

Lecture 20

Linear Transformation II

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Today's Lecture: Outline

Main topic: Linear transformation II

1. Two Definitions of Linear Transformation
2. How to Derive Expression of Linear Transformation
3. Linear Transformation on General Linear Spaces

Strang's book: Sec 8.1, 8.2

Today's Lecture: Learning Goals

After the lecture, you should be able to

1. Tell two definitions and why they are equivalent
2. Verify linear transformation and compute it for Euclidean spaces
3. Tell the relation of linear transformation and matrix
4. Verify linear transformation and compute it for general linear spaces

Comments

Midterm Exam ~~Design Idea~~

Midterm Exam Common Issues

“读” Chinese & English

space

$$C(A) = c_1 a_1 + \dots + c_n a_n \text{ vector}$$

Issue: Confuse space and vectors.

CUHKSZ = Professor.
= Prof. Xu.

space

$$C(A) = n - \dim(N(A)). \text{ number}$$

Issue: LHS is a space, RHS is “dimension” which is a number.
Cannot be the same

→ price = 3 yuan
→ price = 3

~~CUHKSZ = 11,000~~
CUHKSZ has 11,000 ppl.

$$③ C([A,B]) = \{ a_1, \dots, a_n, b_1, \dots, b_n \}$$

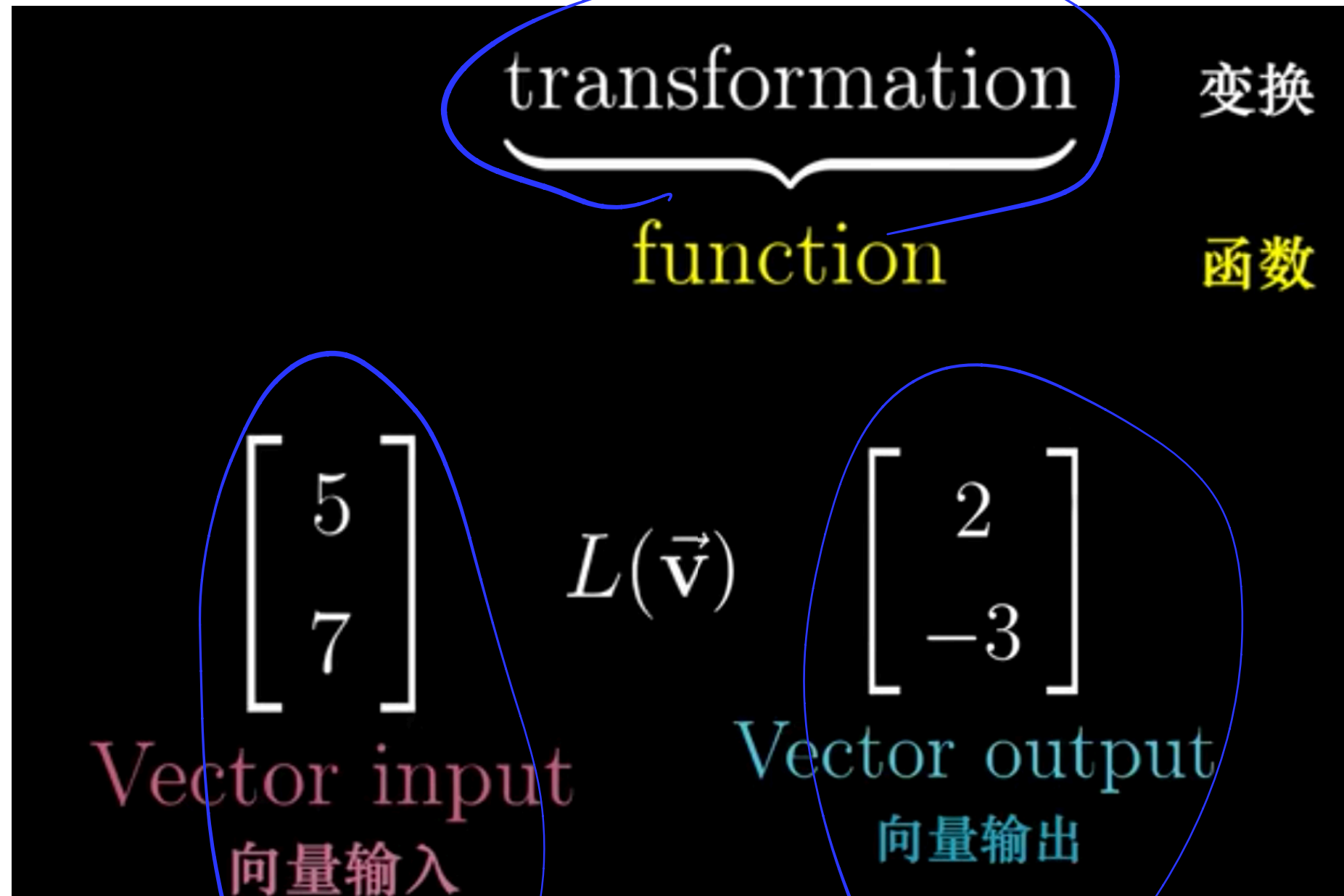
space

set

$$\text{span} \{ a_1, \dots, a_n, b_1, \dots, b_n \} \quad \infty \text{-size set}$$

Review

Transformation: Vector Input, Vector Output



Rotate Photos

How to rotate photos on iPhone



Motivating Question

Yes, I know you can do it on the phones.

How do the phones accomplish the job?



Question 1: Is it a linear transformation?

Question 2: How do I derive its expression?

Part I Another Definition of Linear Transformation

Linear Transformation ==> Superposition (疊加)

Property 19.1 [superposition property]

If f is a linear transformation from \mathbb{R}^n to \mathbb{R}^m , then

$$f(\alpha \mathbf{x} + \beta \mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n. \quad (*)$$

$$f \circ \text{LC} = \text{LC} \circ f$$

In words, transformed LC of vectors = LC of transformed vectors.

Examples:

$$f(x) = 2x,$$

$$f(\alpha x + \beta y) = 2(\alpha x + \beta y) = \alpha \cdot (2x) + \beta \cdot (2y) = \alpha f(x) + \beta f(y)$$

Non-examples:

$$f(x) = x^2,$$

$$f(\alpha x + \beta y) = (\alpha x + \beta y)^2 = \alpha^2 x^2 + \beta^2 y^2 + 2\alpha\beta xy$$
$$\neq \alpha x^2 + \beta y^2 \quad (= \alpha f(x) + \beta f(y))$$

Linear Transformation ==> Superposition (叠加)

Property 19.1 [superposition property]

If f is a linear transformation from \mathbb{R}^n to \mathbb{R}^m , then

$$f(\alpha \mathbf{x} + \beta \mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n. \quad (*)$$

In words, transformed LC of vectors = LC of transformed vectors.

Corollary 19.1

If f is a linear transformation from \mathbb{R}^n to \mathbb{R}^m , then

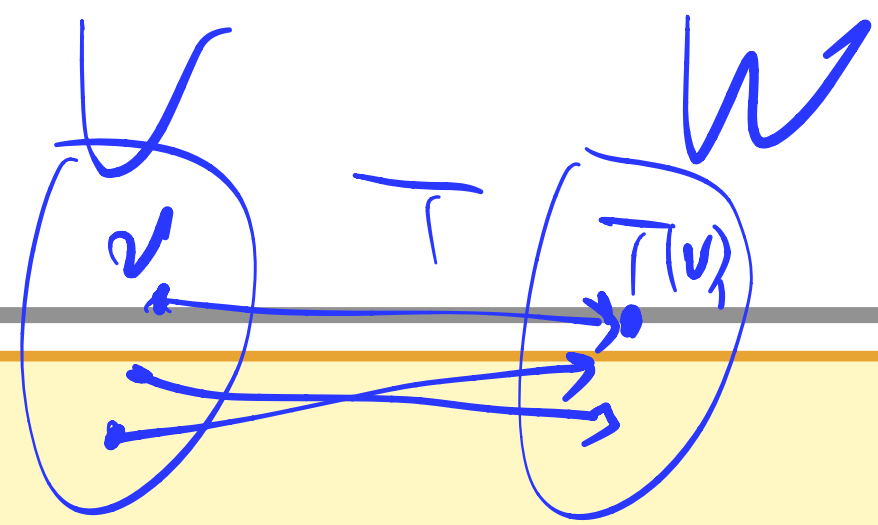
$$f(\alpha \mathbf{x}) = \alpha f(\mathbf{x}). \quad \textcircled{1}$$

$$f(\mathbf{x} + \mathbf{y}) = f(\mathbf{x}) + f(\mathbf{y}), \quad \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n. \quad \textcircled{2}$$

$$(*) \Leftrightarrow \textcircled{1} \ \& \ \textcircled{2}$$

$$\begin{aligned} f(x) &= 2x \\ f(\alpha x) &= 2\alpha x = \alpha f(x) \\ f(x+y) &= 2(x+y) \\ &= 2x + 2y = f(x) + f(y) \end{aligned}$$

Mapping



Definition 20.1 (Mapping)

Let V and W be two **sets**.

