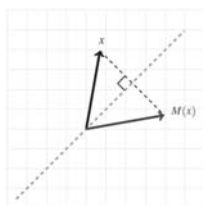


# Linear Transformations and Matrices

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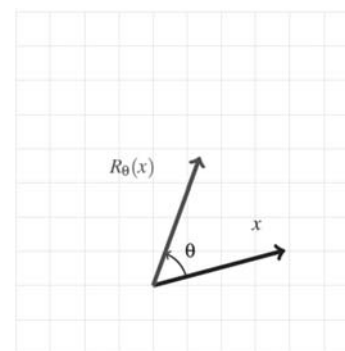
## Example

- A reflection with respect to a 45 degree line is illustrated by
  - Think of the dashed green line as a mirror and  $M: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  as the vector function that maps a vector to its mirror image.
  - If  $x, y \in \mathbb{R}^n$  and  $\alpha \in \mathbb{R}$ , then  $M(\alpha x) = \alpha M(x)$  and  $M(x + y) = M(x) + M(y)$  (in other words,  $M$  is a linear transformation).
    - True
    - False



## Rotating in 2D

- Let  $R_\theta: \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the function that rotates an input vector through an angle  $\theta$



- $R_\theta(\alpha x)$
- $\alpha R_\theta(x)$
- $R_\theta(x + y)$
- $R_\theta(x) + R_\theta(y)$

## Linear Transformations

- What Makes Linear Transformations so Special?
  - Many problems in science and engineering involve vector functions such as:  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ . Given such a function, one often wishes to do the following:
    - Given vector  $x \in \mathbb{R}^n$ , evaluate  $f(x)$ ; or
    - Given vector  $y \in \mathbb{R}^m$ , find  $x$  such that  $f(x) = y$ ; or
    - Find scalar  $\lambda$  and vector  $x$  such that  $f(x) = \lambda x$  (only if  $m = n$ ).

## Linear Transformations

- What is a Linear Transformation?

- Definition 2.1

- A vector function  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is said to be a linear transformation, if for all  $x, y \in \mathbb{R}^n$  and  $\alpha \in \mathbb{R}$

- Transforming a scaled vector is the same as scaling the transformed vector:

$$L(\alpha x) = \alpha L(x)$$

- Transforming the sum of two vectors is the same as summing the two transformed vectors:

$$L(x + y) = L(x) + L(y)$$

## Example

- The transformation  $f\left(\begin{pmatrix} x_0 \\ x_1 \end{pmatrix}\right) = \begin{pmatrix} x_0 + x_1 \\ x_0 \end{pmatrix}$  is a linear transformation?

## Example

- The transformation  $f\left(\begin{pmatrix} x \\ \psi \end{pmatrix}\right) = \begin{pmatrix} x + \psi \\ x + 1 \end{pmatrix}$  is a linear transformation?

## Exercises

- The vector function  $f\left(\begin{pmatrix} x \\ \psi \end{pmatrix}\right) = \begin{pmatrix} x\psi \\ x \end{pmatrix}$  is a linear transformation.
  - TRUE / FALSE
- The vector function  $f\left(\begin{pmatrix} x_0 \\ x_1 \\ x_2 \end{pmatrix}\right) = \begin{pmatrix} x_0 + 1 \\ x_1 + 2 \\ x_2 + 3 \end{pmatrix}$  is a linear transformation.
  - TRUE / FALSE
- The vector function  $f\left(\begin{pmatrix} x_0 \\ x_1 \\ x_2 \end{pmatrix}\right) = \begin{pmatrix} x_0 \\ x_0 + x_1 \\ x_0 + x_1 + x_2 \end{pmatrix}$  is a linear transformation.
  - TRUE / FALSE

## Exercises

- If  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear transformation, then  $L(0) = 0$ . (Recall that 0 equals a vector with zero components of appropriate size.)
  - Always / Sometime / Never
- If  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $f(0) \neq 0$ , Then  $f$  is not a linear transformation.
  - True/False
- If  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  and  $f(0) = 0$ , Then  $f$  is a linear transformation.
  - Always / Sometime / Never

## Exercises

- Find an example of a function  $f$  such that  $f(\alpha x) = \alpha f(x)$ , but for some  $x, y$  it is the case that  $f(x + y) \neq f(x) + f(y)$ . (This is pretty tricky!)
- The vector function  $f\left(\begin{pmatrix} x_0 \\ x_1 \end{pmatrix}\right) = \begin{pmatrix} x_1 \\ x_0 \end{pmatrix}$  is a linear transformation.
  - TRUE / FALSE

