

# Linear Algebra

[KOMS120301] - 2023/2024

## 13.3 - Properties of Linear Transformation

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# Learning objectives

After this lecture, you should be able to:

1. explain various properties of each of linear transformations in a vector space.

# Properties of Matrix Transformations

(page 270 of Elementary LA Applications book)

# Compositions of matrix transformation

Let:

- $T_A$ : a matrix transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^k$
- $T_B$ : a matrix transformation from  $\mathbb{R}^k$  to  $\mathbb{R}^m$

Let  $\mathbf{x} \in \mathbb{R}^n$ , and defines transformation:

$$\mathbf{x} \xrightarrow{T_A} T_A(\mathbf{x}) \xrightarrow{T_B} T_B(T_A(\mathbf{x}))$$

defines the transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ .

It is called the **composition of  $T_B$  with  $T_A$**  and is denoted by  $T_B \circ T_A$ . So:

$$(T_B \circ T_A)(\mathbf{x}) = T_B(T_A(\mathbf{x}))$$

# Compositions of matrix transformation

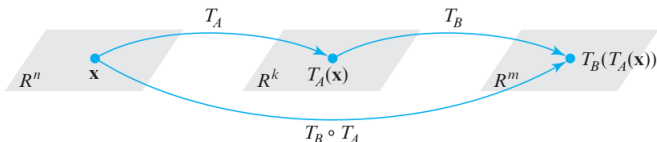
The composition is a matrix transformation, since:

$$(T_B \circ T_A)(\mathbf{x}) = T_B(T_A(\mathbf{x})) = B(T_A(\mathbf{x})) = B(A\mathbf{x}) = (BA)\mathbf{x}$$

meaning that the result of the composition to  $\mathbf{x}$  is obtained by multiplying  $\mathbf{x}$  with  $BA$  on the left.

This is denoted by:

$$T_B \circ T_A = T_{BA}$$



## Composition of three transformations

Compositions can be defined for any finite succession of matrix transformations whose domains and ranges have the appropriate dimensions. For instance, given:

$$T_A : \mathbb{R}^n \rightarrow \mathbb{R}^k, \quad T_B : \mathbb{R}^k \rightarrow \mathbb{R}^\ell, \quad T_C : \mathbb{R}^\ell \rightarrow \mathbb{R}^m$$

we can define the composition:

$$(T_C \circ T_B \circ T_A) : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

by:

$$(T_C \circ T_B \circ T_A)(\mathbf{x}) = T_C(T_B(T_A(\mathbf{x})))$$

It can be shown that this is a matrix transformation with standard matrix  $CBA$ , and:

$$T_C \circ T_B \circ T_A = T_{CBA}$$

# Notation

We can write the standard matrix for transformation  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  without specifying the name of the standard matrix.

It is often written as  $[T]$ .

For instance,

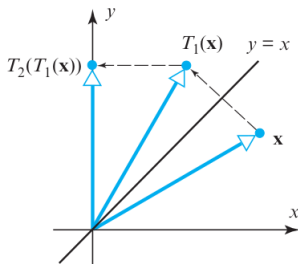
- $T(\mathbf{x}) = [T]\mathbf{x}$
- $[T_2 \circ T_1] = [T_2][T_1]$
- $[T_3 \circ T_2 \circ T_1] = [T_3][T_2][T_1]$

# Composition is not commutative

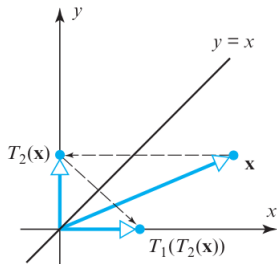
## Example

Let:

- $T_1 : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the reflection about the line  $y = x$ ;
- $T_2 : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the orthogonal projection onto the  $y$ -axis.



$T_2 \circ T_1$



$T_1 \circ T_2$

Geometrically, both transformations have different effect on  $x$



## Composition is not commutative (*cont.*)

Algebraically, we can compute:

$$[T_1 \circ T_2] = [T_1][T_2] = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}$$

$$[T_2 \circ T_1] = [T_2][T_1] = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}$$

Clearly,  $[T_1 \circ T_2] \neq [T_2 \circ T_1]$ .

