

## 2 Points, Vectors, and Lines

### N-Space

**Notation.** Our main setting for this class is  $\mathbb{R}^n$ . We will always use boldface to denote an element of this set, say  $\mathbf{x} \in \mathbb{R}^n$ , and we will most commonly write  $\mathbf{x} = (x_1, \dots, x_n)$  to indicate the  $n$  real number coordinates that comprise  $\mathbf{x}$ . However, we will sometimes instead

use the notation  $\mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix}$  when this is more convenient (usually when we are doing some matrix operations).

**Points or Vectors?** An element  $\mathbf{x} \in \mathbb{R}^n$  may be viewed **both** as a point in space **and** as a “direction”. We will generally call  $\mathbf{x}$  a *point* to emphasize the first context or a *vector* to emphasize the second. However, it is vital to understand that every  $\mathbf{x} \in \mathbb{R}^n$  may be thought of both ways.

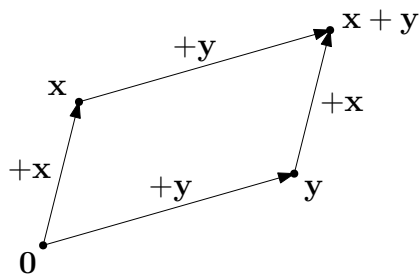
**Zero** We define  $\mathbf{0} = (0, \dots, 0)$ . One may view  $\mathbf{0}$  both as the origin and as a vector indicating the trivial direction.

**Sums** If  $\mathbf{x} = (x_1, \dots, x_n), \mathbf{y} = (y_1, \dots, y_n) \in \mathbb{R}^n$ , the sum of  $\mathbf{x}$  and  $\mathbf{y}$  is

$$\mathbf{x} + \mathbf{y} = (x_1, \dots, x_n) + (y_1, \dots, y_n) = (x_1 + y_1, \dots, x_n + y_n).$$

Addition is commutative:  $\mathbf{x} + \mathbf{y} = \mathbf{y} + \mathbf{x}$  and associative:  $(\mathbf{x} + \mathbf{y}) + \mathbf{z} = \mathbf{x} + (\mathbf{y} + \mathbf{z})$ .

**Note:** In figures we give points labels of the form  $\mathbf{x}$  (as usual). We will label an arrow  $+\mathbf{y}$  to indicate that adding  $\mathbf{y}$  takes you from the point at the start of the arrow to the point at the end. For instance, the following figure demonstrates the commutativity of the sum.



**Scalar Multiplication** If  $\mathbf{x} = (x_1, \dots, x_n) \in \mathbb{R}^n$  and  $t \in \mathbb{R}$ , the scalar product of  $t$  and  $\mathbf{x}$  is defined to be

$$t\mathbf{x} = t(x_1, \dots, x_n) = (tx_1, \dots, tx_n).$$

**Note:** Assuming  $\mathbf{x} \in \mathbb{R}^n$  is nonzero (i.e.  $\mathbf{x} \neq \mathbf{0}$ ) the line through  $\mathbf{0}$  and  $\mathbf{x}$  is precisely the set of all scalar multiples of  $\mathbf{x}$ .

**Example:**

## Lines

If  $L$  is a line through  $\mathbf{0}$  and  $\mathbf{x}$  is any point on  $L$  except  $\mathbf{0}$ , then we have  $L = \{t\mathbf{x} \mid t \in \mathbb{R}\}$ . How can we describe a line that does not go through  $\mathbf{0}$ ?

**Observation 2.1.** *If  $L$  is a line in  $\mathbb{R}^n$ , then for any point  $\mathbf{w} \in L$  and any vector  $\mathbf{x}$  parallel to (or in the same direction as)  $L$  we have*

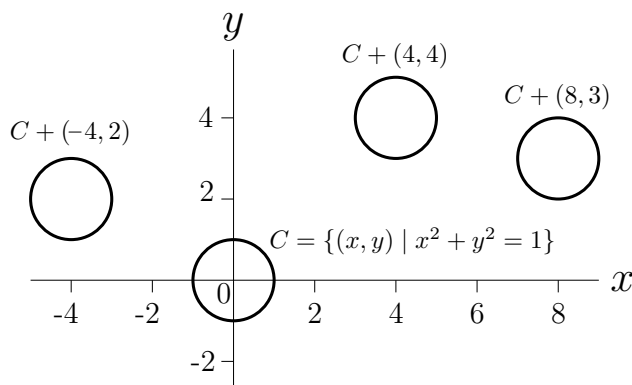
$$L = \{\mathbf{w} + t\mathbf{x} \mid t \in \mathbb{R}\}.$$

Next we will introduce a little more notation to think about this in another way.