If for any $v \in V$, there is a unique $w = T(v) \in W$, then we say T is a **mapping** from V to W .

V is called the **domain** of the mapping.

W is called the **codomain** of the mapping.

Special case:

Suppose $f_i(\mathbf{x})$ is a function from \mathbb{R}^n to \mathbb{R} , $i = 1, \dots, m$.

$f(\mathbf{x}) = (f_1(\mathbf{x}), \dots, f_m(\mathbf{x}))$ is called a **mapping** from \mathbb{R}^n to \mathbb{R}^m .

Remark: Can also call it “**vector function**” (向量函数).

LT \Rightarrow S.P.

Superposition \Rightarrow Linear Transformation

Theorem 20.1

If a mapping f from \mathbb{R}^n to \mathbb{R}^m satisfies

$$f(\alpha \mathbf{x} + \beta \mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n,$$

Then f is a **linear transformation** from \mathbb{R}^n to \mathbb{R}^m .

S.P. \Rightarrow LT. $f(\mathbf{x}) = \mathbf{A} \cdot \mathbf{x}$

\downarrow \downarrow

$m \times n$ \mathbb{R}^n

Proof for $m = 1$

Proposition 20.1 ($n=1$ case for Thm 20.1)

If a function $f: \mathbb{R}^n \rightarrow \mathbb{R}$ satisfies

$$f(\alpha \mathbf{x} + \beta \mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n,$$

then we must have $f(\mathbf{x}) = \mathbf{a}^T \mathbf{x}$ for some \mathbf{a} .

1×1

Exercise 1: Prove that if $f: \mathbb{R} \rightarrow \mathbb{R}$ satisfies (*) for $n=1$, then $f(x) = ax$, $\forall x \in \mathbb{R}$ for some $a \in \mathbb{R}$.

Hint: What is $f(2)$? $f(3)$? $f(2.5)$?

Exercise 2: Prove that if $f: \mathbb{R}^2 \rightarrow \mathbb{R}$ satisfies (*), then

2×1

$f(\mathbf{x}) = a_1 x_1 + a_2 x_2$ for some $a_1, a_2 \in \mathbb{R}$.

Proof for $m=1, n=1$

Known

Goal

Analysis,

Conditions

$$\begin{cases} f(\alpha x) = \alpha f(x) & \textcircled{1} \\ f(x+y) = f(x) + f(y) & \textcircled{2} \end{cases}$$

Conclusion

Want $f(x) = ax$ for some a .

Hint: What is $f(2)$? $f(3)$? $f(2.5)$?

What do we hope $f(2), f(3), f(2.5)$ to be?

$$f(2) = 2 \cdot f(1)$$

$$f(3) = 3 \cdot f(1)$$

$$f(2.5) = 2.5 f(1)$$

Bridge \rightarrow

$$f(2) = 2a$$

$$f(3) = 3a$$

$$f(2.5) = 2.5a$$

for some a

Let $a \stackrel{\text{def}}{=} f(1)$, then

Proof: $f(x) = f(x \cdot 1) \stackrel{\textcircled{1}}{=} x f(1) = ax$,
where $a \stackrel{\text{def}}{=} f(1)$. \square

Proof for $m=1, n=2$

Conditions: $f(\alpha x) = \alpha f(x)$, ① $f(x+y) = f(x) + f(y)$. ② $f(\alpha x + \beta y) = \alpha f(x) + \beta f(y)$ (*)

Want: $f\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) = a_1 x_1 + a_2 x_2$ for some a_1, a_2 .

Analysis. $f\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right), f\left(\begin{bmatrix} 2 \\ 3 \end{bmatrix}\right) = ?$ ~~$f\left(\begin{bmatrix} 2 \\ 2 \end{bmatrix}\right) = 2f\left(\begin{bmatrix} 1 \\ 1 \end{bmatrix}\right)$~~

$$f\left(\begin{bmatrix} 1 \\ 2 \end{bmatrix}\right) \stackrel{||\textcircled{1}}{=} f(1 \cdot \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}) = 1f(e_1) + 2f(e_2)$$

$$f\left(\begin{bmatrix} 2 \\ 3 \end{bmatrix}\right) = f(2e_1 + 3e_2) = 2f(e_1) + 3f(e_2)$$

Key.
basis representn
 $\begin{bmatrix} a \\ b \end{bmatrix} = a \begin{bmatrix} 1 \\ 0 \end{bmatrix} + b \begin{bmatrix} 0 \\ 1 \end{bmatrix}$

Proof. Denote

$$\begin{aligned} \text{Then } f\left(\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}\right) &= f(x_1 e_1 + x_2 e_2) \\ &\stackrel{||\textcircled{1}}{=} x_1 f(e_1) + x_2 f(e_2) \end{aligned}$$

$$= a_1 x_1 + a_2 x_2, \text{ where } a_1 = f(e_1), a_2 = f(e_2)$$

Proof for $m=1$, general n

Conditions: $f(\alpha x) = \alpha f(x)$, ① $f(x+y) = f(x) + f(y)$. ②. $f: \mathbb{R}^n \rightarrow \mathbb{R}$.

Want: $f(x) = a^T x$, $\forall x \in \mathbb{R}^n$, for some $a \in \mathbb{R}^n$.

Proof: $f\left(\begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}\right) = f(x_1 e_1 + \dots + x_n e_n)$

$$= x_1 f(e_1) + \dots + x_n f(e_n)$$

$$= a_1 x_1 + \dots + a_n x_n,$$

where $a_i = f(e_i)$, $\forall i$.

Proof for general m, general n

Conditions: $f(\alpha x) = \alpha f(x)$, ① $f(x+y) = f(x) + f(y)$. ②. $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$

Want: $f(x) = Ax$, $\forall x \in \mathbb{R}^n$, for some $A \in \mathbb{R}^{m \times n}$.

Proof: Suppose $f(x) = \begin{pmatrix} f_1(x) \\ f_2(x) \\ \vdots \\ f_m(x) \end{pmatrix}$.

From ①, ②, we get (for each $f_i(x)$)

$$f_i(\alpha x) = \alpha f_i(x), \quad f_i(x+y) = f_i(x) + f_i(y)$$

From Prop. 20.1, we get:

Denote $A = \begin{bmatrix} a_1^T \\ \vdots \\ a_m^T \end{bmatrix}$ $f_i(x) = a_i^T x$, for some $a_i \in \mathbb{R}^{n \times 1}$.

$$\text{Then } f(x) = \begin{bmatrix} f_1(x) \\ \vdots \\ f_m(x) \end{bmatrix} = \begin{bmatrix} a_1^T x \\ \vdots \\ a_m^T x \end{bmatrix} = A \cdot x. \quad \square$$

Another Definition

Definition 20.2 (alternative definition of LT):

If a mapping f from \mathbb{R}^n to \mathbb{R}^m satisfies

$$f(\alpha \mathbf{x} + \beta \mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n,$$

Then we say f is a **linear transformation** from \mathbb{R}^n to \mathbb{R}^m .

Equivalent definition to Def 20.1.

Using properties to define sth.

Linear Transformation in Euclidean Space: Two Definitions

Two equivalent definitions:

Definition 19.1 (matrix form):

Suppose $A \in \mathbb{R}^{m \times n}$ is a given real matrix.