## Linear Transformations

- Of Linear Transformations and Linear Combinations
  - Lemma 2.4
    - $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear transformation if and only if (iff) for all  $u, v \in \mathbb{R}^n$  and  $\alpha, \beta \in \mathbb{R}$ 

$$L(\alpha u + \beta v) = \alpha L(u) + \beta L(v)$$
  - Lemma 2.5
    - Let  $v_0, v_1, \dots, v_{k-1} \in \mathbb{R}^n$  and let  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear transformation. Then
 
$$L(v_0 + v_1 + \dots + v_{k-1}) = L(v_0) + L(v_1) + \dots + L(v_{k-1})$$

## Mathematical Induction



- What is the Principle of Mathematical Induction?
  - The Principle of Mathematical Induction (weak induction) says that if one can show that
    - (Base case) a property holds for  $n = k_b$ ; and
    - (Inductive step) if it holds for  $n = K$ , where  $K \geq k_b$ , then it is also holds for  $n = K + 1$ ,
  - then one can conclude that the property holds for all integers  $n \geq k_b$ .
  - Often  $k_b = 0$  or  $k_b = 1$ .

## Mathematical Induction

### • Examples

$$\sum_{i=0}^{n-1} i = \frac{n(n-1)}{2}; n \geq 1$$

— Base case :  $n = 1$

— Inductive step : Inductive Hypothesis (IH)

- Assume that the result is true for  $n = k$  where  $k \geq 1$
- Show that the result is then also true for  $n = k + 1$

— Find :  $2 \sum_{i=0}^{n-1} i$

## Exercises

- Let  $n \geq 1$ . Then  $\sum_{i=1}^n i = n(n+1)/2$ .  
— Always / Sometimes / Never
- Let  $n \geq 1$ . Then  $\sum_{i=0}^{n-1} 1 = n$ .  
— Always / Sometimes / Never
- Let  $n \geq 1$  and  $x \in \mathbb{R}^m$ . Then  $\sum_{i=0}^{n-1} x = x + x + \dots + x = nx$ .  
— Always / Sometimes / Never
- Let  $n \geq 1$ .  $\sum_{i=0}^{n-1} i^2 = (n-1)n(2n-1)/6$ .  
— Always / Sometimes / Never

## Representing Linear Transformations as Matrices

### • From Linear Transformation to Matrix-Vector Multiplication

— Theorem 2.6

- Let  $v_0, v_1, \dots, v_{n-1} \in \mathbb{R}^n$ ,  $\alpha_0, \alpha_1, \dots, \alpha_{n-1} \in \mathbb{R}$ , and let  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear transformation.

• Then

$$L(\alpha_0 v_0 + \alpha_1 v_1 + \dots + \alpha_{n-1} v_{n-1}) = \alpha_0 L(v_0) + \alpha_1 L(v_1) + \dots + \alpha_{n-1} L(v_{n-1})$$

## Exercises

Homework 2.4.1.2 Let  $L$  be a linear transformation such that

$$L\left(\begin{pmatrix} 1 \\ 0 \end{pmatrix}\right) = \begin{pmatrix} 3 \\ 5 \end{pmatrix} \quad \text{and} \quad L\left(\begin{pmatrix} 0 \\ 1 \end{pmatrix}\right) = \begin{pmatrix} 2 \\ -1 \end{pmatrix}.$$

Then  $L\left(\begin{pmatrix} 2 \\ 3 \end{pmatrix}\right) =$

For the next three exercises, let  $L$  be a linear transformation such that

$$L\left(\begin{pmatrix} 1 \\ 0 \end{pmatrix}\right) = \begin{pmatrix} 3 \\ 5 \end{pmatrix} \quad \text{and} \quad L\left(\begin{pmatrix} 1 \\ 1 \end{pmatrix}\right) = \begin{pmatrix} 5 \\ 4 \end{pmatrix}.$$

