

# Lecture 7: Linear Transformations

A transformation / function / map

$$T: \mathbb{R}^n \rightarrow \mathbb{R}^m$$

is a rule assigning, to each vector in  $\mathbb{R}^n$ , a vector in  $\mathbb{R}^m$ :

$$\vec{x} \longmapsto T(\vec{x}).$$

The range / image of  $T$  is

$$T(\mathbb{R}^n) = \{T(\vec{x}) \mid \vec{x} \in \mathbb{R}^n\} \subset \mathbb{R}^m,$$

the set of all vectors in the image of  $T$ .

## Examples

①  $T: \mathbb{R}^2 \rightarrow \mathbb{R}$  defined by  $T\begin{pmatrix} x \\ y \end{pmatrix} := x^2 + y^2$ .

②  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  "  $T\begin{pmatrix} x \\ y \end{pmatrix} := \begin{pmatrix} x \\ -y \end{pmatrix}$

③  $T: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  "  $T\begin{pmatrix} x \\ y \end{pmatrix} := \begin{pmatrix} y^{-1} \\ x+y \end{pmatrix}$

④  $T: \mathbb{R}^3 \rightarrow \mathbb{R}^3$  "  $T\begin{pmatrix} x \\ y \\ z \end{pmatrix} := \begin{pmatrix} x \\ y \\ 0 \end{pmatrix}$

(What are these geometrically?)

A transformation is linear if it satisfies

$$(*) \left. \begin{array}{l} T(c\vec{v}) = cT(\vec{v}) \\ T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v}) \end{array} \right\} \text{for all } c \in \mathbb{R} \text{ and } \vec{u}, \vec{v} \in \mathbb{R}^n.$$

(Equivalently, you can check that  $T(c\vec{u} + d\vec{v}) = cT(\vec{u}) + dT(\vec{v})$ .)

An immediate consequence of (\*) is that  $T(\vec{0}) = \vec{0}$ :

$$T(\vec{0}) = T(0 \cdot \vec{0}) = 0 \cdot T(\vec{0}) = \vec{0}. \\ \text{(take } c = 0 \text{ in } (*))$$

### Examples (cont'd.)

(2) is linear since  $T\left(c\begin{pmatrix} u_1 \\ u_2 \end{pmatrix} + d\begin{pmatrix} v_1 \\ v_2 \end{pmatrix}\right) = T\left(\begin{pmatrix} cu_1 + dv_1 \\ cu_2 + dv_2 \end{pmatrix}\right)$   
 $= \begin{pmatrix} cu_1 + dv_1 \\ -(cu_2 + dv_2) \end{pmatrix} = \begin{pmatrix} cu_1 \\ -cu_2 \end{pmatrix} + \begin{pmatrix} dv_1 \\ -dv_2 \end{pmatrix} = c\begin{pmatrix} u_1 \\ -u_2 \end{pmatrix} + d\begin{pmatrix} v_1 \\ -v_2 \end{pmatrix}$   
 $= cT\begin{pmatrix} u_1 \\ u_2 \end{pmatrix} + dT\begin{pmatrix} v_1 \\ v_2 \end{pmatrix}.$

(4) is linear (similar reasoning)

(3) is not because  $T(\vec{0}) = \begin{pmatrix} -1 \\ 0 \end{pmatrix} \neq \vec{0}$ . This may seem strange: the basic point is that a linear transformation from  $\mathbb{R}$  to  $\mathbb{R}$  is of the form  $x \mapsto ax$ , not  $x \mapsto ax + b$ .

(1) is also non linear; this is called affine rather than linear  
while  $T(\vec{0}) = 0$ ,  $T(2\begin{pmatrix} 1 \\ 0 \end{pmatrix}) = T\begin{pmatrix} 2 \\ 0 \end{pmatrix} = 2^2 + 0^2 = 4 \neq 2 = 2T\begin{pmatrix} 1 \\ 0 \end{pmatrix}$ .

