

Quiz 3 solutions are up.

Due to medical issues, HW 4 is delayed until Thurs Oct 22.

This means all HW and Quizzes are delayed 1 week.



Recall: A function $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is called "linear" if

- it preserves vector addition:

$$f(\vec{u} + \vec{v}) = f(\vec{u}) + f(\vec{v})$$

- it preserves scalar multiplication:

$$f(t\vec{v}) = t f(\vec{v}).$$

[In particular, taking $t=0$ gives

$$f(\vec{0}) = \vec{0}.]$$

We can also summarize this definition in one step:

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

" f preserves linear combinations"

Geometric Meaning: A linear function sends $\vec{0}$ to $\vec{0}$ and sends parallelograms to parallelograms.

Theorem: If $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is linear, then there exists an $m \times n$ matrix $[f]$ such that

$$f(\vec{x}) = [f] \vec{x} \quad \text{for all } \vec{x} \in \mathbb{R}^n$$

matrix times column vector.

Proof: Let $\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n \in \mathbb{R}^n$ be the "standard basis vectors":

$$\vec{e}_i = (0, \dots, 0, 1, 0, \dots, 0).$$

i th position.

So for any vector $\vec{x} = (x_1, x_2, \dots, x_n)$ we observe that

$$\vec{x} = x_1 \vec{e}_1 + x_2 \vec{e}_2 + \dots + x_n \vec{e}_n$$

$$\begin{aligned}
 &= x_1(1, 0, \dots, 0) \\
 &\quad + x_2(0, 1, 0, \dots, 0) \\
 &\quad + \dots + x_n(0, \dots, 0, 1) \quad \checkmark
 \end{aligned}$$

Let $[f]$ be the $m \times n$ matrix whose j th column vector is $f(\vec{e}_j) \in \mathbb{R}^m$:

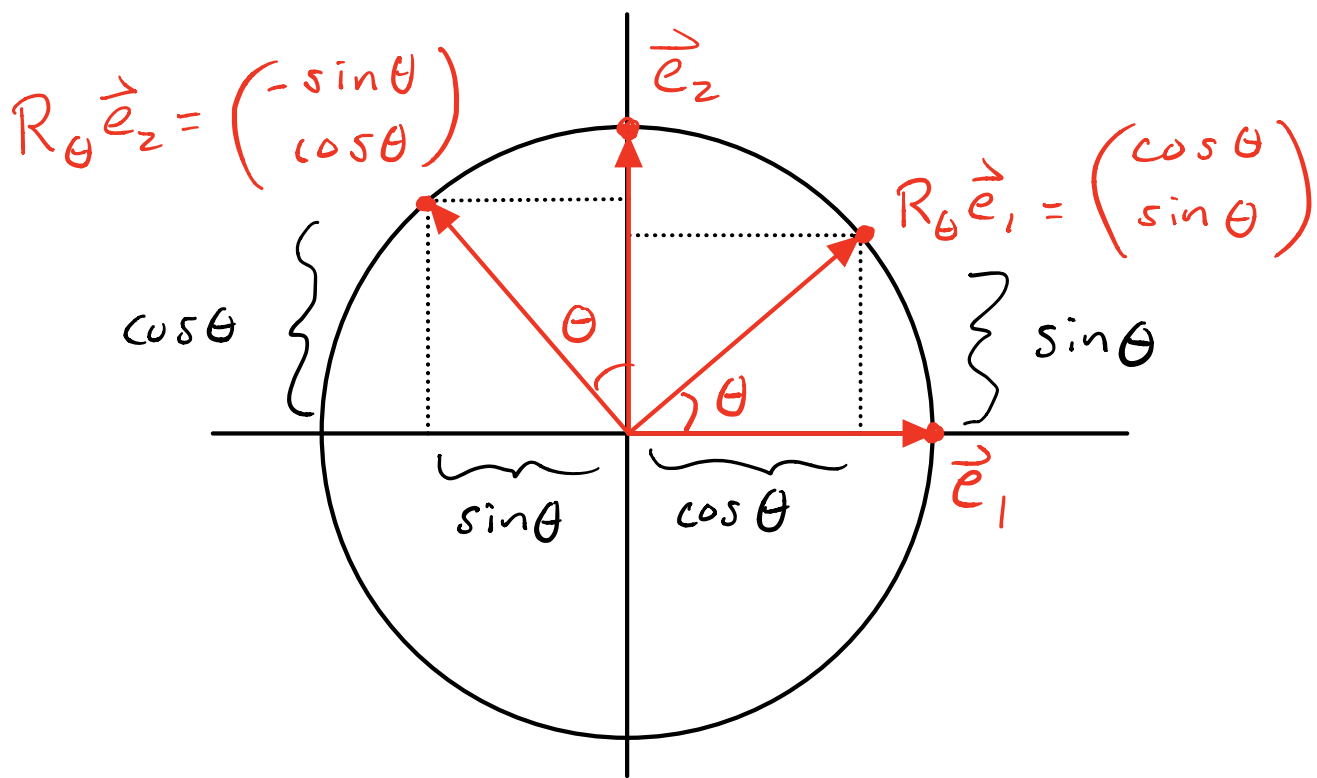
$$[f] = \underbrace{\left(f(\vec{e}_1) \ \dots \ f(\vec{e}_n) \right)}_n \Bigg\}^m$$

Then by the definition of linearity & matrix times column vector, we have

$$\begin{aligned}
 [f] \vec{x} &= \left(f(\vec{e}_1) \ \dots \ f(\vec{e}_n) \right) \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \\
 &= x_1 f(\vec{e}_1) + \dots + x_n f(\vec{e}_n) \quad \text{def} \\
 &= f(x_1 \vec{e}_1 + \dots + x_n \vec{e}_n) \quad \text{def} \\
 &= f(\vec{x}) \quad \text{as desired.} \quad \checkmark
 \end{aligned}$$

Example: Let $R_\theta: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be the function that rotates by angle θ counter-clockwise around the origin, which is a linear function.

Let's find the matrix:



It follows that the matrix is

$$R_\theta = \left(R_\theta \vec{e}_1 \mid R_\theta \vec{e}_2 \right)$$
$$= \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}.$$

To rotate a general point $\vec{x} = (x, y)$
we multiply by the matrix:

$$\begin{aligned} R_{\theta} \vec{x} &= \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \\ &= \begin{pmatrix} x \cos \theta - y \sin \theta \\ x \sin \theta + y \cos \theta \end{pmatrix} \end{aligned}$$

That's pretty good!

Example from last time:

R_{90° = rotate c.c.w. by 90°

$$= \begin{pmatrix} \cos 90^\circ & -\sin 90^\circ \\ \sin 90^\circ & \cos 90^\circ \end{pmatrix}$$

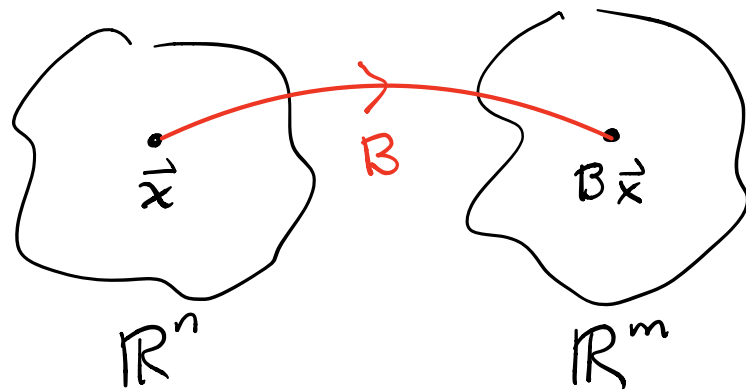
$$= \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \quad \checkmark$$



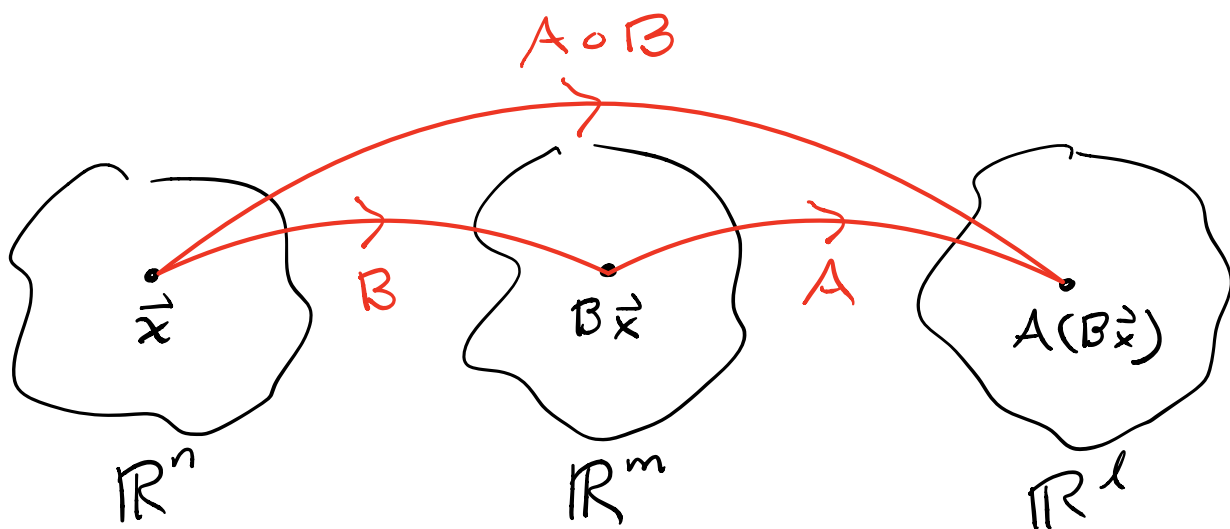
So any linear function is a matrix.

Conversely, any $m \times n$ matrix B

can be thought of as a function
from $\mathbb{R}^n \rightarrow \mathbb{R}^m$:



And you will show that this function
is linear. If $A : \mathbb{R}^m \rightarrow \mathbb{R}^l$ is
any $l \times m$ matrix, then we consider
the composite function $A \circ B : \mathbb{R}^n \rightarrow \mathbb{R}^l$:



The function $A \circ B$ is also linear,
so it corresponds to an $l \times n$ matrix:

$$[A \circ B] = \left(\underbrace{A(B\vec{e}_1) \dots A(B\vec{e}_n)}_n \right) \Bigg\} l$$

Important Definition:

If A is $l \times m$ matrix & B is $m \times n$ then we define $l \times n$ matrix AB as the matrix of the composite function:

$$AB := [A \circ B]$$

= the matrix of the linear function $A \circ B$.

In other words, we have

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

for all vectors $\vec{x} \in \mathbb{R}^n$.

Fine. But how do we compute this matrix?

Well, we know that

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

which is already defined. ☺

It follows that

$$\begin{aligned} i, j \text{ entry of } AB &= i\text{th entry of } A(\text{jth col } B) \\ &= (\text{i\text{th row } A})(\text{jth col } B) \end{aligned}$$

↑
dot product.

Examples :

• Let $A = \begin{pmatrix} 1 & 2 & 0 \\ 2 & 3 & 1 \\ 1 & 0 & 1 \end{pmatrix}$, $B = \begin{pmatrix} 2 & -1 \\ 0 & 1 \\ 1 & 2 \end{pmatrix}$

Compute AB :

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

But note that BA is not defined.

Reason: # cols $B \neq$ # rows A .

In other words, the function

$B \circ A$ is not defined because

$$A: \mathbb{R}^3 \rightarrow \mathbb{R}^3$$

$$B: \mathbb{R}^2 \rightarrow \mathbb{R}^3$$

Emphasis: IF A is $k \times l$
 B is $m \times n$

then AB is only defined when

$$l = m$$

cols $A =$ # rows B .

In which case, AB is $k \times n$.

$$\begin{array}{ccc} A \cdot B & = & AB \\ k \times \boxed{l} \quad \boxed{m} \times n & & k \times n \\ \text{these} & & \\ \text{must} & & \\ \text{match} & & \end{array}$$

• Let $R = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}$
= rotate 90° c.c.w.