$f(\mathbf{x}) = A\mathbf{x}$ is called a linear transformation from \mathbb{R}^n to \mathbb{R}^m .

$$\begin{array}{c} a\mathbf{x} \\ \downarrow \\ A\mathbf{x} \end{array}$$

Definition 20.1 (alternative definition of LT):

If a mapping f from \mathbb{R}^n to \mathbb{R}^m satisfies

$$f(\alpha\mathbf{x} + \beta\mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n,$$

Then we say f is a linear transformation from \mathbb{R}^n to \mathbb{R}^m .

Why Equivalent

$Ax \Rightarrow$ property def

Property 20.1 [Def 19.1 \Rightarrow Def 20.1]

Suppose $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ is defined as $f(\mathbf{x}) = A\mathbf{x}$, then

$$f(\alpha\mathbf{x} + \beta\mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n.$$

Reverse:

Property Def \rightarrow Ax def.

Theorem 20.1 [Def 20.1 \Rightarrow Def 19.1]

If a mapping f from \mathbb{R}^n to \mathbb{R}^m satisfies

$$f(\alpha\mathbf{x} + \beta\mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n,$$

Then $f(\mathbf{x}) = A\mathbf{x}$ for some matrix A .

rank \rightarrow def. # of pivots
 \rightarrow def. dim (Col span)

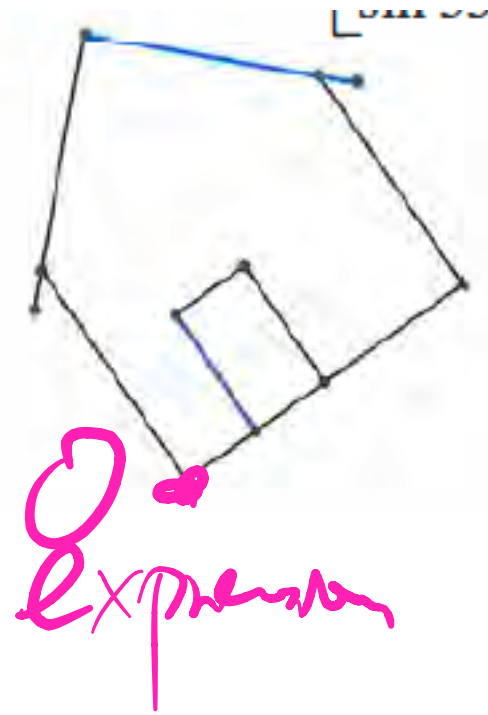
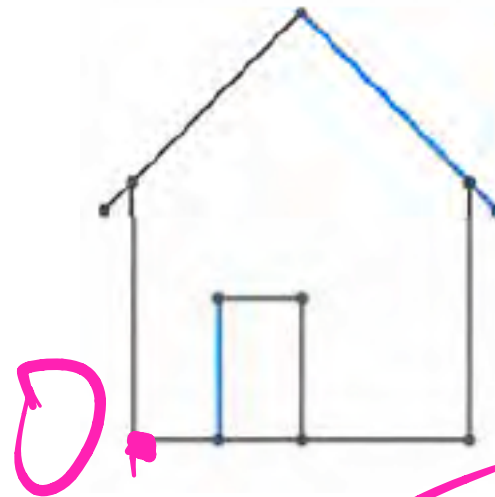
Part II Expression of Linear Transformation

Transformation: Movement

“变换”这个词在暗示你用运动去思考

The word “transformation” suggests
that you think using movement

Description of "Rotation"



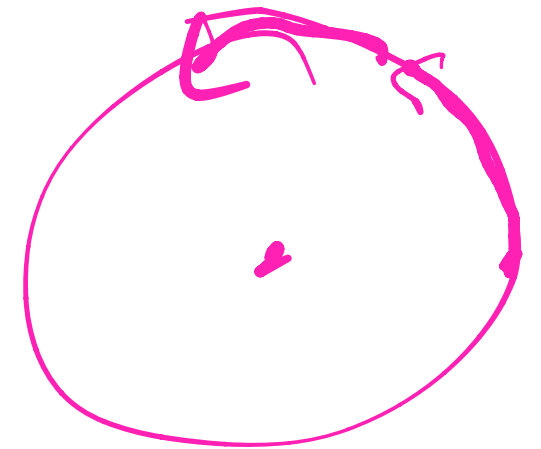
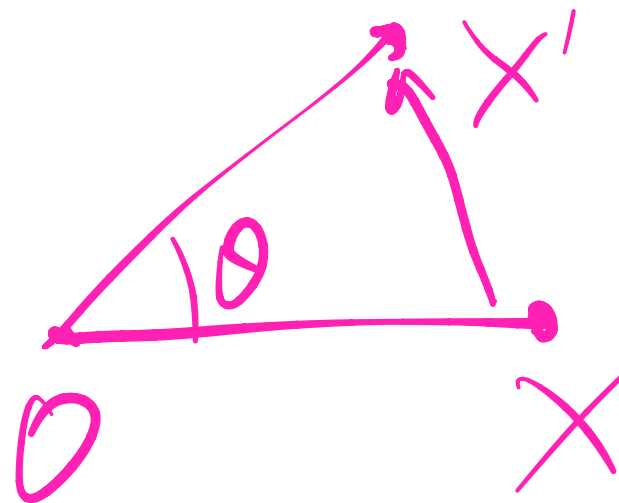
Rotation is a "transformation".

Precise description of rotation:

"Rotating an image around point O by degree θ " means:

For any point X , the new point X' satisfies $\angle(OX, OX') = \theta$.

$$|OX| = |OX'|$$



Q1 How to Check Linear Transformation

Question 1: Is it a **linear** transformation?

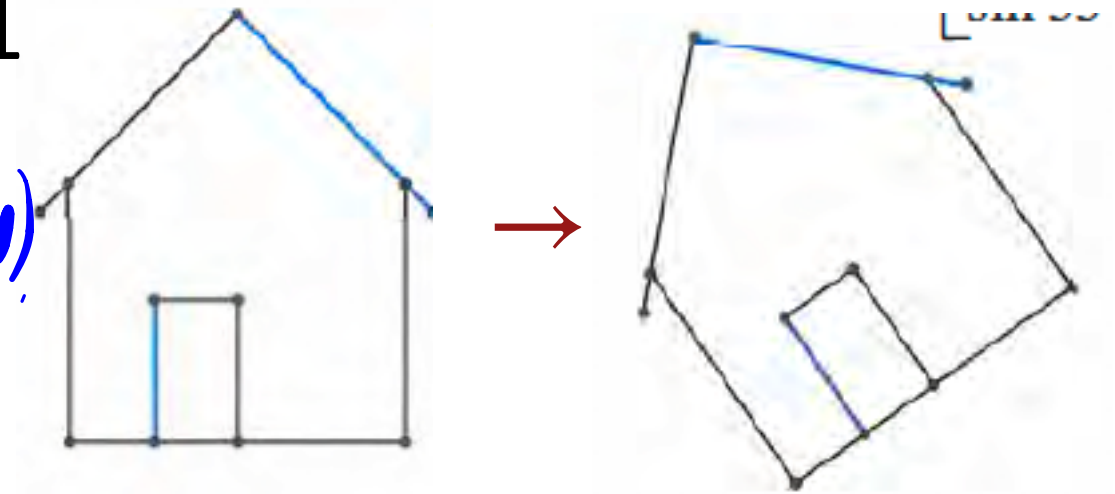
Check: It satisfies Property 19.1

$$f(\alpha x + \beta y) = \alpha f(x) + \beta f(y)$$

algebraic.



geometrical way of checking.



