. 4 Show that for any $m \times n$ matrix A ,
 $B = A^T A$ and $C = A A^T$ are always
symmetric.

Proof: 0) Make sure B & C are defined:

$B = A^T A$ is $n \times n$; similarly C is $m \times m$.
 $(m \times n)^T \quad m \times n$
 $n \times m = m \times n$
 $n \times n$ ✓

Do for B , → 1)

Have:

$$B = A^T A$$

Want:

$$B^T = B.$$

Similar
for C

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So compute $\underbrace{B^T}_{\text{l.h.s.}}$ and compare with $\underbrace{B}_{\text{r.h.s.}}$.

$$B^T = (A^T A)^T \stackrel{2)}{=} A^T (A^T)^T \stackrel{3)}{=} A^T A = B.$$

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Ex. 5 Find A, B s.t. A & B are symmetric, but AB is not.

Sol'n:

Have:

$$A^T = A, B^T = B \\ C = AB$$

Want:

$$C^T \neq C.$$

So, compute C^T and compare it with C .

$$C^T = (AB)^T \stackrel{2)}{=} B^T A^T \stackrel{\text{given}}{=} BA \\ C = AB.$$

We know that $BA \neq AB$ in general, so it should be easy to find such A, B.

For example:

$$A = \begin{pmatrix} 1 & 2 \\ 2 & 1 \end{pmatrix}, B = \begin{pmatrix} 1 & 2 \\ 2 & -1 \end{pmatrix}$$

$$AB = \begin{pmatrix} 5 & 0 \\ 4 & 3 \end{pmatrix} \leftarrow \text{not symmetric.}$$

HW: 7, 13, 15, 17, 19, 21, 26, 28, 27, 31 [30 was done in class], 35, 41(a), 43, 47, 49(b), 57.

EC #2: 46 + Sec. 1.5 #67

Note: B is $n \times n$, not $m \times n$, as book says.

Hint for 41(a): write this as a l.s.

for 43(b): Use Thm. 8 repeatedly. First do the problem for $A^2 u$, then move on to $A^5 u$.