Homework 2.4.1.3  $L\left(\begin{pmatrix} 3 \\ 3 \end{pmatrix}\right) =$

Homework 2.4.1.4  $L\left(\begin{pmatrix} -1 \\ 0 \end{pmatrix}\right) =$

Homework 2.4.1.5  $L\left(\begin{pmatrix} 2 \\ 3 \end{pmatrix}\right) =$

## Representing Linear Transformations as Matrices

Now we are ready to link linear transformations to matrices and matrix-vector multiplication. Recall that any vector  $x \in \mathbb{R}^n$  can be written as

$$x = \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{n-1} \end{pmatrix} = x_0 \underbrace{\begin{pmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{pmatrix}}_{e_0} + x_1 \underbrace{\begin{pmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{pmatrix}}_{e_1} + \cdots + x_{n-1} \underbrace{\begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{pmatrix}}_{e_{n-1}} = \sum_{j=0}^{n-1} x_j e_j.$$

Let  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  be a linear transformation. Given  $x \in \mathbb{R}^n$ , the result of  $y = L(x)$  is a vector in  $\mathbb{R}^m$ . But then

$$y = L(x) = L\left(\sum_{j=0}^{n-1} x_j e_j\right) = \sum_{j=0}^{n-1} x_j L(e_j) = \sum_{j=0}^{n-1} x_j a_j,$$

where we let  $a_j = L(e_j)$ .

**The Big Idea.** The linear transformation  $L$  is completely described by the vectors

$$a_0, a_1, \dots, a_{n-1}, \quad \text{where } a_j = L(e_j)$$

because for any vector  $x$ ,  $L(x) = \sum_{j=0}^{n-1} x_j a_j$ .

By arranging these vectors as the columns of a two-dimensional array, which we call the matrix  $A$ , we arrive at the observation that the matrix is simply a representation of the corresponding linear transformation  $L$ .

## Representing Linear Transformations as Matrices

### —Definition 2.7 ( $\mathbb{R}^{m \times n}$ )

- The set of all  $m \times n$  real valued matrices is denoted by  $\mathbb{R}^{m \times n}$ .
- Thus,  $A \in \mathbb{R}^{m \times n}$  means that  $A$  is a real valued matrix of size  $m \times n$ .

## Representing Linear Transformations as Matrices

### —Definition 2.8 (Matrix-vector multiplication or product)

- Let  $A \in \mathbb{R}^{m \times n}$  and  $x \in \mathbb{R}^n$  with

$$A = \begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \cdots & \alpha_{0,n-1} \\ \alpha_{1,0} & \alpha_{1,1} & \cdots & \alpha_{1,n-1} \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_{m-1,0} & \alpha_{m-1,1} & \cdots & \alpha_{m-1,n-1} \end{pmatrix} \quad \text{and} \quad x = \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{n-1} \end{pmatrix}$$

- then

$$\begin{pmatrix} \alpha_{0,0} & \alpha_{0,1} & \cdots & \alpha_{0,n-1} \\ \alpha_{1,0} & \alpha_{1,1} & \cdots & \alpha_{1,n-1} \\ \vdots & \vdots & \ddots & \vdots \\ \alpha_{m-1,0} & \alpha_{m-1,1} & \cdots & \alpha_{m-1,n-1} \end{pmatrix} \begin{pmatrix} x_0 \\ x_1 \\ \vdots \\ x_{n-1} \end{pmatrix} = \begin{pmatrix} \alpha_{0,0}x_0 + \alpha_{0,1}x_1 + \cdots + \alpha_{0,n-1}x_{n-1} \\ \alpha_{1,0}x_0 + \alpha_{1,1}x_1 + \cdots + \alpha_{1,n-1}x_{n-1} \\ \vdots \\ \alpha_{m-1,0}x_0 + \alpha_{m-1,1}x_1 + \cdots + \alpha_{m-1,n-1}x_{n-1} \end{pmatrix}$$

## Representing Linear Transformations as Matrices

### • Practice with Matrix-Vector Multiplication

**Homework 2.4.2.1** Compute  $Ax$  when  $A = \begin{pmatrix} -1 & 0 & 2 \\ -3 & 1 & -1 \\ -2 & -1 & 2 \end{pmatrix}$  and  $x = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$ .

**Homework 2.4.2.2** Compute  $Ax$  when  $A = \begin{pmatrix} -1 & 0 & 2 \\ -3 & 1 & -1 \\ -2 & -1 & 2 \end{pmatrix}$  and  $x = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ .

**Homework 2.4.2.3** If  $A$  is a matrix and  $e_j$  is a unit basis vector of appropriate length, then  $Ae_j = a_j$ , where  $a_j$  is the  $j$ th column of matrix  $A$ .