Checking Linear Transformation: Intuition

Question 1: Is it a **linear** transformation?

$$f(\alpha \mathbf{x} + \beta \mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}), \quad \forall \alpha, \beta \in \mathbb{R}, \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n$$

Fix \mathbf{x}, \mathbf{y} , consider all $\alpha, \beta \in \mathbb{R}$.

$$\alpha + \beta = 1$$

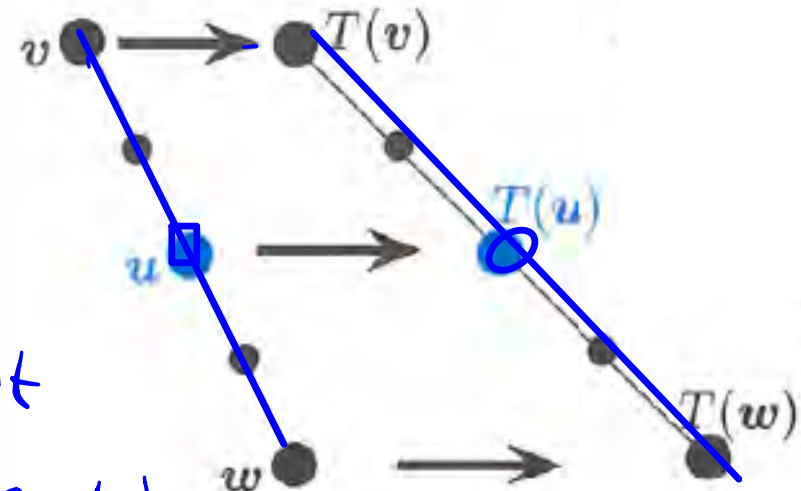
Rule 1: Lines to Lines.

Rule 2: Equally spaced points to equally spaced points.

$$\alpha = \beta = 0$$

Rule 3: Origin to Origin.

$$f(\mathbf{0}) = 0 + 0 = 0.$$



Derivation of "Lines to Lines"

The superposition property implies

$$\{f(\alpha \vec{x} + \beta \vec{y}) : \alpha + \beta = 1\} = \{\alpha f(\vec{x}) + \beta f(\vec{y}) : \alpha + \beta = 1\} \quad (1)$$

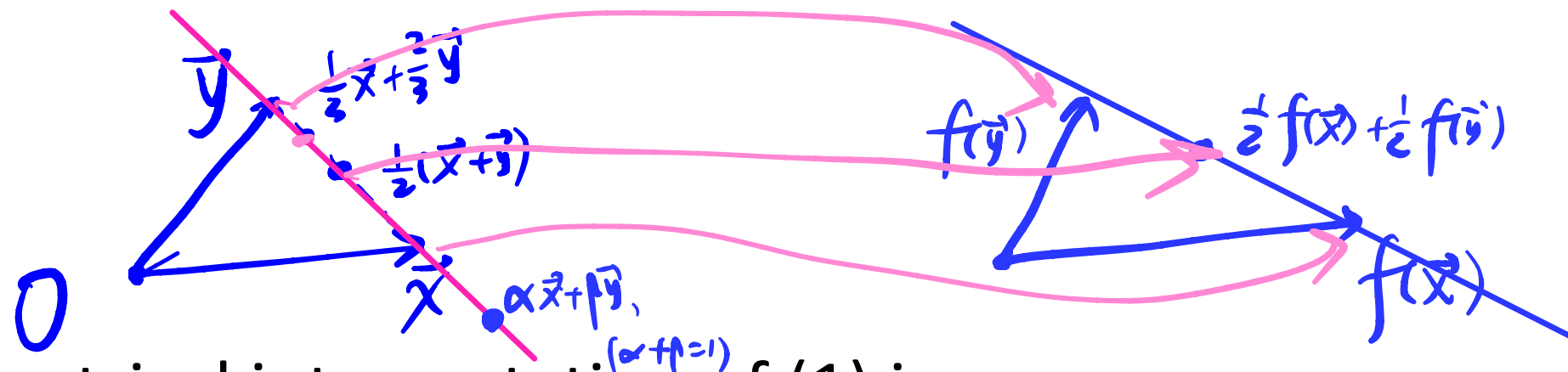
$$(1) \Leftrightarrow \Omega_1 \xrightarrow{f} \Omega_2 \stackrel{(2)}{\Leftrightarrow} \text{line} \xrightarrow{f} \text{line}$$

$$\Omega_1 \triangleq \{\alpha \vec{x} + \beta \vec{y} \mid \alpha + \beta = 1\}$$

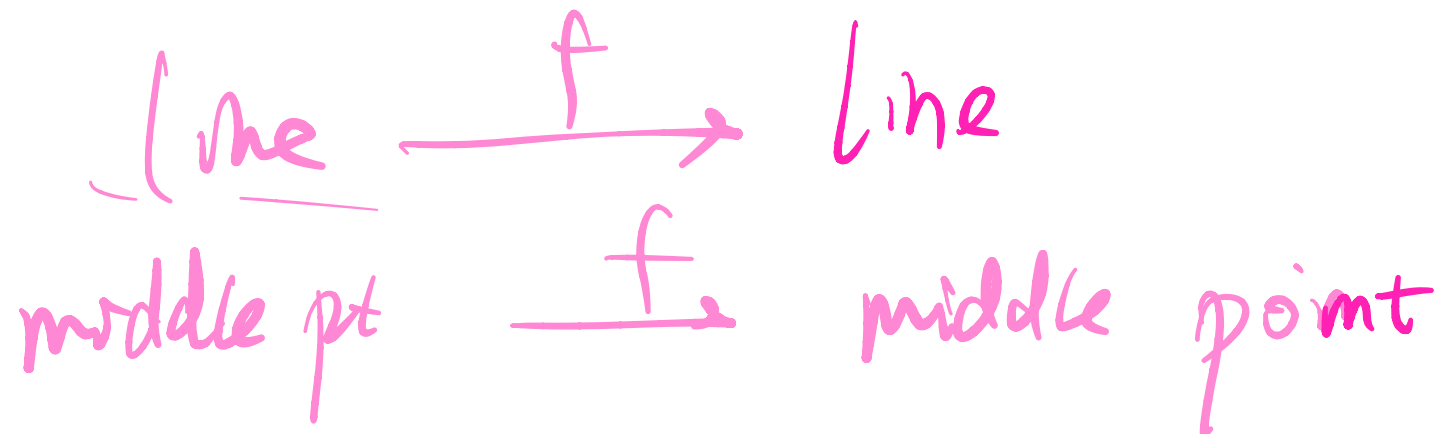
$$\Omega_2 \triangleq \{\alpha f(\vec{x}) + \beta f(\vec{y}) \mid \alpha + \beta = 1\}$$

To show: $\Omega_1 = \text{line}$ (2)

$\Omega_2 = \text{line}$, (2)



Geometrical interpretation of (1) is

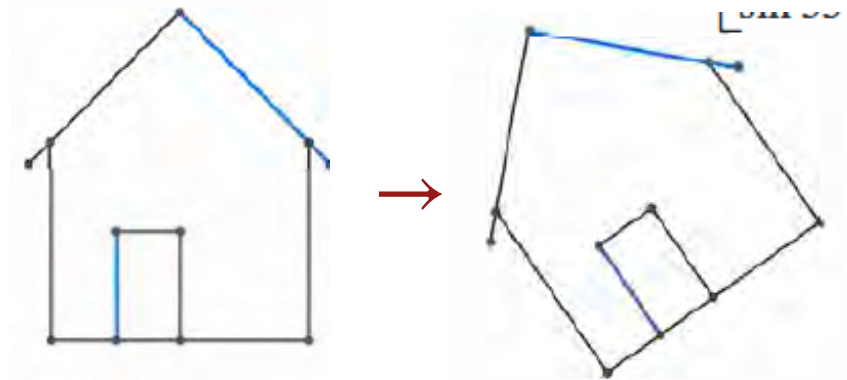


Use 3 Rules to Check

Rule: $O \rightarrow O$.

