

HW 4 is up, due Thurs Oct 15.

Last time: Defined the notation

$$“A\vec{x} = \vec{b}”$$

as follows: Let A be $m \times n$ matrix

$$A = m \left\{ \begin{pmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{pmatrix} \right. \underbrace{\quad}_{n}$$

$$= \left(\vec{c}_1 \vec{c}_2 \cdots \vec{c}_n \right)$$

where $\vec{c}_1, \vec{c}_2, \dots, \vec{c}_n \in \mathbb{R}^m$ are
the column vectors of A . Then
for any $n \times 1$ column vector $\vec{x} \in \mathbb{R}^n$
we define the $m \times 1$ column $A\vec{x} \in \mathbb{R}^m$:

$$A\vec{x} := x_1 \vec{c}_1 + x_2 \vec{c}_2 + \cdots + x_n \vec{c}_n \in \mathbb{R}^m$$

maybe the most important definition
in the course ...

Alternatively, let

$$A = \begin{pmatrix} \vec{r}_1^T \\ \vdots \\ \vec{r}_m^T \end{pmatrix}$$

where $\vec{r}_1, \dots, \vec{r}_m \in \mathbb{R}^n$ are the
"row vectors" of the matrix A .

[Convention: \vec{v} always denotes a
column, \vec{v}^T is always a row.]

Then for any $\vec{x} \in \mathbb{R}^n$ we have

$$A\vec{x} = \begin{pmatrix} \vec{r}_1^T \vec{x} \\ \vdots \\ \vec{r}_m^T \vec{x} \end{pmatrix}$$

where $\vec{r}_i^T \vec{x} = (\text{row})(\text{column})$ is the dot product $\vec{r}_i \cdot \vec{x}$.

Example : Express the following system as matrix equation $A\vec{x} = \vec{b}$:

$$\begin{cases} x + y + z = 5, \\ x + 2y + 4z = 3. \end{cases}$$

Solution :

$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 4 \end{pmatrix}, \vec{x} = \begin{pmatrix} x \\ y \\ z \end{pmatrix}, \vec{b} = \begin{pmatrix} 5 \\ 3 \end{pmatrix}$$

Compute $A\vec{x}$ in terms of rows :

$$(1 \ 1 \ 1) \begin{pmatrix} x \\ y \\ z \end{pmatrix} = x + y + z$$

$$(1 \ 2 \ 4) \begin{pmatrix} x \\ y \\ z \end{pmatrix} = x + 2y + 4z$$

Hence

$$A\vec{x} = \begin{pmatrix} x+y+z \\ x+2y+4z \end{pmatrix}.$$

Or compute $A\vec{x}$ in terms of cols:

$$A\vec{x} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 4 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix}$$

$$= x \begin{pmatrix} 1 \\ 1 \end{pmatrix} + y \begin{pmatrix} 1 \\ 2 \end{pmatrix} + z \begin{pmatrix} 1 \\ 4 \end{pmatrix}$$

$$= \begin{pmatrix} x \\ x \end{pmatrix} + \begin{pmatrix} y \\ 2y \end{pmatrix} + \begin{pmatrix} z \\ 4z \end{pmatrix}$$

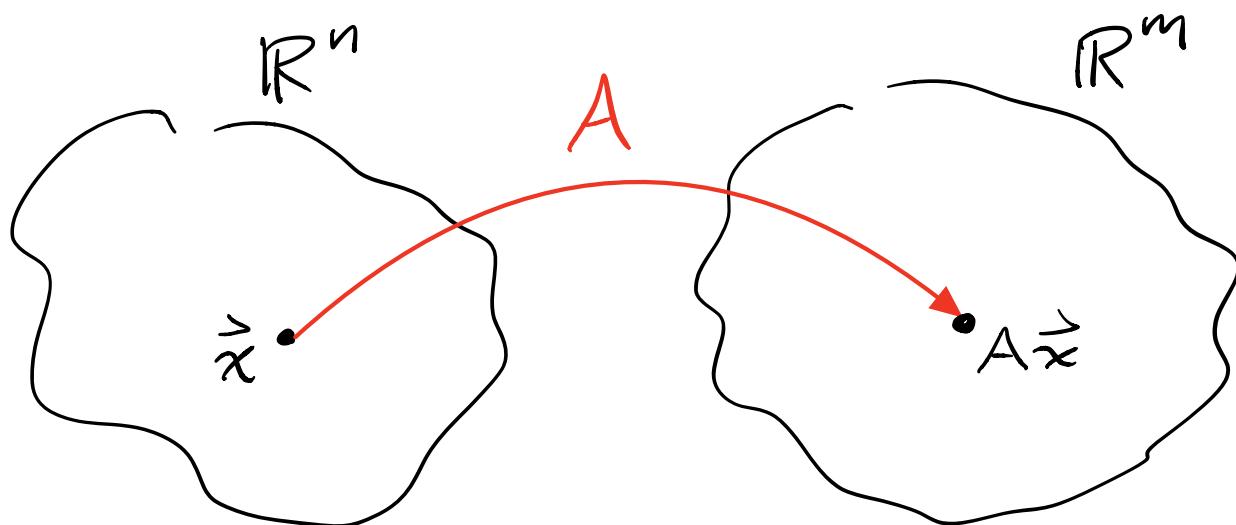
$$= \begin{pmatrix} x+y+z \\ x+2y+4z \end{pmatrix} \quad \text{SAME } \checkmark$$

