# Chapter 1: Intro to Vectors

#### Sarah Helmbrecht

#### 1 Vectors

**Definition 1.1.** A (column) vector is a list of real numbers arranged in a column. We write vectors with arrows over them, as in

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix},$$

where  $x_1, x_2, \ldots, x_n \in \mathbb{R}$  are real numbers called the **components** of  $\vec{x}$ . An **n-dimensional vector** is a vector with n components.

**Definition 1.2.** The set of all n-dimensional vectors is n-dimensional Euclidean space, denoted  $\mathbb{R}^n$ .

**Definition 1.3.** (Vector addition) For any  $x_1, \ldots, x_n, y_1, \ldots, y_n \in \mathbb{R}$ , define

$$\begin{bmatrix} x_1\\x_2\\\vdots\\x_n \end{bmatrix} + \begin{bmatrix} y_1\\y_2\\\vdots\\y_n \end{bmatrix} = \begin{bmatrix} x_1+y_1\\x_2+y_2\\\vdots\\x_n+y_n \end{bmatrix}.$$

**Definition 1.4.** (Scalar multiplication) For any  $x_1, \ldots, x_n, c \in \mathbb{R}$ , define

$$c \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} cx_1 \\ cx_2 \\ \vdots \\ cx_n \end{bmatrix}.$$

**Definition 1.5.** Let  $\vec{0}$  denote the zero vector, whose components are all zero:

$$\vec{0} = \begin{bmatrix} 0\\0\\\vdots\\0 \end{bmatrix},$$

where the number of components is in context. For example, if  $\vec{0} \in \mathbb{R}^2$ , then  $\vec{0} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$ .

**Definition 1.6.** Let  $(\vec{v_1}, \vec{v_2}, \dots, \vec{v_n})$  be a list of vectors in  $\mathbb{R}^m$ . A linear combination of  $(\vec{v_1}, \vec{v_2}, \dots, \vec{v_n})$  is a vector of the form

 $c_1\vec{v_1} + c_2\vec{v_2} + \dots + c_n\vec{v_n}$ 

for some scalars  $c_1, c_2, \ldots, c_n \in \mathbb{R}$ , which are called the weights or coefficients.

### 2 Lengths and Dot Products

**Definition 2.1.** Let 
$$\vec{x}, \vec{y} \in \mathbb{R}^n$$
, where  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$ , and  $\vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix}$ . The **dot product**  $\vec{x} \cdot \vec{y}$  of  $\vec{x}$  and  $\vec{y}$  is given by

$$\vec{x} \cdot \vec{y} = x_1 y_1 + x_2 y_2 + \dots + x_n y_n.$$

**Proposition 2.2.** For all  $\vec{x}, \vec{y}, \vec{z} \in \mathbb{R}^n$  and scalars c,

- 1.  $\vec{x} \cdot \vec{y} = \vec{y} \cdot \vec{x}$  (commutative).
- 2.  $(\vec{x} + \vec{y}) \cdot \vec{z} = \vec{x} \cdot \vec{z} + \vec{y} \cdot \vec{z}$  (distributive over addition).
- 3.  $(c\vec{x}) \cdot \vec{y} = c(\vec{x} \cdot \vec{y}).$
- 4.  $\vec{x} \cdot \vec{x} \ge 0$ .
- 5.  $\vec{x} \cdot \vec{x} = 0$  if and only if  $\vec{x} = \vec{0}$ .

**Definition 2.3.** The length or norm of a vector  $\vec{x} \in \mathbb{R}^n$  is given by

$$\|\vec{x}\| = \sqrt{\vec{x} \cdot \vec{x}} = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$$

or equivalently,  $\|\vec{x}\|^2 = \vec{x} \cdot \vec{x}$ .

**Proposition 2.4.** For all  $\vec{x}, \vec{y} \in \mathbb{R}^n$  and scalars c,

- 1.  $\|\vec{x}\| \ge 0$ .
- 2.  $\|\vec{x}\| = 0$  if and only if  $\vec{x} = \vec{0}$ .
- 3.  $||c\vec{x}|| = |c| \cdot ||\vec{x}||.$
- 4.  $\|\vec{x} + \vec{y}\|^2 + \|\vec{x} \vec{y}\|^2 = 2(\|\vec{x}\|^2 + \|\vec{y}\|^2)$  (parallelogram law).
- 5.  $\vec{x} \cdot \vec{y} = \frac{1}{4} (\|\vec{x} + \vec{y}\|^2 \|\vec{x} \vec{y}\|^2).$

**Definition 2.5.** The distance between two points  $x, y \in \mathbb{R}^n$  is

$$dist(x,y) = \|y - x\|$$

This is just the length of the vector from x to y.

**Definition 2.6.** A vector  $\vec{x} \in \mathbb{R}^n$  is a unit vector if  $\|\vec{x}\| = 1$ .

**Definition 2.7.** Let  $\vec{x} \in \mathbb{R}^n$  be a nonzero vector. The unit vector in the direction of  $\mathbf{x}$  is the vector  $\frac{\vec{x}}{\|\vec{x}\|}$ .