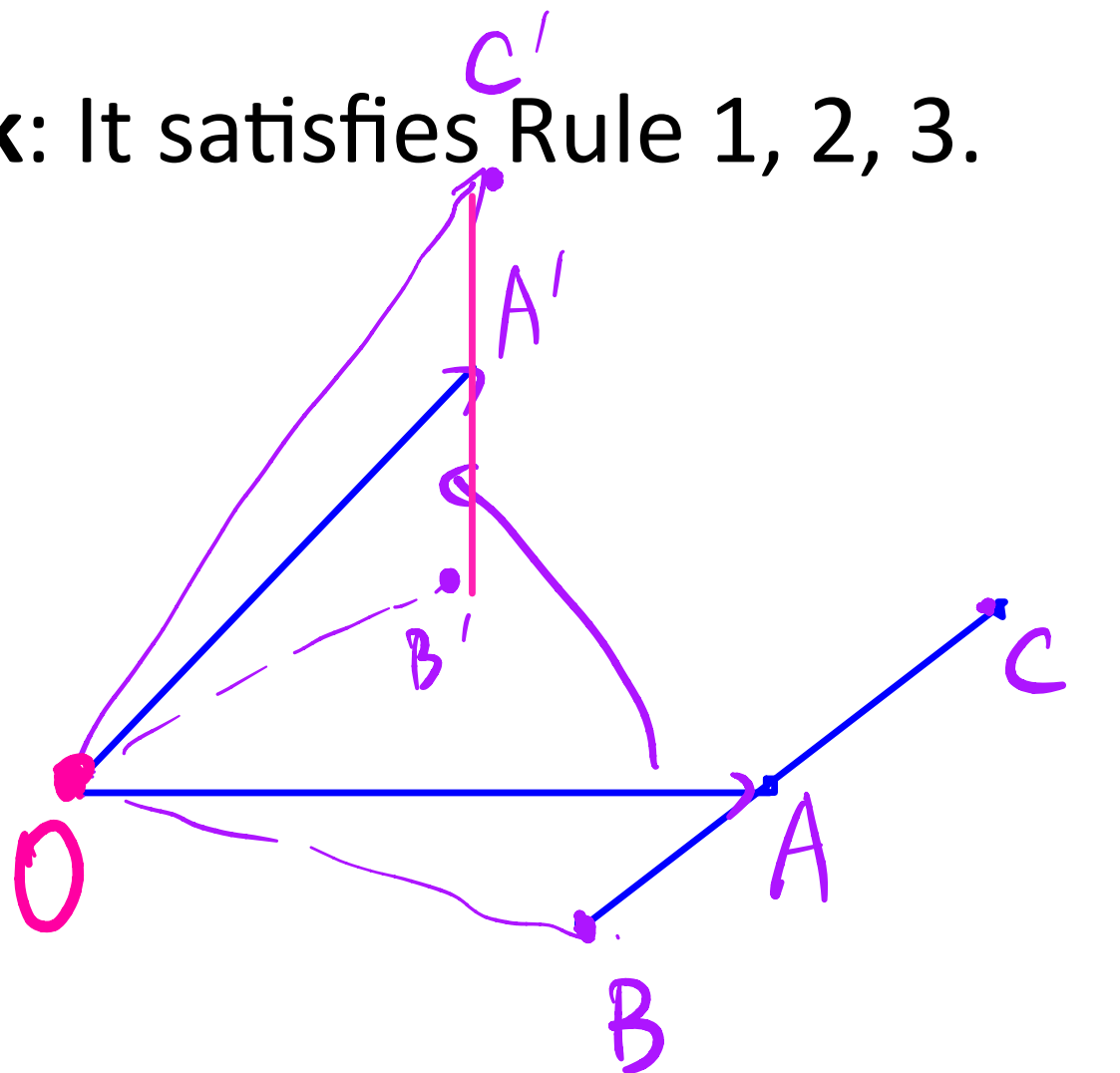
Question 1: Is it a **linear** transformation?

Check: It satisfies Rule 1, 2, 3.



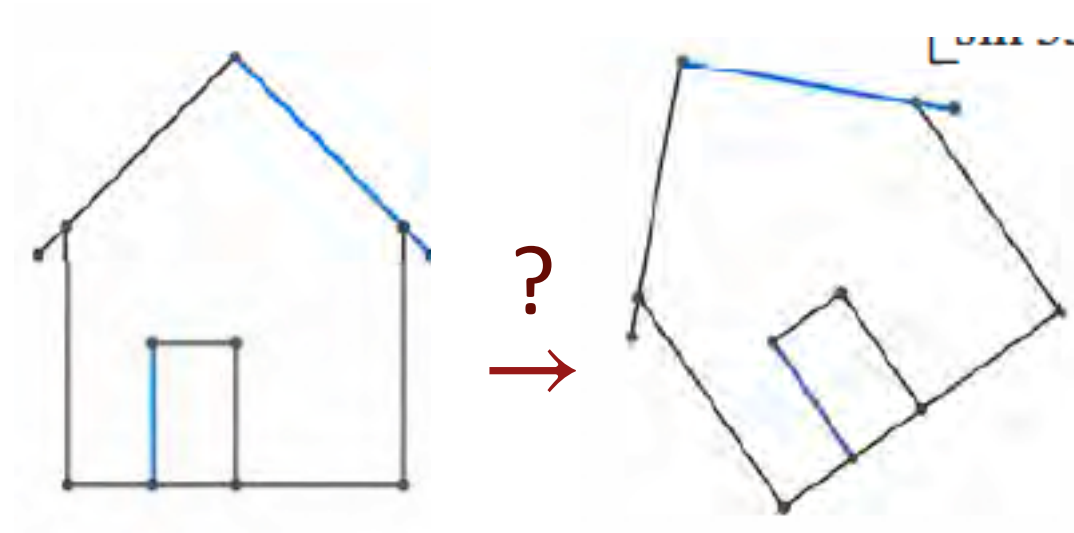
rotate θ

A	→	A'
B	→	B'
C	→	C'



- ① A', B', C' same line. (can prove by similar triangles, 全等三角形)
- ② A' is the middle point of $B'C'$.

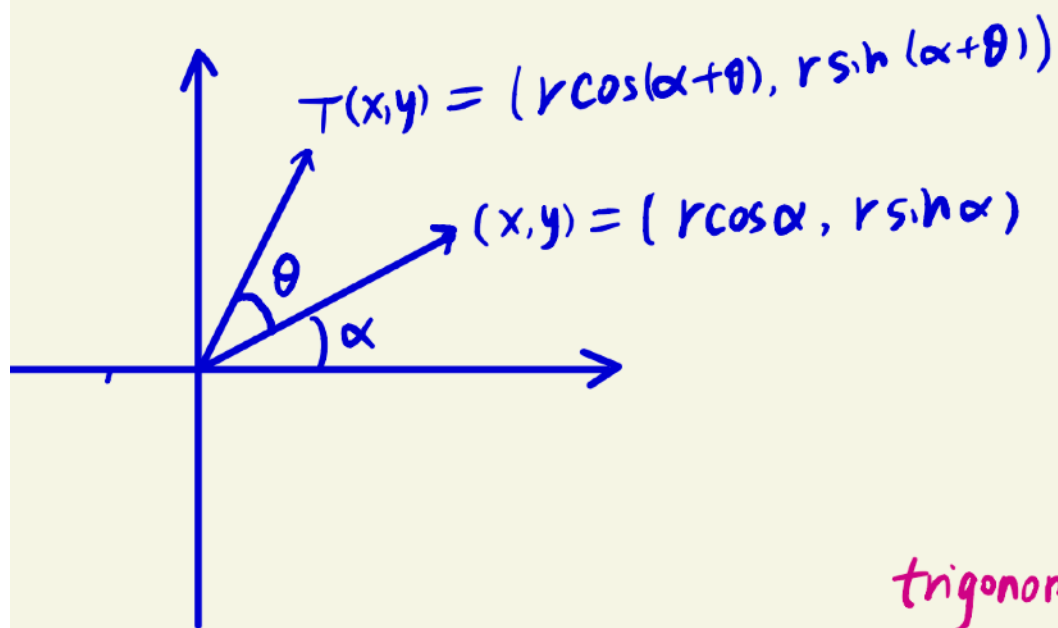
Q2: Math Form of Rotation



Q2: How to **derive** its expression?

Reading: Direct Derivation (high school student)

Method 1.



What's the relation of $T(x, y)$ and (x, y) ?

$$(x, y) = (r \cos \alpha, r \sin \alpha)$$

for some α , some $r > 0$.

$$T(x, y) = \begin{bmatrix} r \cos(\alpha + \theta) \\ r \sin(\alpha + \theta) \end{bmatrix}$$

trigonometric identities
[三角恒等式]

$$= \begin{bmatrix} r \cos \alpha \cos \theta - r \sin \alpha \sin \theta \\ r \sin \alpha \cos \theta + r \cos \alpha \sin \theta \end{bmatrix}$$

$$= \begin{bmatrix} x \cos \theta - y \sin \theta \\ y \cos \theta + x \sin \theta \end{bmatrix}$$

Remark This method requires knowing the trigonometric identities.

