

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 ($4 > 3$)
must be lin. dependent.

Determine if the set is linearly dependent or linearly independent

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

Contains the zero vector!

The set is lin. 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

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

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 .

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

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

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

We seek a solution \vec{x} to $T(\vec{x}) = \begin{bmatrix} -4 \\ -4 \\ 4 \end{bmatrix}$.

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

Using an augmented matrix

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

$$x_1 = 2$$

$$x_2 = -2$$

So a preimage $\vec{x} = \begin{bmatrix} 2 \\ -2 \end{bmatrix}$.

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

(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, i.e. is $A\vec{x} = \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$

consistent?

using 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

The system is inconsistent.

$\begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix}$ 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 transformation¹.

Show that T is a linear transformation.

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

Let \vec{u}, \vec{v} be in \mathbb{R}^2 and c in \mathbb{R} .

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

↗ $T(\vec{u})$ $T(\vec{v})$ ↖

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

$$\begin{aligned} T(c\vec{u}) &= r(c\vec{u}) = r c\vec{u} = c r\vec{u} \\ &= c (r\vec{u}) \\ &= c T(\vec{u}). \end{aligned}$$

So T satisfies both properties and is a linear transformation.

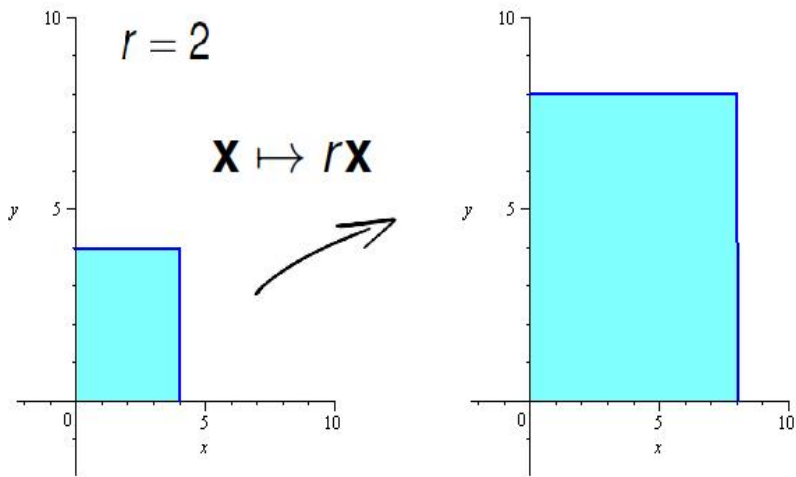


Figure: Geometry of dilation $\mathbf{x} \mapsto 2\mathbf{x}$. The 4 by 4 square maps to an 8 by 8 square.