Hence $A\vec{x} = \vec{b}$ says:

$$\begin{pmatrix} x+y+z \\ x+2y+4z \end{pmatrix} = \begin{pmatrix} 5 \\ 3 \end{pmatrix} \quad \checkmark$$

H
Today: A new point of view.

Given an $m \times n$ matrix A , we will think of this as a **function** from n -dimensional space to m -dimensional space:



This function has a very specific property called "linearity":

For all $\vec{u}, \vec{v} \in \mathbb{R}^n$ & $s, t \in \mathbb{R}$,

$$A(s\vec{u} + t\vec{v}) = sA\vec{u} + tA\vec{v}$$

"preserves linear combinations"

We can also express this property in two steps:

- $A(\vec{u} + \vec{v}) = A\vec{u} + A\vec{v}$

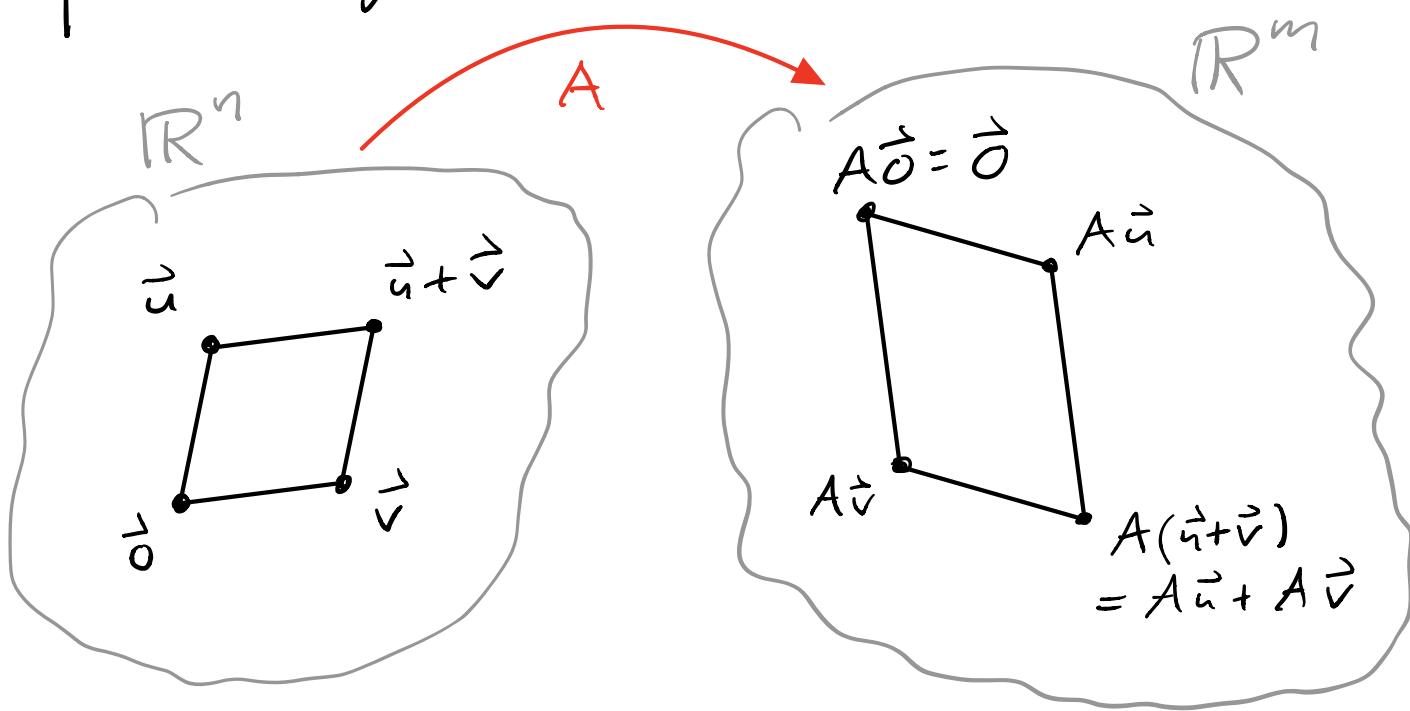
"preserves vector addition"

- $A(t\vec{v}) = tA\vec{v}$

"preserves scalar multiplication"

[You will prove this on HW 4.1.]