**Definition.** If  $S \subseteq \mathbb{R}^n$  and  $\mathbf{x} \in \mathbb{R}^n$ , the translate of  $S$  by  $\mathbf{x}$  is defined to be

$$\mathbf{x} + S = \{\mathbf{x} + \mathbf{s} \mid \mathbf{s} \in S\}$$

**Example.**

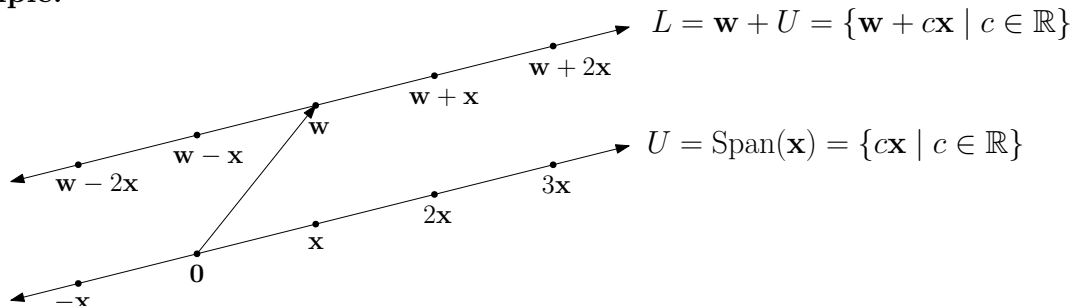


**Note:** With this framework, we have another way of describing a line, namely

$$L = \{\mathbf{w} + t\mathbf{x} \mid t \in \mathbb{R}\} = \mathbf{w} + \{t\mathbf{x} \mid t \in \mathbb{R}\}.$$

So the line in the direction of  $\mathbf{x}$  going through the point  $\mathbf{w}$  can also be thought of as a translate of the line in the direction of  $\mathbf{x}$  going through  $\mathbf{0}$ .

**Example.**



## Linear Combinations

**Definition:** If  $\mathbf{x}_1, \dots, \mathbf{x}_k \in \mathbb{R}^n$  and  $c_1, \dots, c_k \in \mathbb{R}$  then we call

$$c_1\mathbf{x}_1 + \dots + c_k\mathbf{x}_k$$

a *linear combination* of  $\mathbf{x}_1, \dots, \mathbf{x}_k$ . The set of all linear combinations of  $\mathbf{x}_1, \dots, \mathbf{x}_k$  is called the *span* of  $\mathbf{x}_1, \dots, \mathbf{x}_k$  and we express by

$$\text{Span}(\mathbf{x}_1, \dots, \mathbf{x}_k) = \{c_1\mathbf{x}_1 + \dots + c_k\mathbf{x}_k \mid c_1, \dots, c_k \in \mathbb{R}\}.$$

**Example:** For a single nonzero vector  $\mathbf{x}$  we have

$$\text{Span}(\mathbf{x}) = \{t\mathbf{x} \mid t \in \mathbb{R}\}$$

So the span of  $\mathbf{x}$  is a line through  $\mathbf{0}$ .

**Example:** Suppose (for the sake of concreteness) that we are in  $\mathbb{R}^3$ . If  $\mathbf{x}_1$  and  $\mathbf{x}_2$  are

**Definition.** A set  $U \subseteq \mathbb{R}^n$  is called a *subspace* if it satisfies the following three properties:

**zero**  $\mathbf{0} \in U$ .

**additive closure** If  $\mathbf{x}, \mathbf{y} \in U$  then  $\mathbf{x} + \mathbf{y} \in U$ .

**scalar mult. closure** If  $\mathbf{x} \in U$  and  $t \in \mathbb{R}$  then  $t\mathbf{x} \in U$ .

**Theorem 2.2.** A set  $U \subseteq \mathbb{R}^n$  is a subspace if and only if  $U = \text{span}(\mathbf{x}_1, \dots, \mathbf{x}_k)$  holds for some list of vectors.

**Definition:** If  $\mathbf{x}_1, \dots, \mathbf{x}_k \in \mathbb{R}^n$ , then we define

$$\text{Int}(\mathbf{x}_1, \dots, \mathbf{x}_k) = \{t_1\mathbf{x}_1 + t_2\mathbf{x}_2 + \dots + t_k\mathbf{x}_k \mid \text{where } t_1, \dots, t_k \in \mathbb{Z} \}$$

If  $\Lambda = \text{Int}(\mathbf{x}_1, \dots, \mathbf{x}_k)$  and  $\Lambda$  does not contain arbitrarily small nonzero vectors, we call  $\Lambda$  a lattice.

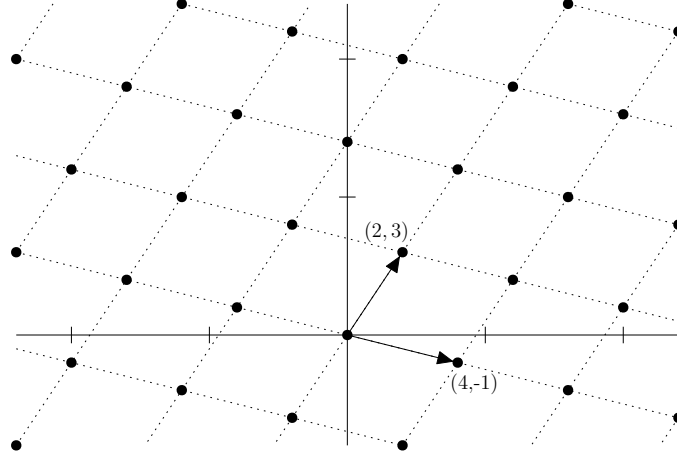


Figure 1: \*  
The lattice  $\Lambda((2, 3), (4, -1))$