Ex / Which are linear / nonlinear transformations from  $\mathbb{R}^2 \rightarrow \mathbb{R}^2$ :

$$S \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} y \\ x \end{pmatrix}, \quad T \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} xy \\ x \end{pmatrix}, \quad R \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x+y \\ y \end{pmatrix}.$$

Answer: the middle one is nonlinear:

$$T \begin{pmatrix} 1 \\ 0 \end{pmatrix} + T \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} + \begin{pmatrix} 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \neq \begin{pmatrix} 1 \\ 1 \end{pmatrix} = T \begin{pmatrix} 1 \\ 1 \end{pmatrix} = T \left( \begin{pmatrix} 1 \\ 0 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right) //$$

A matrix transformation is a transformation of the form

$$T(\vec{x}) := A \vec{x}.$$

In Lecture 4, we checked that

$$A(c\vec{u} + d\vec{v}) = cA\vec{u} + dA\vec{v};$$

that is, matrix transformations are linear.

Note that in the matrix-vector product  $A\vec{x}$ , an  $m \times n$  matrix must be multiplied by an  $n \times 1$  vector, yielding an  $m \times 1$  vector. That is, the matrix transformation maps  $\mathbb{R}^n \rightarrow \mathbb{R}^m$  if  $A$  is  $m \times n$ .

Ex / Let  $A = \begin{pmatrix} 2 & -3 & 1 \\ 1 & 0 & -1 \end{pmatrix}$ , and let  $T: \mathbb{R}^3 \rightarrow \mathbb{R}^2$  be the corresponding transformation:  $T \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2 & -3 & 1 \\ 1 & 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} 2x - 3y + z \\ x - z \end{pmatrix} //$

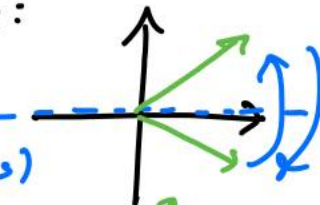
# Examples (2) and (4)

I claim that we can

write these as matrix transformations:

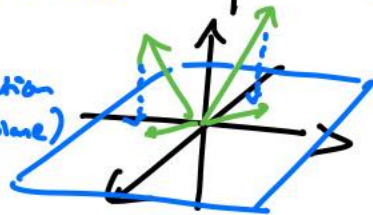
$$\text{and } \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} x \\ -y \end{pmatrix}$$

(reflection about x-axis)



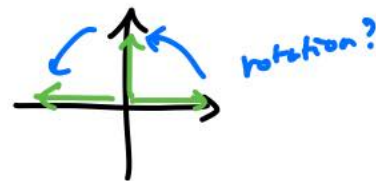
$$\begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{pmatrix} x \\ y \\ 0 \end{pmatrix}$$

(projection to xy-plane)

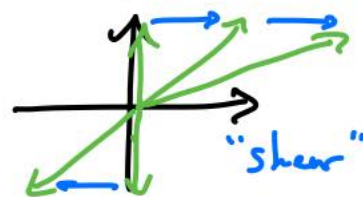


What about  $S$  &  $R$  from the example on the last page?

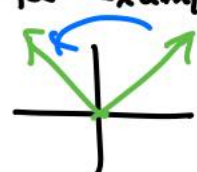
$$\begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \begin{pmatrix} -y \\ x \end{pmatrix} = S \begin{pmatrix} x \\ y \end{pmatrix}$$



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



(Is  $S$  a rotation? Sure looks like it: for example,

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


How does one go about finding these matrices for a linear transformation? Can one always do this?

Those are questions for the next lecture! But this would be useful, because then we could rephrase questions like "is  $\vec{y}$  in the range of  $T$ " as "is  $\vec{y} = A\vec{x}$  consistent" which we already know how to solve.

It's natural to ask what happens to linear (in)dependence under a linear transformation.

If  $c_1\vec{v}_1 + \dots + c_n\vec{v}_n = \vec{0}$  is a dependence relation on  $\{\vec{v}_1, \dots, \vec{v}_n\}$ ,

$$\vec{0} = T(\vec{0}) = T(c_1\vec{v}_1 + \dots + c_n\vec{v}_n) = c_1T(\vec{v}_1) + \dots + c_nT(\vec{v}_n)$$

is a dependence relation on  $\{T(\vec{v}_1), \dots, T(\vec{v}_n)\}$ . So

" $T$  preserves dependency". Does it preserve independence?

To see that the answer is NO for an arbitrary transformation, just consider the projection above:

