

## Section 1.7: Linear Independence

**Definition:** An indexed set of vectors  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  in  $\mathbb{R}^n$  is said to be **linearly independent** if the vector equation

$$x_1\mathbf{v}_1 + x_2\mathbf{v}_2 + \cdots + x_p\mathbf{v}_p = \mathbf{0}$$

has only the trivial solution.

The set  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  is said to be **linearly dependent** if there exists a set of weights  $c_1, c_2, \dots, c_p$  *at least one of which is nonzero* such that

$$c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_p\mathbf{v}_p = \mathbf{0}.$$

(i.e. Provided the homogeneous equation possesses a nontrivial solution.)

An equation  $c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \cdots + c_p\mathbf{v}_p = \mathbf{0}$ , with at least one  $c_i \neq 0$ , is called a **linear dependence relation**.

## Three Theorems on Linear Independence

**Theorem:** The columns of a matrix  $A$  are linearly **independent** if and only if the homogeneous equation  $A\mathbf{x} = \mathbf{0}$  has only the trivial solution.

**Theorem:** If a set contains more vectors than there are entries in each vector, then the set is linearly **dependent**. That is, if  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  is a set of vector in  $\mathbb{R}^n$ , and  $p > n$ , then the set is linearly dependent.

**Theorem:** Any set of vectors that contains the zero vector is linearly **dependent**.

Determine if the set is linearly dependent or linearly independent

$$(a) \left\{ \begin{bmatrix} 1 \\ 2 \\ -1 \end{bmatrix}, \begin{bmatrix} 3 \\ 3 \\ -5 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \\ 3 \end{bmatrix} \right\}$$

4 vectors in  $\mathbb{R}^3$   $4 > 3$

They are lin. dependent.

Determine if the set is linearly dependent or linearly independent

↖ 0

$$(b) \left\{ \begin{bmatrix} 2 \\ 2 \\ 2 \\ 0 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 4 \\ -8 \\ 1 \end{bmatrix} \right\}$$

Set contains the zero vector. It is linearly dependent.

## Section 1.8: Intro to Linear Transformations

Recall that the product  $A\mathbf{x}$  is a linear combination of the columns of  $A$ —turns out to be a vector. If the columns of  $A$  are vectors in  $\mathbb{R}^m$ , and there are  $n$  of them, then

- ▶  $A$  is an  $m \times n$  matrix,
- ▶ the product  $A\mathbf{x}$  is defined for  $\mathbf{x}$  in  $\mathbb{R}^n$ , and
- ▶ the vector  $\mathbf{b} = A\mathbf{x}$  is a vector in  $\mathbb{R}^m$ .

So we can think of  $A$  as an **object that acts** on vectors  $\mathbf{x}$  in  $\mathbb{R}^n$  (via the product  $A\mathbf{x}$ ) to produce vectors  $\mathbf{b}$  in  $\mathbb{R}^m$ .

## Transformation from $\mathbb{R}^n$ to $\mathbb{R}^m$

**Definition:** A transformation  $T$  (a.k.a. **function** or **mapping**) from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  is a rule that assigns to each vector  $\mathbf{x}$  in  $\mathbb{R}^n$  a vector  $T(\mathbf{x})$  in  $\mathbb{R}^m$ .

**Function Notation:** If a transformation  $T$  takes a vector  $\mathbf{x}$  in  $\mathbb{R}^n$  and maps it to a vector  $T(\mathbf{x})$  in  $\mathbb{R}^m$ , we can write

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

which reads “ $T$  maps  $\mathbb{R}^n$  into  $\mathbb{R}^m$ .”

And we can write

$$\vec{x} \mapsto T(\mathbf{x})$$

which reads “ $\mathbf{x}$  maps to  $T$  of  $\mathbf{x}$ .”

# Terms and Notation

For  $T : \mathbb{R}^n \longrightarrow \mathbb{R}^m$ ,

- ▶  $\mathbb{R}^n$  is the **domain**, and
- ▶  $\mathbb{R}^m$  is called the **codomain**.
- ▶ For  $\mathbf{x}$  in the domain,  $T(\mathbf{x})$  is called the **image** of  $\mathbf{x}$  under  $T$ . (We can call  $\mathbf{x}$  a **pre-image** of  $T(\mathbf{x})$ .)
- ▶ The collection of all images is called the **range**.
- ▶ If  $T(\mathbf{x})$  is defined by multiplication by the  $m \times n$  matrix  $A$ , we may denote this by  $\mathbf{x} \mapsto A\mathbf{x}$ .

## Matrix Transformation Example

Let  $A = \begin{bmatrix} 1 & 3 \\ 2 & 4 \\ 0 & -2 \end{bmatrix}$ . Define the transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^3$  by the mapping  $T(\mathbf{x}) = A\mathbf{x}$ .

(a) Find the image of the vector  $\mathbf{u} = \begin{bmatrix} 1 \\ -3 \end{bmatrix}$  under  $T$ .

$$\begin{aligned} T(\mathbf{u}) &= A\mathbf{u} \\ &= \begin{bmatrix} 1 & 3 \\ 2 & 4 \\ 0 & -2 \end{bmatrix} \begin{bmatrix} 1 \\ -3 \end{bmatrix} = \begin{bmatrix} 1-9 \\ 2-12 \\ 0+6 \end{bmatrix} = \begin{bmatrix} -8 \\ -10 \\ 6 \end{bmatrix} \end{aligned}$$



$$A = \begin{bmatrix} 1 & 3 \\ 2 & 4 \\ 0 & -2 \end{bmatrix}$$

(b) Determine a vector  $\mathbf{x}$  in  $\mathbb{R}^2$  whose image under  $T$  is  $\begin{bmatrix} -4 \\ -4 \\ 4 \end{bmatrix}$ .