**Definition 2.8.** Two vectors  $\vec{x}, \vec{y} \in \mathbb{R}^n$  are orthogonal or perpendicular if  $\vec{x} \cdot \vec{y} = 0$ .

**Proposition 2.9.** (Cosine Formula) The angle  $\theta$  between two nonzero vectors  $\vec{x}, \vec{y} \in \mathbb{R}^n$  is given by:

$$\frac{\vec{x} \cdot \vec{y}}{\|\vec{x}\| \|\vec{y}\|} = \cos \theta.$$

**Theorem 2.10.** (Cauchy-Schwarz Inequality) For all  $\vec{x}, \vec{y} \in \mathbb{R}^n$ ,

 $|\vec{x} \cdot \vec{y}| \le \|\vec{x}\| \|\vec{y}\|.$ 

**Proposition 2.11.** (*Triangle Inequality*) For all  $\vec{x}, \vec{y} \in \mathbb{R}^n$ ,

$$\|\vec{x} + \vec{y}\| \le \|\vec{x}\| + \|\vec{y}\|.$$

#### 3 Spans, Linear Dependence, and Linear Independence

**Definition 3.1.** Let  $\vec{v_1}, \vec{v_2}, \ldots, \vec{v_m}$  be a set of vectors in  $\mathbb{R}^n$ . The **span** of  $\vec{v_1}, \vec{v_2}, \ldots, \vec{v_m}$  is the set of all linear combinations of  $\vec{v_1}, \vec{v_2}, \ldots, \vec{v_m}$ , and is denoted span $\{\vec{v_1}, \vec{v_2}, \ldots, \vec{v_m}\}$ .

In symbols:  $span\{\vec{v_1}, \vec{v_2}, \dots, \vec{v_m}\} = \{c_1\vec{v_1} + c_2\vec{v_2} + \dots + c_m\vec{v_m} : c_1, c_2, \dots, c_m \in \mathbb{R}\}.$ 

**Definition 3.2.** Suppose  $\vec{v_1}, \vec{v_2}, \ldots, \vec{v_m} \in \mathbb{R}^n$  and consider the equation

$$c_1 \vec{v_1} + c_2 \vec{v_2} + \dots + c_m \vec{v_m} = \vec{0}$$

This equation always has the solution  $c_1 = c_2 = \cdots = c_m = 0$ , called the trivial solution.

- 1. If  $c_1\vec{v_1} + c_2\vec{v_2} + \cdots + c_m\vec{v_m} = \vec{0}$  has only the trivial solution, then  $\{\vec{v_1}, \vec{v_2}, \dots, \vec{v_m}\}$  is linearly independent.
- 2. If  $c_1\vec{v_1} + c_2\vec{v_2} + \cdots + c_m\vec{v_m} = \vec{0}$  has nontrivial solutions, then  $\{\vec{v_1}, \vec{v_2}, \dots, \vec{v_m}\}$  is linearly dependent.

**Proposition 3.3.** A set  $\{\vec{v_1}, \vec{v_2}, \dots, \vec{v_m}\}$  of vectors in  $\mathbb{R}^n$  is **linearly dependent** if and only if one of the vectors is in the span of the others.

## 4 Matrices

**Definition 4.1.** A matrix is a rectangular array of numbers. We say a matrix is  $m \times n$ , i.e. "m by n" if it has m rows and n columns. If A is  $m \times n$  with entries  $a_{ij}$ , then we write

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix}.$$

A matrix is called **square** if it is  $n \times n$ .

**Definition 4.2.** The entries  $a_{11}, a_{22}, a_{33}, \ldots$  are the **diagonal entries**, which form the **main diagonal** of the matrix. A **diagonal matrix** is a square matrix whose non-diagonal entries are all zeros.

**Definition 4.3.** The  $n \times n$  identity matrix  $I_n$  is the diagonal matrix with all diagonal entries equal to 1.  $I_n$  is special because  $I_n \vec{v} = \vec{v}$  for all  $\vec{v} \in \mathbb{R}^n$ .

**Definition 4.4.** The  $m \times n$  zero matrix is the  $m \times n$  matrix 0 with all zero entries.

**Definition 4.5.** The transpose of an  $m \times n$  matrix A is the  $n \times m$  matrix  $A^T$  whose rows are the columns of A. In other words, the *ij*-th entry of  $A^T$  is  $a_{ji}$ .

**Proposition 4.6.** A linear combination of n vectors  $\vec{v_1}, \vec{v_2}, \ldots, \vec{v_n} \in \mathbb{R}^m$  can be expressed as an  $m \times n$  matrix  $A = \begin{bmatrix} \vec{v_1} & \vec{v_2} & \cdots & \vec{v_n} \end{bmatrix}$  multiplying a vector  $\vec{x} \in \mathbb{R}^n$ :

$$\begin{bmatrix} \vec{v_1} & \vec{v_2} & \cdots & \vec{v_n} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} = x_1 \vec{v_1} + x_2 \vec{v_2} + \cdots + x_n \vec{v_n}$$