We'll provide a more general method next.



② (where)

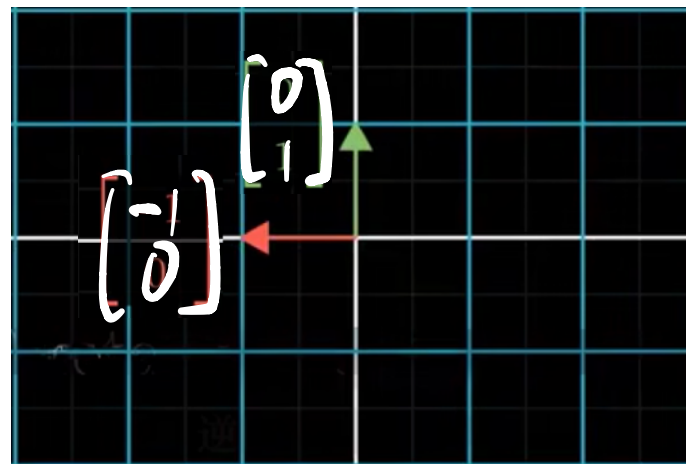
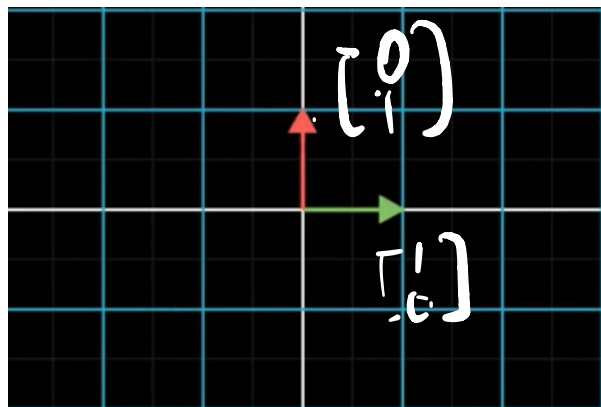
(LT) / Use method

- ① $T(e_1), T(e_2)$
- ② $A = [T(e_1), T(e_2)]$

Special Case: Rotating by 90 degree

Example: Rotation by 90°

$$\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$



$$\begin{aligned} e_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} & \xrightarrow{\text{rot } 90^\circ} T(e_1) = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \\ e_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} & \xrightarrow{\text{rot } 90^\circ} T(e_2) = \begin{bmatrix} -1 \\ 0 \end{bmatrix} \end{aligned}$$

$$x = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$T(x) = T\left(x_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \end{bmatrix}\right)$$

$$= x_1 T(e_1) + x_2 T(e_2)$$

$$= \begin{bmatrix} 0 \\ 1 \end{bmatrix} x_1 + \begin{bmatrix} -1 \\ 0 \end{bmatrix} x_2$$

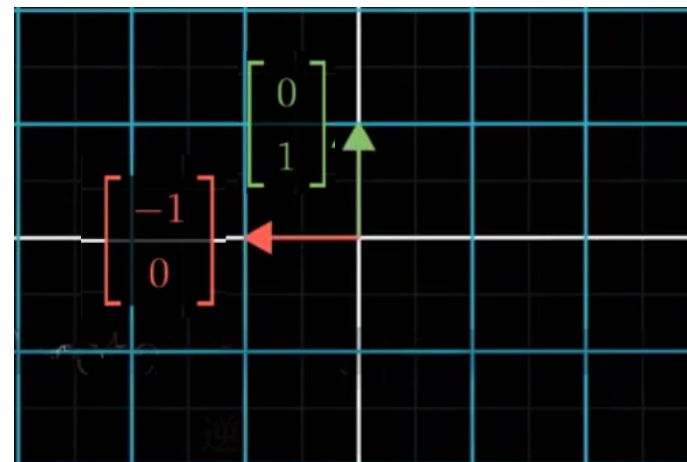
$$= \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} T(e_1) & T(e_2) \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

proof
20 mins ago

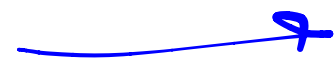
Special Case: Rotating by 90 degree

Example: Rotation by 90°

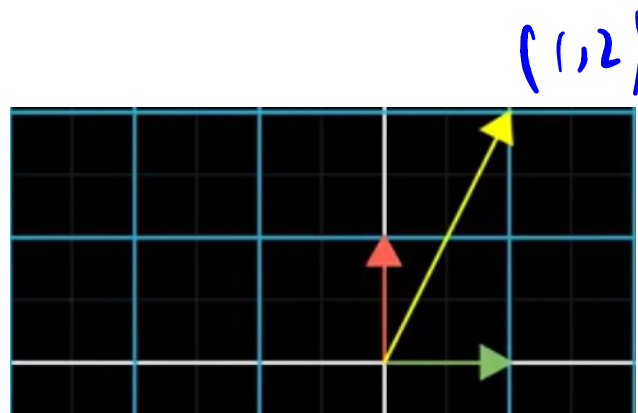
$$\begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix}$$



basis (e_1, e_2)



basis $(T(e_1), T(e_2))$



$x = (e_1, e_2) \begin{pmatrix} 1 \\ 2 \end{pmatrix}$
 basis change,

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

$T(x) = (T(e_1), T(e_2)) \begin{pmatrix} 1 \\ 2 \end{pmatrix}$
 coordinates same.

$$LT: \quad x \rightarrow y = Ax.$$

1) Change basis

$$e_1, \dots, e_n \longrightarrow T(e_1), \dots, T(e_n).$$

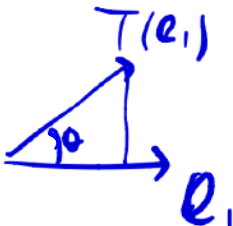
2) coord mate stay

$$y = [T(e_1), \dots, T(e_n)] \vec{x}.$$

Q2: Math Form of Rotation

Q2: How to **derive** its expression?

From basis!

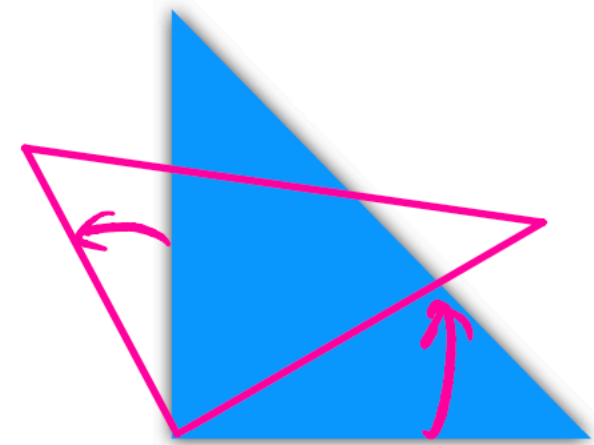
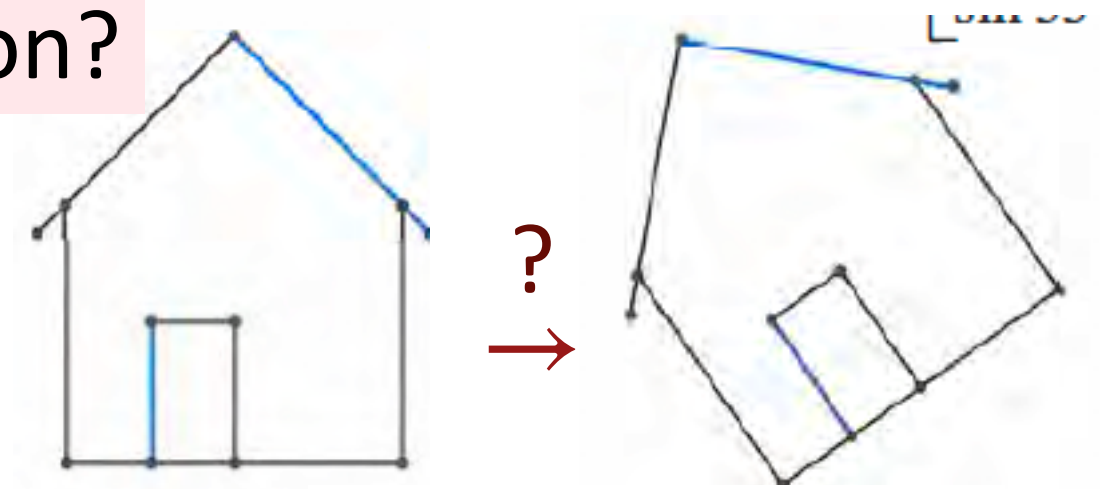
$$e_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \mapsto T(e_1) = \begin{bmatrix} \cos \theta \\ \sin \theta \end{bmatrix}$$