Find  $\vec{x}$  such that  $T(\vec{x}) = \begin{bmatrix} -4 \\ -4 \\ 4 \end{bmatrix}$

Solve  $A\vec{x} = \vec{b}$  where  $\vec{b} = \begin{bmatrix} -4 \\ -4 \\ 4 \end{bmatrix}$ .

Using an augmented matrix

$$[A \ \vec{b}] = \begin{bmatrix} 1 & 3 & -4 \\ 2 & 4 & -4 \\ 0 & -2 & 4 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & -2 \\ 0 & 0 & 0 \end{bmatrix} \quad \begin{array}{l} x_1 = 2 \\ x_2 = -2 \end{array}$$

A pre-image  $\vec{x} = \begin{bmatrix} 2 \\ -2 \end{bmatrix}$ .

$$A = \begin{bmatrix} 1 & 3 \\ 2 & 4 \\ 0 & -2 \end{bmatrix}$$

(c) Determine if  $\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$  is in the range of  $T$ .

Is  $T(\vec{x}) = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$  solvable?  $A\vec{x} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ ,

we can use an augmented matrix

$$\begin{bmatrix} 1 & 3 & 1 \\ 2 & 4 & 0 \\ 0 & -2 & 1 \end{bmatrix} \xrightarrow{\text{rref}} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

pivot column!

$A\vec{x} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$  is consistent

$\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$  is not in the range of  $T$ .

# Linear Transformations

**Definition:** A transformation  $T$  is **linear** provided

- (i)  $T(\mathbf{u} + \mathbf{v}) = T(\mathbf{u}) + T(\mathbf{v})$  for every  $\mathbf{u}, \mathbf{v}$  in the domain of  $T$ , and
- (ii)  $T(c\mathbf{u}) = cT(\mathbf{u})$  for every scalar  $c$  and vector  $\mathbf{u}$  in the domain of  $T$ .

Every matrix transformation (e.g.  $\mathbf{x} \mapsto A\mathbf{x}$ ) is a linear transformation. And it turns out that every linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$  can be expressed in terms of matrix multiplication.

## A Theorem About Linear Transformations:

If  $T$  is a linear transformation, then

$$T(\mathbf{0}) = \mathbf{0},$$

$$T(c\mathbf{u} + d\mathbf{v}) = cT(\mathbf{u}) + dT(\mathbf{v})$$

for scalars  $c, d$  and vectors  $\mathbf{u}, \mathbf{v}$ .

And in fact

$$T(c_1\mathbf{u}_1 + c_2\mathbf{u}_2 + \cdots + c_k\mathbf{u}_k) = c_1T(\mathbf{u}_1) + c_2T(\mathbf{u}_2) + \cdots + c_kT(\mathbf{u}_k).$$

## Example

Let  $r$  be a nonzero scalar. The transformation  $T : \mathbb{R}^2 \rightarrow \mathbb{R}^2$  defined by

$$T(\mathbf{x}) = r\mathbf{x}$$

is a linear transformation<sup>1</sup>.

Show that  $T$  is a linear transformation.

We have to show that  $T(\vec{u} + \vec{v}) = T(\vec{u}) + T(\vec{v})$   
and  $T(c\vec{u}) = cT(\vec{u})$

Note  $T(\vec{u} + \vec{v}) = r(\vec{u} + \vec{v}) = r\vec{u} + r\vec{v} = T(\vec{u}) + T(\vec{v})$

$$T(c\vec{u}) = r(c\vec{u}) = rc\vec{u} = c r\vec{u} = c(r\vec{u}) = cT(\vec{u})$$

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<sup>1</sup>It's called a **contraction** if  $0 < r < 1$  and a **dilation** when  $r \geq 1$

These follow from the algebraic properties of scalar multiplication.

Hence  $T$  is a linear transformation.

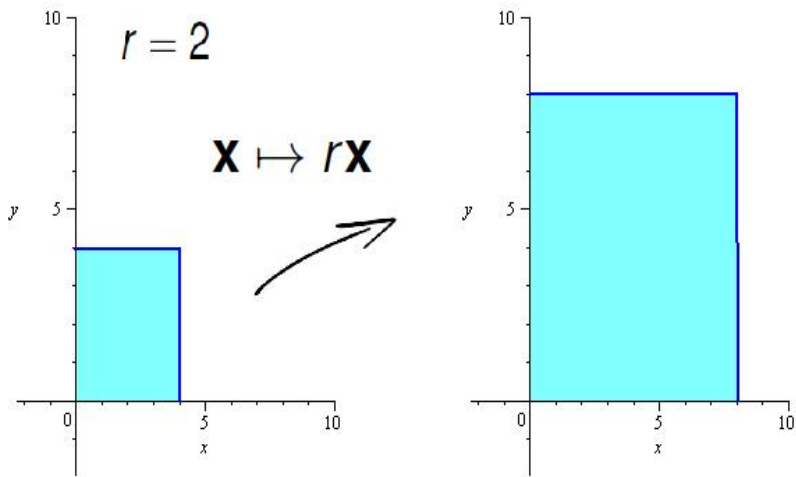


Figure: Geometry of dilation  $\mathbf{x} \mapsto 2\mathbf{x}$ . The 4 by 4 square maps to an 8 by 8 square.