# Composition of rotation is commutative

## Example

Given :

$$T_1 : \mathbb{R}^2 \rightarrow \mathbb{R}^2 \quad \text{and} \quad T_2 : \mathbb{R}^2 \rightarrow \mathbb{R}^2$$

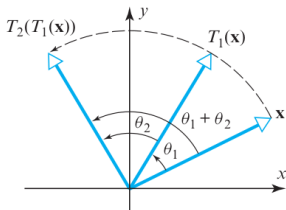
the matrix operators that rotate vectors about the origin through the angles  $\theta_1$  and  $\theta_2$  respectively.

So, the operation:

$$T_2 \circ T_1(\mathbf{x}) = T_2(T_1(\mathbf{x}))$$

first rotates  $\mathbf{x}$  through the angle  $\theta_1$ , then rotates  $T_1(\mathbf{x})$  through the angle  $\theta_2$ .

Hence,  $(T_2 \circ T_1)(\mathbf{x})$  defines rotation of  $\mathbf{x}$  through the angle  $\theta_1 + \theta_2$ .



## Composition of rotation is commutative (*cont.*)

In this case, we have:

$$[T_1] = \begin{bmatrix} \cos \theta_1 & -\sin \theta_1 \\ \sin \theta_1 & \cos \theta_1 \end{bmatrix} \quad \text{and} \quad [T_2] = \begin{bmatrix} \cos \theta_2 & -\sin \theta_2 \\ \sin \theta_2 & \cos \theta_2 \end{bmatrix}$$

We show that:  $[T_2 \circ T_1] = [T_2][T_1]$

$$\text{Note that } [T_2 \circ T_1] = \begin{bmatrix} \cos(\theta_1 + \theta_2) & -\sin(\theta_1 + \theta_2) \\ \sin(\theta_1 + \theta_2) & \cos(\theta_1 + \theta_2) \end{bmatrix}$$

Furthermore:

$$\begin{aligned} [T_2][T_1] &= \begin{bmatrix} \cos \theta_2 & -\sin \theta_2 \\ \sin \theta_2 & \cos \theta_2 \end{bmatrix} \begin{bmatrix} \cos \theta_1 & -\sin \theta_1 \\ \sin \theta_1 & \cos \theta_1 \end{bmatrix} \\ &= \begin{bmatrix} \cos \theta_2 \cos \theta_1 - \sin \theta_2 \sin \theta_1 & -(\cos \theta_2 \sin \theta_1 + \sin \theta_2 \cos \theta_1) \\ \sin \theta_2 \cos \theta_1 + \cos \theta_2 \sin \theta_1 & -\sin \theta_2 \sin \theta_1 + \cos \theta_2 \cos \theta_1 \end{bmatrix} \\ &= \begin{bmatrix} \cos(\theta_1 + \theta_2) & -\sin(\theta_1 + \theta_2) \\ \sin(\theta_1 + \theta_2) & \cos(\theta_1 + \theta_2) \end{bmatrix} \\ &= [T_2 \circ T_1] \end{aligned}$$

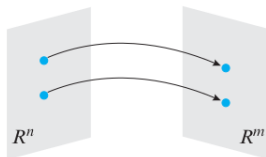
It can be easily seen that  $[T_2 \circ T_1] = [T_1 \circ T_2]$  (hence, commutative).

# Exercise

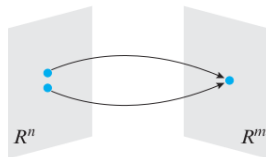
Read Example 3 and Example 4 (page 272-273)

# One-to-one matrix transformation

A matrix transformation  $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is said to be **one-to-one** if  $T_A$  maps distinct vectors (points) in  $\mathbb{R}^n$  into distinct vectors (points) in  $\mathbb{R}^m$ .



One-to-one



Not one-to-one

Equivalent statements:

- $T_A$  is one-to-one if  $\forall \mathbf{b}$  in the range of  $A$ , there is exactly one vector  $\mathbf{x} \in \mathbb{R}^n$ , s.t.  $T_A \mathbf{x} = \mathbf{b}$ .
- $T_A$  is one-to-one if the equality  $T_A(\mathbf{u}) = T_A(\mathbf{v})$  implies that  $\mathbf{u} = \mathbf{v}$ .