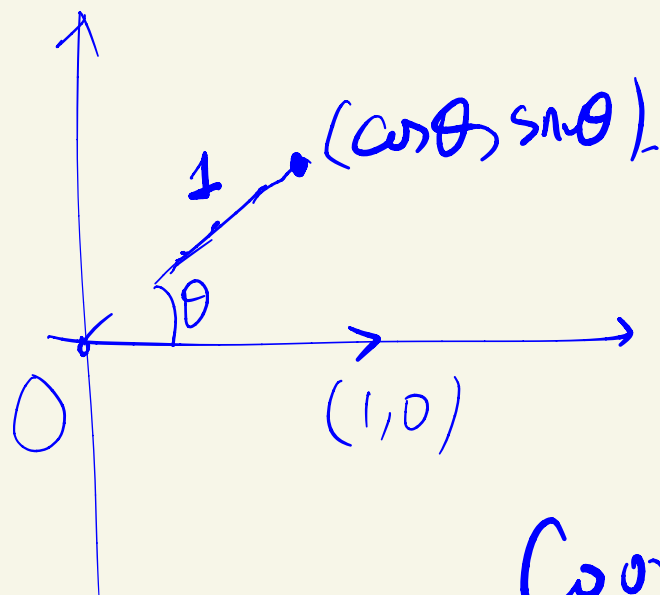
$$e_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \mapsto T(e_2) = \begin{bmatrix} -\sin \theta \\ \cos \theta \end{bmatrix}$$

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mapsto T(x) = x_1 T(e_1) + x_2 T(e_2)$$

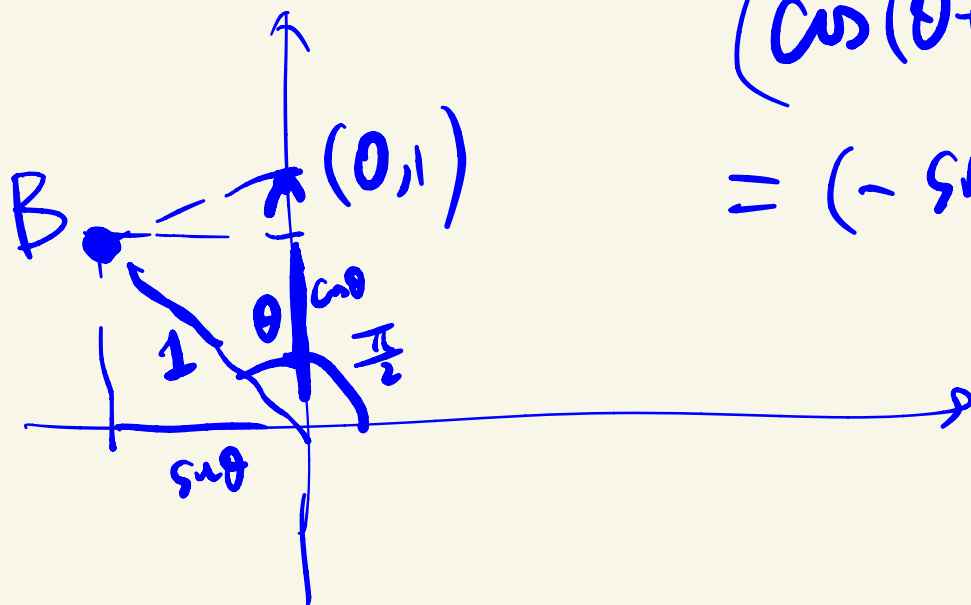
$$= \begin{bmatrix} T(e_1) & T(e_2) \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$= \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} x_1 \cos \theta - x_2 \sin \theta \\ x_1 \sin \theta + x_2 \cos \theta \end{pmatrix}$$





Coordinate of B:
 $(\cos(\theta + \frac{\pi}{2}), \sin(\theta + \frac{\pi}{2}))$
 $= (-\sin \theta, \cos \theta)$.



Computing Linear Transformation

Proposition 20.1

If f is a linear transformation from \mathbb{R}^n to \mathbb{R}^m , then

$$f\left(\sum_{i=1}^k \alpha_i \mathbf{v}_i\right) = \sum_{i=1}^k \alpha_i f(\mathbf{v}_i), \quad \forall \alpha_i \in \mathbb{R}, \forall \mathbf{v}_i \in \mathbb{R}^n, i = 1, \dots, k.$$

Algorithm 20.1

Step 1: Pick the standard basis $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ of \mathbb{R}^n .

Step 2: Compute $f(\mathbf{e}_i) \in \mathbb{R}^m, \forall i$.

Step 3 (?): For any $\mathbf{x} \in \mathbb{R}^n$, we have

$$f(\mathbf{x}) = \sum_i x_i f(\mathbf{e}_i).$$

Enough to compute $f(\mathbf{x})$.

But if want to write $f(\mathbf{x}) = A\mathbf{x}$, need extra step.

Q1: How to
compute f ?
(If you know f
is L.T.)

Matrix of Linear Transformation

Step 3 (?): For any $\mathbf{x} \in \mathbb{R}^n$, we have

$$f(\mathbf{x}) = \sum_i x_i f(\mathbf{e}_i) \in \mathbb{R}^m.$$

Q2 How to write $f(\mathbf{x})$ as $f(\mathbf{x}) = A\mathbf{x}$?

This formula computes $f(\mathbf{x})$ for any \mathbf{x} .

Can we write it as $f(\mathbf{x}) = A\mathbf{x}$ for some A ?

$$f(\mathbf{x}) = x_1 f(\mathbf{e}_1) + \dots + x_n f(\mathbf{e}_n)$$
$$= \underbrace{\left[f(\mathbf{e}_1) \quad \dots \quad f(\mathbf{e}_n) \right]}_A \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$$

Computing Linear Transformation

Algorithm 20.1 (2nd variant)

Answer Q2

What is A

m $f(x) = Ax?$

Step 1: Pick the standard basis $\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ of \mathbb{R}^n .

Step 2: Compute $f(\mathbf{e}_i) \in \mathbb{R}^m, \forall i$.

Step 3: Form a matrix $A = [f(\mathbf{e}_1), f(\mathbf{e}_2), \dots, f(\mathbf{e}_n)] \in \mathbb{R}^{m \times n}$

Conclusion: For any $\mathbf{x} \in \mathbb{R}^n$, we have

$$f(\mathbf{x}) = \sum_i x_i f(\mathbf{e}_i) = A\mathbf{x}.$$

A is the matrix of the linear transformation f .

For rotation, we've used Algorithm 20.1 to find A .

