

Matrix operations &

Manipulating matrices

Beezer: p 125 - 147 print version

Strang: Sec 1.4 & 1.6.

Definition: M_{mn} is the set of all $m \times n$ matrices.

Matrix Equality "what does it mean to say that two matrices are equal?"

Definition: We say $A = B$ if

$$\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{bmatrix} \quad \begin{bmatrix} b_{11} & \cdots & b_{1n} \\ \vdots & \ddots & \vdots \\ b_{m1} & \cdots & b_{mn} \end{bmatrix}$$

$$a_{11} = b_{11}, \quad a_{12} = b_{12}, \quad \dots, \quad a_{ij} = b_{ij} \text{ for all } i, j.$$

$$\begin{bmatrix} a_{m1} & \dots & a_{mn} \end{bmatrix} \quad \begin{bmatrix} a_{m1} & & a_{mn} \end{bmatrix}$$

i.e. $A = B$ if $[A]_{ij} = [B]_{ij}$ for all i, j

Ex: $A = \begin{bmatrix} 3 & 2 \\ 1 & 1 \end{bmatrix} \quad B = \begin{bmatrix} 3 & 2 \\ 1 & 2 \end{bmatrix}$

$A \neq B$ because $[A]_{22} = 1 \neq 2 = [B]_{22}$

Definition: Matrix addition: suppose A, B are $m \times n$ matrices.

$A + B$ is the $m \times n$ matrix with entries

$$[a_{ij}] - [a_{ij}] + [b_{ij}]$$

$$[H+I]_{ij} = L^{(i)}_{ij} - \dots$$

Ex: $A = \begin{bmatrix} 1 & 1 \\ 2 & 0 \end{bmatrix}$ $B = \begin{bmatrix} 0 & 1 \\ 1 & 3 \end{bmatrix}$

$$A+B = \begin{bmatrix} 1+0 & 1+1 \\ 2+1 & 0+3 \end{bmatrix} = \begin{bmatrix} 1 & 2 \\ 3 & 3 \end{bmatrix}$$

Note: we can only add matrices of the same size!

Ex: If $A = \begin{bmatrix} 2 & 2 \\ 3 & 1 \\ 2 & 1 \end{bmatrix}$ $B = \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}$,

$A+B$ is not defined.

Next... Scalar multiplication.

Definition:
Let α be a scalar (i.e. a real number). We write
this as $\alpha \in \mathbb{R}$. Let $A \in M_{mn}$ be an $m \times n$ matrix.

Then αA is also an $m \times n$ matrix with

$$[\alpha A]_{ij} = \alpha [A]_{ij}.$$

Example: $\alpha = 2$ $A = \begin{bmatrix} 3 & 3 & 1 \\ 2 & 0 & 1 \end{bmatrix}$

$$\alpha A = \begin{bmatrix} 2 \times 3 & 2 \times 3 & 2 \times 1 \\ 2 \times 2 & 2 \times 0 & 2 \times 1 \end{bmatrix} = \begin{bmatrix} 6 & 6 & 2 \\ 4 & 0 & 2 \end{bmatrix}$$

Combine addition & scalar multiplication:

Let A, B be $m \times n$ matrices & let $\alpha \in \mathbb{R}$.

$A + \underline{\alpha B}$: We'll get back to this shortly.

Recall that there is a real number (0) that

satisfies $0 + r = r$ for all $r \in \mathbb{R}$.

Definition: the $m \times n$ zero matrix 0 or $0_{m \times n}$
is the matrix all of whose entries are

zero: $[0]_{ij} = 0$ for all i, j .

$\begin{bmatrix} 0 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 0 & 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 0 & 0 & 0 & 0 \end{bmatrix} \quad \text{etc}$

$$\text{Ex: } \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \text{ is } \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

Definition: The additive inverse of $A \in M_{mn}$ is the matrix $-A$ with entries

$$[-A]_{ij} = -[A]_{ij}.$$

$$\text{Ex: } -\begin{bmatrix} 3 & 2 \\ 2 & 1 \end{bmatrix} = \begin{bmatrix} -3 & -2 \\ -2 & -1 \end{bmatrix}.$$

Note: $-A = \underbrace{(-1)A}_{\text{Scalar multiple of } A \text{ by } \alpha = (-1)}.$

Some properties satisfied by these operations.

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

1. $A + B = B + A$ commutativity

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

3. $A + \mathbb{O} = A$

4. $A + (-A) = \mathbb{O}$.

Check properties 1 & 4 in the next 2 minutes.

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

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

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

$$8. \quad 1 \cdot A = A.$$

Check properties 5 8 7 in the next 2 minutes.