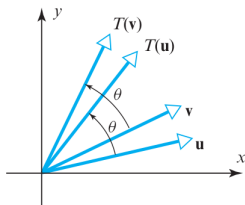
## Examples: one-to-one and not one-to-one transformations

*Rotation operators on  $\mathbb{R}^2$  are one-to-one.*

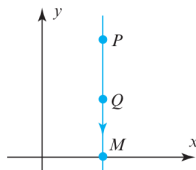
since distinct vectors that are rotated through the same angle have distinct images.

*The orthogonal projection of  $\mathbb{R}^2$  onto the  $x$ -axis is not one-to-one.*

since it maps distinct points on the same vertical line into the same point.



▲ **Figure 4.10.6** Distinct vectors  $\mathbf{u}$  and  $\mathbf{v}$  are rotated into distinct vectors  $T(\mathbf{u})$  and  $T(\mathbf{v})$ .



▲ **Figure 4.10.7** The distinct points  $P$  and  $Q$  are mapped into the same point  $M$ .

## Kernel and range

If  $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a matrix transformation, then the set of all vectors in  $\mathbb{R}^n$  that  $T_A$  maps into 0 is called the **kernel of  $T_A$**  and is denoted by  $\ker(T_A)$ , i.e.:

$$\ker(T_A) = \{\mathbf{x} \in \mathbb{R}^n \text{ s.t. } A\mathbf{x} = \mathbf{0}\}$$

The set of all vectors in  $\mathbb{R}^m$  that are images under this transformation of at least one vector in  $\mathbb{R}^n$  is called the **range of  $T_A$**  and is denoted by  $R(T_A)$ , i.e.:

$$R(T_A) = \{\mathbf{b} \in \mathbb{R}^m \text{ s.t. } \exists \mathbf{x} \in \mathbb{R}^n, \text{ where } A\mathbf{x} = \mathbf{b}\}$$

In brief:

$$\ker(T_A) = \text{null space of } A$$

$$R(T_A) = \text{column space of } A$$

# Matrix - linear system - transformation

Let  $A$  be an  $(m \times n)$  matrix.

Three ways of viewing the same subspace of  $\mathbb{R}^n$ :

- **Matrix view:** the null space of  $A$
- **System view:** the solution space of  $Ax = 0$
- **Transformation view:** the kernel of  $T_A$

Three ways of viewing the same subspace of  $\mathbb{R}^m$ :

- **Matrix view:** the column space of  $A$
- **System view:** all  $\mathbf{b} \in \mathbb{R}^m$  for which  $Ax = \mathbf{b}$  is consistent
- **Transformation view:** the range of  $T_A$



# Exercise

Read Example 5 and Example 6 on page 275.

## One-to-one matrix operator

Let  $T_A : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a one-to-one matrix operator. So,  $A$  is invertible.

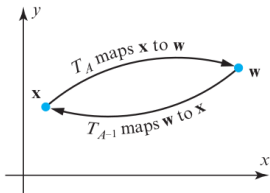
The **inverse operator** or the **inverse** of  $T_A$  is defined as:

$$T_{A^{-1}} : \mathbb{R}^n \rightarrow \mathbb{R}^n$$

In this case:

$$T_A(T_{A^{-1}}(\mathbf{x})) = AA^{-1}\mathbf{x} = I\mathbf{x} = \mathbf{x} \quad \text{or, equivalently} \quad T_A \circ T_{A^{-1}} = T_{AA^{-1}} = T_I$$

$$T_{A^{-1}}(T_A(\mathbf{x})) = A^{-1}A\mathbf{x} = I\mathbf{x} = \mathbf{x} \quad \text{or, equivalently} \quad T_{A^{-1}} \circ T_A = T_{A^{-1}A} = T_I$$



$T_A$  maps  $\mathbf{x}$  to  $\mathbf{w}$  and  $T_{A^{-1}}$  maps  $\mathbf{w}$  back to  $\mathbf{x}$ , i.e.,  $T_{A^{-1}}(\mathbf{w}) = T_{A^{-1}}(T_A(\mathbf{x})) = \mathbf{x}$

# Exercise

Read Example 7 and Example 8 on page 276.