Geometric Meaning: A linear function sends parallelograms to parallelograms:



+
So what? Why do we care?

① Because every linear function
comes from a matrix, and this
tells us how to

"multiply matrices"

② Many basic geometric functions
are linear, e.g. rotations,
reflections, projections, ...

This is important in physics &
computer graphics, but also
(surprisingly) in probability &
statistics. [Least Squares Approx.]

③ We can also think of linear
functions between exotic kinds
of vector spaces $f: U \rightarrow V$,

e.g., derivatives & integrals,

$$\frac{d}{dx} (af(x) + bg(x)) = a \frac{d}{dx} f(x) + b \frac{d}{dx} g(x)$$

$$\int (af(x) + bg(x)) dx = a \int f(x) dx + b \int g(x) dx.$$



① Linear functions $\mathbb{R}^n \rightarrow \mathbb{R}^m$ = $m \times n$ matrices.

We've already seen that $m \times n$ matrix A determines a linear function $A: \mathbb{R}^n \rightarrow \mathbb{R}^m$.

Conversely, let $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ be any function satisfying

$$f(s\vec{u} + t\vec{v}) = sf(\vec{u}) + tf(\vec{v}).$$

Goal: Find some $m \times n$ matrix A such that $f(\vec{x}) = A\vec{x}$ for all \vec{x} .

Solution: Let $\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n \in \mathbb{R}^n$
 be the "standard basis vectors"
 $\vec{e}_1 = \begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}, \vec{e}_2 = \begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}, \dots, \vec{e}_n = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}$

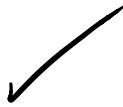
Applying f to these gives some
 vectors $f(\vec{e}_1), \dots, f(\vec{e}_n) \in \mathbb{R}^m$.

Define the matrix

$$A = \left(\underbrace{\begin{matrix} f(\vec{e}_1) & \cdots & f(\vec{e}_n) \end{matrix}}_n \right) \}^m$$

Then one can check [but not right now] that

$$f(\vec{x}) = A\vec{x} \text{ for all } \vec{x} \in \mathbb{R}^n,$$

as desired. 

↗

Example : Let $\text{id} : \mathbb{R}^n \rightarrow \mathbb{R}^n$ be
the "identity function"

$$\text{id}(\vec{x}) = \vec{x}$$

The corresponding matrix is called
the $n \times n$ "identity matrix":

$$\underline{I} = \left(\begin{array}{cccc} \text{id}(\vec{e}_1) & \text{id}(\vec{e}_2) & \dots & \text{id}(\vec{e}_n) \end{array} \right)$$

$$= \left(\vec{e}_1 \ \vec{e}_2 \ \dots \ \vec{e}_n \right)$$

$$= \left(\begin{array}{ccccc} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & & \vdots \\ \vdots & 0 & 1 & & \\ \vdots & \vdots & \vdots & 1 & 0 \\ 0 & 0 & \dots & 0 & 1 \end{array} \right)$$

$$= \left(\begin{array}{ccccc} 1 & & & & \\ & 1 & & & \\ & & \ddots & & \\ & & & 1 & \end{array} \right) \quad \text{"}$$

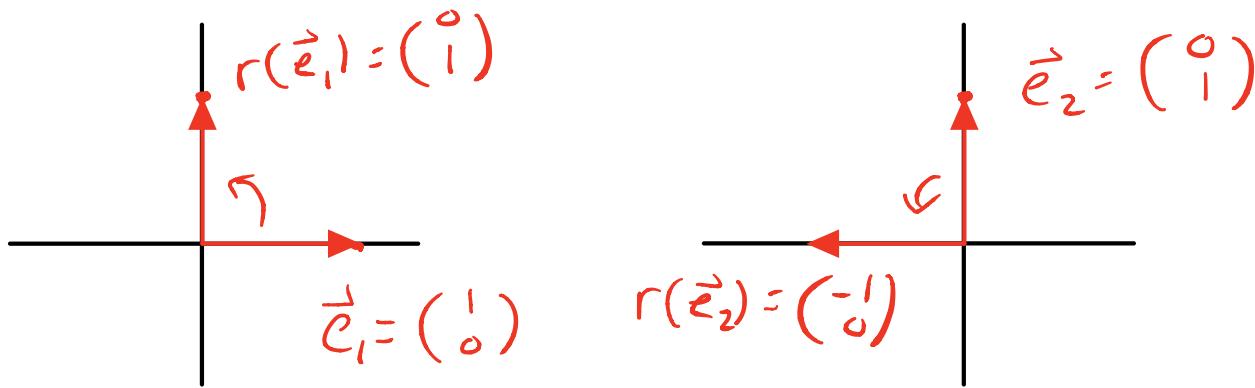
Check : $I\vec{x} = \vec{x}$ for all \vec{x} .