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

= reflect across line $x=y$.

Compute RF & FR :

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

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

Observe $RF \neq FR$.

Geometrically:

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

"reflect across line $x=0$ "

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

"reflect across line $y=0$ "

We conclude that

"reflect across $x=y$ and then
rotate ccw 90° "

$$= RF = \begin{pmatrix} -1 & 0 \\ 0 & 1 \end{pmatrix} = \text{"reflect across } x=0 \text{"}$$

"rotate ccw 90° and then reflect across line $x=y$ "

$$= FR = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = \text{"reflect across"} \\ y=0$$

These geometric facts are not so obvious, but the algebra was easy to compute. Slogan:

"Algebra is smarter than geometry"



Next Topic: Inverse Matrices.

$$\text{Let } f: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

$$g: \mathbb{R}^m \rightarrow \mathbb{R}^n \text{ be functions.}$$

We say f & g are inverses if

$$f \circ g: \mathbb{R}^m \rightarrow \mathbb{R}^m$$

$$g \circ f: \mathbb{R}^n \rightarrow \mathbb{R}^n$$

are both the identity (do-nothing) function. Now let's suppose that f & g are linear functions with corresponding matrices:

$$[f] \quad m \times n$$

$$[g] \quad n \times m$$

If f & g are inverse functions, then this implies:

$$[f][g] = \mathbf{I} \quad (m \times m \text{ identity})$$

$$[g][f] = \mathbf{I} \quad (n \times n \text{ identity})$$

[Subtle Fact: In this case we must have $m = n$, so $[f]$ & $[g]$ are square matrices of the same size.]

Definition: Let A be a square $n \times n$ matrix.

We say that A is invertible if there exists a square $n \times n$ matrix B such that

$$AB = \underline{I} \quad (n \times n)$$

$$BA = \underline{I} \quad (n \times n).$$

In this case, the matrix B is unique, and we call it

" A^{-1} " = the inverse matrix of A

Examples:

- Let $R_\theta = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix}$.

What is the inverse matrix?

Geometrically, we see that

$$\begin{aligned} R_\theta^{-1} &= \text{rotate } \underline{\text{clockwise}} \text{ by } \theta \\ &= \text{rotate c.c.w. by } -\theta \end{aligned}$$

$$= R_{-\theta}$$

$$= \begin{pmatrix} \cos(-\theta) & -\sin(-\theta) \\ \sin(-\theta) & \cos(-\theta) \end{pmatrix}$$

$$= \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}.$$

Check :

$$R_{\theta} R_{-\theta} = \begin{pmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{pmatrix} \begin{pmatrix} \cos \theta & \sin \theta \\ -\sin \theta & \cos \theta \end{pmatrix}$$

$$= \begin{pmatrix} \cos^2 \theta + \sin^2 \theta & \cos \theta \sin \theta - \cos \theta \sin \theta \\ \cos \theta \sin \theta - \cos \theta \sin \theta & \sin^2 \theta + \cos^2 \theta \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \mathbf{I} \quad \checkmark$$

The algebra & the geometry agree,
as they should. 😊

- Let F = reflect across line $x=y$
What is the inverse?

Claim : $F^{-1} = F$.

Indeed, to undo the reflection we should just do the same reflection again!

Check that the algebra works :

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

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

[Remark : The identity matrix is not the only matrix that is equal to its own inverse ; any reflection also has this property.]

- The $n \times n$ zero matrix O is never invertible because for any $n \times n$ matrix A we have

$$OA = O \neq I \quad \& \quad AO = O \neq \underline{I}$$

• I claim that the matrix

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

is not invertible. To see this
suppose that we have

$$NA = I$$

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

for some matrix $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$.

But then we compute

$$\begin{pmatrix} 0 & 1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} c & d \\ 0 & 0 \end{pmatrix},$$

which is a contradiction because
the bottom-right entry is never
equal to 1.

[Remark : There exist nonzero matrices that are not invertible. In fact, we will see later that any "projection matrix" is not invertible.

The theory of matrix arithmetic is quite a bit more interesting than the usual arithmetic of numbers (i.e. 1×1 matrices) . }