Let's prove property 7, to show how most of these properties can be checked.

Want to show: $\alpha(A + B) = \underbrace{\alpha A + \alpha B}_{m \times n \text{ matrix}} \quad m \times n \text{ matrix}$

need to show that each entry $[\alpha(A + B)]_{ij}$ equals the corresponding entry $[\alpha A + \alpha B]_{ij}$.

$$[\alpha(A + B)]_{..} = \alpha[A + B]_{..} \quad \text{by def. of}$$

L L L L scalar mult.

$$= \alpha([A]_{ij} + [B]_{ij})$$

by def. of matrix addition.


$$= \alpha[A]_{ij} + \alpha[B]_{ij}$$

by distributivity of mult. of real numbers

$$= [\alpha A]_{ij} + [\alpha B]_{ij}$$

by def of scalar mult.

$$= [\alpha A + \alpha B]_{ij}$$

by def of addition of matrices.

□  symbol means our proof is complete.

Notice how we used distributivity of multiplication of real numbers to show distributivity of scalar mult. of matrices.

multiplication ~ ~ ~

Definition: let $A \in M_{m \times n}$. The Transpose of A , written A^t , is the $n \times m$ matrix with

$$[A^t]_{ij} = [A]_{ji}.$$

(in other words, transposing A swaps its rows & columns)

Ex: $A = \begin{bmatrix} 2 & 2 & 3 \\ 0 & 0 & -1 \end{bmatrix} \in M_{2 \times 3}$

$A^t =$

$$F = \begin{bmatrix} -1 & 0 \\ 2 & 0 \\ 3 & -1 \end{bmatrix}$$

Question: Prove that $A^t + B^t = (A+B)^t$.

Answer: Need to show that

$$[A^t + B^t]_{ij} = [(A+B)^t]_{ij}$$

for all $1 \leq i \leq m$
 $1 \leq j \leq n$.

$$[A^t + B^t]_{ij} = [A^t]_{ij} + [B^t]_{ij} \quad (\text{def of matrix add})$$

$$= [A]_{ii} + [B]_{ii} \quad (\text{def of transpose})$$

$$= [A + B]_{ji} \quad (\text{def of matrix add})$$

$$= [(A + B)^t]_{ij} \quad (\text{def of transpose})$$

□.

Theorem: Let $\alpha \in \mathbb{R}$ & let $A \in M_{nm}$

$$(\alpha A)^t = \alpha A^t.$$

Proof: exercise.

Theorem: $(A^t)^t = A$ for any $A \in M_{mn}$.

Proof: $\left[(A^t)^t \right]_{ij} = \left[A^t \right]_{ji}$ (by def of transpose)

$$= \left[A \right]_{ij} \quad (\text{by def of transpose})$$

□

Definitions Matrix - vector product.

Let $A \in M_{m \times n}$ with columns $\vec{A}_1, \dots, \vec{A}_n$.

Let \vec{u} be a vector of size n . Then

$$A \vec{u} = [\vec{u}]_1 \vec{A}_1 + [\vec{u}]_2 \vec{A}_2 + \dots + [\vec{u}]_n \vec{A}_n.$$

by definition

Ex: $A = \begin{bmatrix} 3 & 2 & 2 \\ 0 & -1 & 2 \end{bmatrix}$

$\brace{m=2 \text{ rows}}$
 $\brace{n=3 \text{ columns}}$

$$\vec{A}_1 = \begin{bmatrix} 3 \\ 0 \end{bmatrix} \quad \vec{A}_2 = \begin{bmatrix} 2 \\ -1 \end{bmatrix} \quad \vec{A}_3 = \begin{bmatrix} 2 \\ 2 \end{bmatrix}.$$

$$\vec{u} = \begin{bmatrix} 0 \\ 1 \\ 2 \end{bmatrix}$$

Then $A\vec{u} = [\vec{u}]_1 \vec{A}_1 + [\vec{u}]_2 \vec{A}_2 + [\vec{u}]_3 \vec{A}_3$

$$= 0 \cdot \begin{bmatrix} 3 \\ 0 \end{bmatrix} + 1 \cdot \begin{bmatrix} 2 \\ -1 \end{bmatrix} + 2 \cdot \begin{bmatrix} 2 \\ 2 \end{bmatrix}$$

used
scalar
mult

$$\begin{aligned} &= \begin{bmatrix} 0 \\ 0 \end{bmatrix} + \begin{bmatrix} 2 \\ -1 \end{bmatrix} + \begin{bmatrix} 4 \\ 4 \end{bmatrix} \\ &= \begin{bmatrix} 0+2+4 \\ 0-1+4 \end{bmatrix} = \begin{bmatrix} 6 \\ 3 \end{bmatrix}. \end{aligned}$$

Notice: If $A \in M_{m \times n}$, then \vec{u} needs to be a vector with n entries and $A\vec{u}$ will have m entries.