Always/Sometimes/Never

**Homework 2.4.2.4** If  $x$  is a vector and  $e_i$  is a unit basis vector of appropriate size, then their dot product,  $e_i^T x$ , equals the  $i$ th entry in  $x$ ,  $x_i$ .

Always/Sometimes/Never

**Homework 2.4.2.5** Compute

$$\begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}^T \left( \begin{pmatrix} -1 & 0 & 2 \\ -3 & 1 & -1 \\ -2 & -1 & 2 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \right) = \underline{\hspace{2cm}}$$

## Representing Linear Transformations as Matrices

### • It Goes Both Ways

#### — Theorem 2.9

- Let  $L: \mathbb{R}^n \rightarrow \mathbb{R}^m$  be defined by  $L(x) = Ax$  where  $A \in \mathbb{R}^{m \times n}$ . Then  $L$  is a linear transformation.
- A function  $f: \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear transformation if and only if it can be written as a matrix-vector multiplication.

## Exercises

**Homework 2.4.3.1** Give the linear transformation that corresponds to the matrix

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

**Homework 2.4.3.2** Give the linear transformation that corresponds to the matrix

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

## Representing Linear Transformations as Matrices

### • Example 2.10 (from 2.2)

— The transformation  $f\left(\begin{pmatrix} x_0 \\ x_1 \end{pmatrix}\right) = \begin{pmatrix} x_0 + x_1 \\ x_0 \end{pmatrix}$  is a linear transformation.

### • Example 2.11 (from 2.3)

— The transformation  $f\left(\begin{pmatrix} x \\ \psi \end{pmatrix}\right) = \begin{pmatrix} x + \psi \\ x + 1 \end{pmatrix}$  is not a linear transformation.

## Exercises

**Homework 2.4.3.3** Let  $f$  be a vector function such that  $f\left(\begin{pmatrix} x_0 \\ x_1 \end{pmatrix}\right) = \begin{pmatrix} x_0^2 \\ x_1 \end{pmatrix}$ . Then

- (a)  $f$  is a linear transformation.
- (b)  $f$  is not a linear transformation.
- (c) Not enough information is given to determine whether  $f$  is a linear transformation.

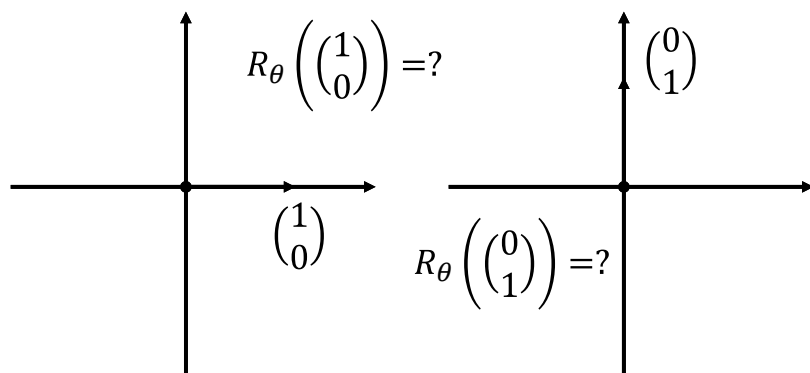
How do you know?

**Homework 2.4.3.4** For each of the following, determine whether it is a linear transformation or not:

- $f\left(\begin{pmatrix} x_0 \\ x_1 \\ x_2 \end{pmatrix}\right) = \begin{pmatrix} x_0 \\ 0 \\ x_2 \end{pmatrix}.$
- $f\left(\begin{pmatrix} x_0 \\ x_1 \end{pmatrix}\right) = \begin{pmatrix} x_0^2 \\ 0 \end{pmatrix}.$

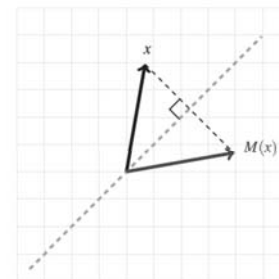
## Representing Linear Transformations as Matrices

- Rotations and Reflections, Revisited



## Representing Linear Transformations as Matrices

Homework 2.4.4.2 A reflection with respect to a 45 degree line is illustrated by



Again, think of the dashed green line as a mirror and let  $M : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  be the vector function that maps a vector to its mirror image. Compute the matrix that represents  $M$  (by examining the picture).

## Questions and Answers