# Conclusion

## THEOREM 4.10.2 Equivalent Statements

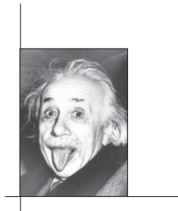
If  $A$  is an  $n \times n$  matrix, then the following statements are equivalent.

- (a)  $A$  is invertible.
- (b)  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.
- (c) The reduced row echelon form of  $A$  is  $I_n$ .
- (d)  $A$  is expressible as a product of elementary matrices.
- (e)  $A\mathbf{x} = \mathbf{b}$  is consistent for every  $n \times 1$  matrix  $\mathbf{b}$ .
- (f)  $A\mathbf{x} = \mathbf{b}$  has exactly one solution for every  $n \times 1$  matrix  $\mathbf{b}$ .
- (g)  $\det(A) \neq 0$ .
- (h) The column vectors of  $A$  are linearly independent.
- (i) The row vectors of  $A$  are linearly independent.
- (j) The column vectors of  $A$  span  $\mathbb{R}^n$ .
- (k) The row vectors of  $A$  span  $\mathbb{R}^n$ .
- (l) The column vectors of  $A$  form a basis for  $\mathbb{R}^n$ .
- (m) The row vectors of  $A$  form a basis for  $\mathbb{R}^n$ .
- (n)  $A$  has rank  $n$ .
- (o)  $A$  has nullity 0.
- (p) The orthogonal complement of the null space of  $A$  is  $\mathbb{R}^n$ .
- (q) The orthogonal complement of the row space of  $A$  is  $\{\mathbf{0}\}$ .
- (r) The kernel of  $T_A$  is  $\{\mathbf{0}\}$ .
- (s) The range of  $T_A$  is  $\mathbb{R}^n$ .
- (t)  $T_A$  is one-to-one.

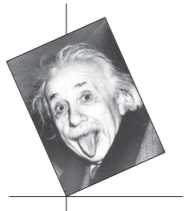
# Geometry of Matrix Operators on $\mathbb{R}^2$

(page 280 of Elementary LA Applications book)

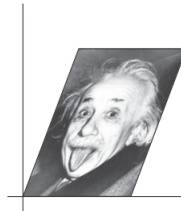
*to be continued...*



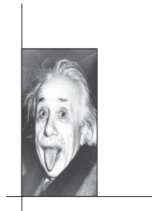
Digitized scan



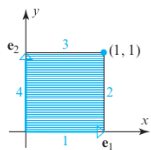
Rotated



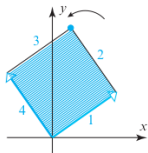
Sheared horizontally



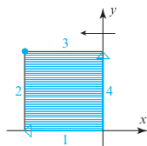
Compressed horizontally



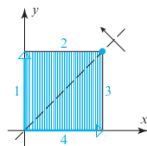
Unit square



Unit square rotated



Unit square reflected about the  $y$ -axis



Unit square reflected about the line  $y = x$



## Exercise

Given a transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  which is multiplication by an invertible matrix. Determine the image of:

1. A straight line
2. A line through the origin
3. Parallel lines
4. The line segment joining points  $P$  and  $Q$
5. Three points lie on a line

### Task:

*Divide yourselves into 5 groups, and examine each of the question!*

# Exercises

## Question 1

Given a transformation matrix:

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

Find the image of line  $y = 2x + 1$  under the transformation.

## Question 2

Given a transformation matrix:

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

Find the image of the unit square on the *first quadrant* under the transformation.

## Exercises

Determine the image of the unit square under the following transformation:

- Reflection about the  $y$ -axis
- Reflection about the  $x$ -axis
- Reflection about the line  $y = x$
- Rotation about the origin through a positive angle  $\theta$
- Compression in the  $x$ -direction with factor  $k$  with  $0 < k < 1$
- Compression in the  $y$ -direction with factor  $k$  with  $0 < k < 1$
- Expansion in the  $x$ -direction with factor  $k$  with  $k > 1$
- Expansion in the  $y$ -direction with factor  $k$  with  $k > 1$
- Shear in the  $x$ -direction with factor  $k$  with  $k > 0$
- Shear in the  $x$ -direction with factor  $k$  with  $k < 0$
- Shear in the  $y$ -direction with factor  $k$  with  $k > 0$
- Shear in the  $y$ -direction with factor  $k$  with  $k < 0$

*This is the end of slide...*