Example : Let $r: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be the function that rotates by 90° counterclockwise, which is a linear function (sends parallelograms to parallelograms). Find the 2×2 matrix $R = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ such that

$$\begin{aligned} r \begin{pmatrix} x \\ y \end{pmatrix} &= \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \\ &= \begin{pmatrix} ax + by \\ cx + dy \end{pmatrix}. \end{aligned}$$

How ? We apply the function r to the standard basis vectors

$$\vec{e}_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \text{ & } \vec{e}_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}.$$



So now we know the matrix. Let's call it R :

$$R = \begin{pmatrix} r(\vec{e}_1) & r(\vec{e}_2) \end{pmatrix} = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}.$$

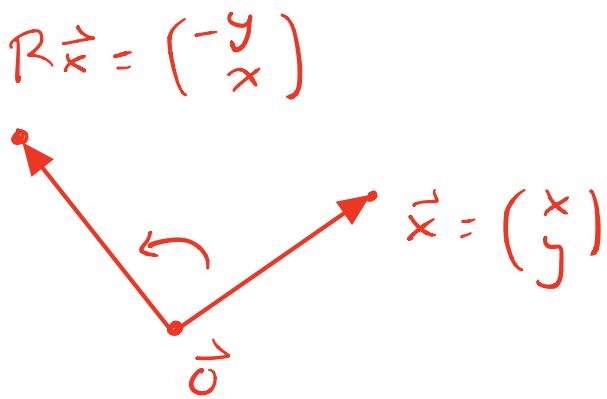
Conclusion: To rotate a general vector $\vec{x} = (x, y)$ multiply by R :

$$r(\vec{x}) = R \vec{x}$$

$$= \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix}$$

$$= x \begin{pmatrix} 0 \\ 1 \end{pmatrix} + y \begin{pmatrix} -1 \\ 0 \end{pmatrix} = \begin{pmatrix} -y \\ x \end{pmatrix}$$

Picture :



What if we rotate twice ?

$$r^2(\vec{x}) = (r \circ r)(\vec{x}) = r(r(\vec{x}))$$

"rotate by 180° "

To get the matrix :

$$r^2 \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \quad \& \quad r^2 \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ -1 \end{pmatrix}$$

We call this matrix

$$"R^2" = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}$$

This is an example of
matrix multiplication !

$$R \cdot R = R^2$$

$$\begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} -1 & 0 \\ 0 & -1 \end{pmatrix}$$

rotate twice by $=$ rotate by
 90° c.c.w. 180°

What about R^3 & R^4 ?

One can check that

$$R^3 = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} = \text{rotate by } 90^\circ \text{ clockwise}$$