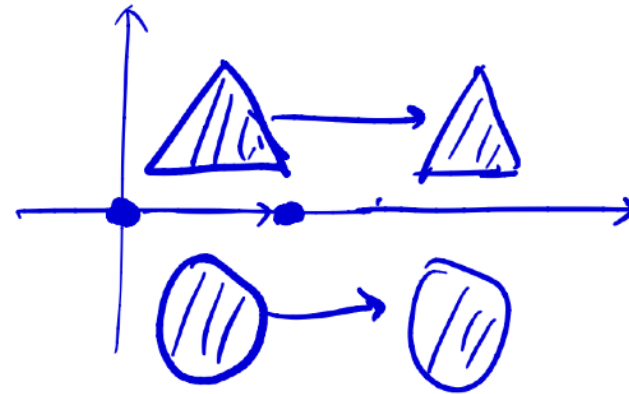
Analyzing Transformation: Eg1

Eg (translation, 平移):

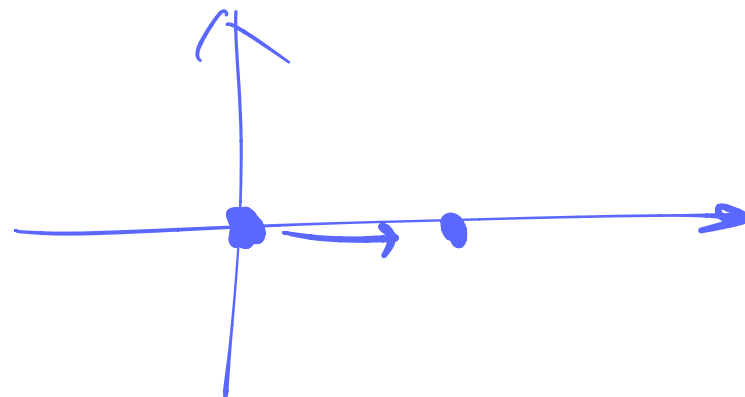
Suppose $T(x_1, x_2) = (x_1 + 1, x_2)$

Is it a linear transformation?

If so, find A such that $T(x) = Ax$



Intuition: Lines to lines; but origin NOT to origin .



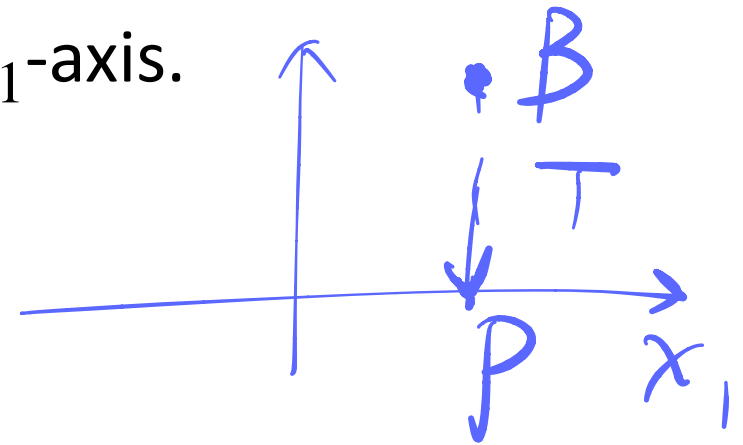
(affine transformation, $f(x) = Ax + b$),
($Ax = b$; shifted linear space)

Analyzing Transformation: Eg2

Eg: Suppose $T(x)$ is the **projection** of (x_1, x_2) onto x_1 -axis.

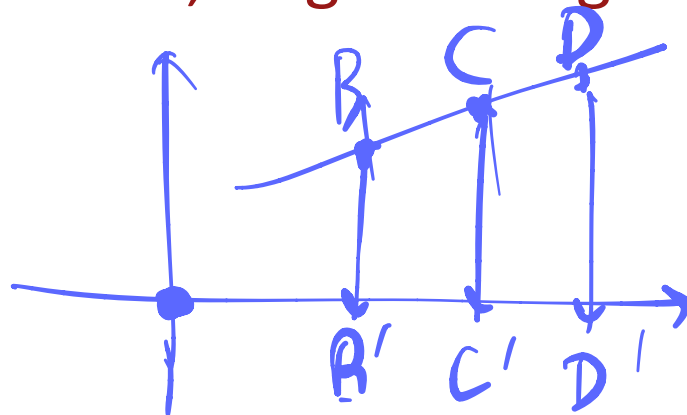
Is it a linear transformation?

If so, find A such that $T(x) = Ax$



Intuition: Lines to lines; origin to origin.

Method 1:



C is middle pt
 \downarrow
 middle pt C'

Method 2: $T(x_1, x_2) = x_1$

$$T(x_1, x_2) = \underbrace{\begin{bmatrix} 1 & 0 \end{bmatrix}}_A \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}, \quad \text{so n L.T.}$$

Method 3, $T(x_1, x_2) = x_1 = T(e_1) + x_2 \cdot T(e_2) = \begin{bmatrix} T(e_1) & T(e_2) \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$

Analyzing Transformation: Eg3

Eg: Let $L : \mathbb{R}^2 \rightarrow \mathbb{R}$ be a mapping defined as:

norm

$$L((x, y)^T) = \sqrt{x^2 + y^2}$$

Is it a linear transformation?

If so, find A such that $T(x) = Ax$

$$L(\vec{v}) = \|\vec{v}\|$$

L is not a linear transformation since

$$\begin{aligned} L(-\vec{v}) &= L(-(x, y)^T) = L((-x, -y)^T) \\ &= \sqrt{(-x)^2 + (-y)^2} = \sqrt{x^2 + y^2} \neq -L((x, y)^T) \\ &= \|\vec{v}\| = L(\vec{v}) \neq (-1) \cdot L(\vec{v}) \end{aligned}$$

Eg: Projection onto Circles

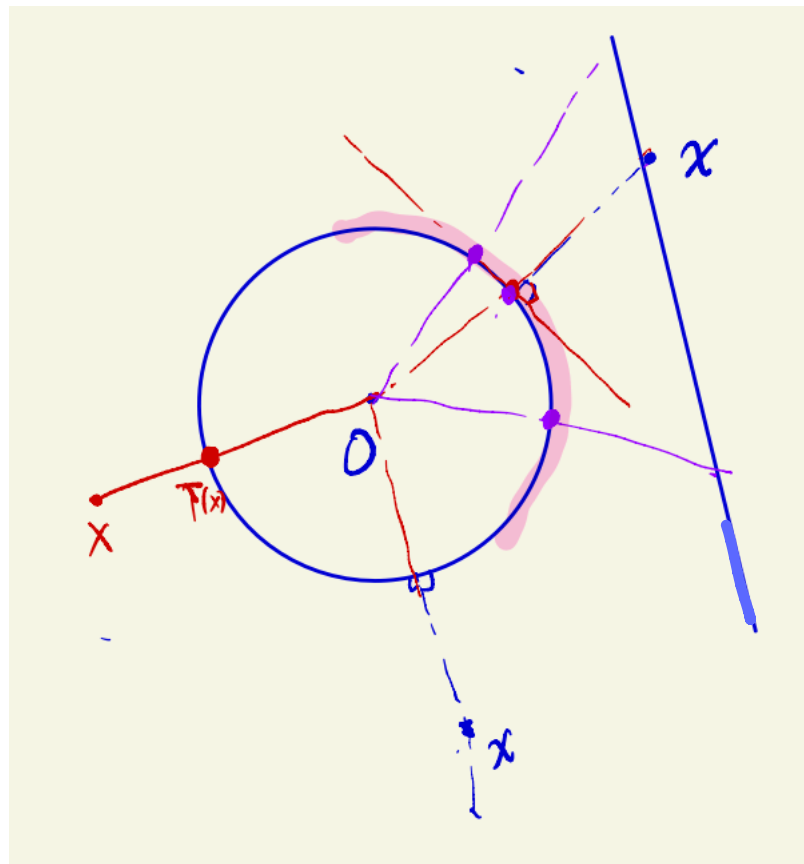
(Dropping apples to earth)

Transformation T : Given a circle C .

For any $X \in \mathbb{R}^2$, connect OX , and denote the intersection of OX and C as X' .


This X' is defined as $T(X)$.

Question: Is T a linear transformation?



Answer: No.

line \xrightarrow{T} part of circle
NOT LT.



Summary Today (Write Your Own)

One sentence summary:

Detailed summary:

Summary Today (Write Your Own)

One sentence summary:

We learned general linear transformation and how to “express” it.

Detailed summary: ① Two definitions of Linear transformations.

② Deriving Expression of Linear transformation (Euclidean space)

Find basis; Find the transformed vectors. Put in columns of a matrix.
—> Matrix of linear transformation.

③ Defining Linear transformation in Euclidean space

—> find basis of the Euclidean space

—> find transformed vectors of each element

—> find matrix of transformed vectors

—> find linear transformation matrix via transformed vectors; basis;

—> combine all the above to find matrix