Theorem (Systems of linear equations as Matrix Multiplication).

Consider the sys of lin. eq.
 $\begin{array}{l} \dots \\ \dots \\ \dots \end{array} \quad x_1 = b_1$

$$a_{11}x_1 + \dots + a_{1n}x_n$$

⋮

$$a_{m1}x_1 + \dots + a_{mn}x_n = b_m.$$

Let $A = \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix}$ $\vec{b} = \begin{bmatrix} b_1 \\ \vdots \\ b_m \end{bmatrix}$ $\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$

Then the set of solutions x_1, \dots, x_n of our system
equals the set of solutions \vec{x} of the matrix equation

$$A \cdot \vec{x} = \vec{b}.$$

(i.e. if x_1, \dots, x_n solves our system of linear eq,
then $\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$ solve our matrix equation,

$\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$
& vice-versa).

Ex: Solutions to $3x_1 + 2x_2 = 0$
 $5x_1 + 7x_2 = 3$

are the same as vector solutions to

$$\begin{bmatrix} 3 & 2 \\ 5 & 7 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 3 \end{bmatrix}.$$

Definition: Matrix multiplication.

Let $A \in M_{m \times n}$. Let $B \in M_{n \times p}$.

Then $AB \in M_{m \times p}$ is the matrix

with entries

$$[AB]_{ij} = \sum_{k=1}^m a_{ik} b_{kj}.$$

We can reformulate this definition as follows:

Let B have columns $\vec{B}_1, \dots, \vec{B}_p$. Then

AB is the $m \times p$ matrix with columns

$$\underline{\vec{A}\vec{B}_1, \vec{A}\vec{B}_2, \dots, \vec{A}\vec{B}_p}$$

Ex: $A = \begin{bmatrix} 3 & 2 \\ 1 & 1 \end{bmatrix}$ $B = \begin{bmatrix} 2 & 5 \\ 2 & 1 \end{bmatrix}$

$$\vec{B}_1 = [2] \quad \vec{B}_2 = [5]$$

-' [2] ~ L ~

$$A\bar{B}_1 = 2 \begin{bmatrix} 3 \\ 1 \end{bmatrix} + 2 \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 2 \cdot 3 + 2 \cdot 2 \\ 2 \cdot 1 + 2 \cdot 1 \end{bmatrix} = \begin{bmatrix} 10 \\ 4 \end{bmatrix}$$

$$A\bar{B}_2 = 5 \begin{bmatrix} 3 \\ 1 \end{bmatrix} + 1 \cdot \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \cdot 3 + 1 \cdot 2 \\ 5 \cdot 1 + 1 \cdot 1 \end{bmatrix} = \begin{bmatrix} 17 \\ 6 \end{bmatrix}$$

$$AB = \begin{bmatrix} 10 & 17 \\ 4 & 6 \end{bmatrix}.$$

Note: $AB = \begin{bmatrix} 2 \cdot 3 + 2 \cdot 2 & 5 \cdot 3 + 1 \cdot 2 \\ 2 \cdot 1 + 2 \cdot 1 & 5 \cdot 1 + 1 \cdot 1 \end{bmatrix}$ as in first version
of our definition.

Note: In order that AB be defined,
need the number of columns of A

to equal number of rows of B.

i.e. $A \in M_{m \times n}$

$B \in M_{n \times p}$

↑
columns

↓
rows

Note: you might expect matrix multiplication to

be defined by $[AB]_{ij} = [A]_{ij} \cdot [B]_{ij}$.

This operation \uparrow turns out to be mostly
useless!

Exercise: let $A = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$ $B = \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix}$.

Do ... have an equality

AB = BA ?

$$\begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 2 \cdot 3 + 1 \cdot 0 & 2 \cdot 0 + 1 \cdot (-1) \\ 1 \cdot 3 + 1 \cdot 0 & 1 \cdot 0 + 1 \cdot (-1) \end{bmatrix} = \begin{bmatrix} 6 & -1 \\ 3 & -1 \end{bmatrix}$$

$$\begin{bmatrix} 3 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} = \begin{bmatrix} 3 \cdot 2 + 0 \cdot 1 & 3 \cdot 1 + 0 \cdot 1 \\ 0 \cdot 2 - 1 \cdot 1 & 0 \cdot 1 - 1 \cdot 1 \end{bmatrix} = \begin{bmatrix} 6 & 3 \\ -1 & -1 \end{bmatrix}$$

hence $AB \neq BA$.