$$R^4 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \text{do nothing}$$

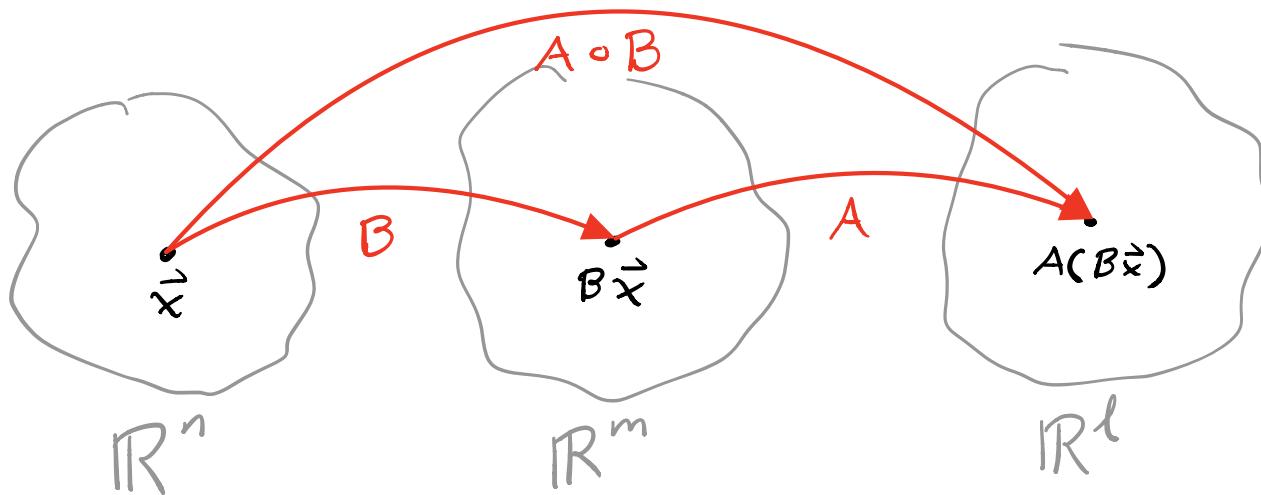
$= I$ the 2×2 identity



How does matrix multiplication
work in general ?

Let A be $l \times m$ matrix
 B be $m \times n$ matrix.

Think of these as linear functions:



The composite function

$$A \circ B : \mathbb{R}^n \rightarrow \mathbb{R}^l$$

is also linear, so it corresponds to some $l \times n$ matrix. We call this matrix AB "A times B."

In other words, we define the matrix AB so that

$$A(B\vec{x}) = (AB)\vec{x}$$

[That seems reasonable.]

How to compute it?

Recall : For standard basis

$\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n \in \mathbb{R}^n$, we have

$B\vec{e}_j = j^{\text{th}}$ column of B .

$(AB)\vec{e}_j = j^{\text{th}}$ column of AB .

Hence

$$\begin{aligned} \text{jth col } AB &= (AB)\vec{e}_j \\ &= A(B\vec{e}_j) \\ &= A(\text{jth col of } B). \end{aligned}$$

That's it!



Example : $A = \begin{pmatrix} 1 & 0 & 2 \\ 1 & 1 & 1 \end{pmatrix}$, $B = \begin{pmatrix} 1 & -1 \\ 0 & 1 \\ 2 & 0 \end{pmatrix}$

Compute AB & BA .

AB :

$$1\text{st col } AB = A(1\text{st col } B)$$

$$= \begin{pmatrix} 1 & 0 & 2 \\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 2 \end{pmatrix} = \begin{pmatrix} 1+0+4 \\ -1+0+2 \end{pmatrix} = \begin{pmatrix} 5 \\ 3 \end{pmatrix}$$

$$2\text{nd col } AB = A(2\text{nd col } B)$$

$$= \begin{pmatrix} 1 & 0 & 2 \\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} -1+0+0 \\ -1+1+0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \end{pmatrix}$$

Hence

$$AB = \begin{pmatrix} 1 & 0 & 2 \\ -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 0 & 1 \\ 2 & 0 \end{pmatrix} = \begin{pmatrix} 5 & -1 \\ 3 & 0 \end{pmatrix}$$

Next BA :

$$1\text{st col } BA = B(1\text{st col } A)$$

$$= \begin{pmatrix} 1 & -1 \\ 0 & 1 \\ 2 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & -1 \\ 0 & 1 \\ 2 & 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 2 \end{pmatrix}$$

$$\text{2nd col } BA = B \text{ (2nd col A)}$$

$$= \begin{pmatrix} \cancel{1} & \cancel{-1} \\ \cancel{0} & \cancel{1} \\ \cancel{2} & \cancel{0} \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$= \begin{pmatrix} 0 & -1 \\ 0 & 1 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 1 \\ 0 \end{pmatrix}$$

$$\text{3rd col } BA = B \text{ (3rd col A)}$$

$$= \begin{pmatrix} \cancel{1} & \cancel{-1} \\ \cancel{0} & \cancel{1} \\ \cancel{2} & \cancel{0} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$= \begin{pmatrix} 2 & -1 \\ 0 & 1 \\ 4 & 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 1 \\ 4 \end{pmatrix}.$$

$$\text{Hence, } BA = \begin{pmatrix} 0 & -1 & 1 \\ 1 & 1 & 1 \\ 2 & 0 & 4 \end{pmatrix}.$$

Important Observation :

$$AB \neq BA$$

$$\begin{pmatrix} 5 & -1 \\ 3 & 0 \end{pmatrix} \neq \begin{pmatrix} 0 & -1 & 1 \\ 1 & 1 & 1 \\ 2 & 0 & 4 \end{pmatrix}$$

These matrices don't even have
the same shape!