

## Vectors in Euclidean Spaces

### Vectors.

- Economists usually work in the vector space  $\mathbb{R}^n$ . A point in this space is called a *vector*, and is typically defined by its rectangular coordinates.
- For instance, let  $v \in \mathbb{R}^n$ . We define this vector by its  $n$  coordinates,  $v_1, v_2, \dots, v_n$ . It is common to write  $v = (v_1, v_2, \dots, v_n)$  or to display a vector as a column matrix:

$$v = \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix}$$

- It is common to distinguish between locations and displacements by writing a location as a row vector and a displacement as a column vector. However, we can use the same algebraic operations to work with each.
- A vector can be also be defined by its origin and end points.
- Suppose the vector  $v$  links the point  $P = (p_1, \dots, p_n)$  to the point  $Q = (q_1, \dots, q_n)$  in  $\mathbb{R}^n$ . Then  $v = Q - P$ , i.e.  $v_i = p_i - q_i, \forall i \in \{1, 2, \dots, n\}$ .

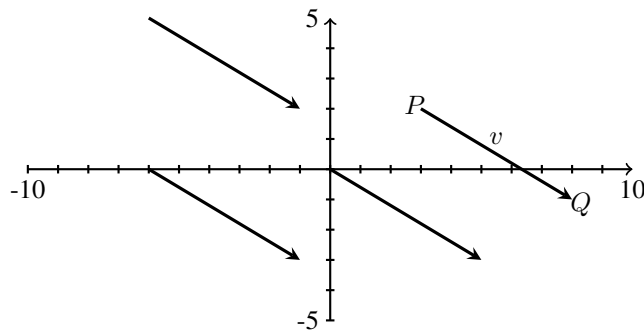


Figure 1: The displacement  $(5, -3)$

### Addition.

- Given two vectors  $u$  and  $v$ , with coordinates, we add them like so:

$$u + v = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} + \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} = \begin{pmatrix} u_1 + v_1 \\ u_2 + v_2 \\ \vdots \\ u_n + v_n \end{pmatrix}.$$

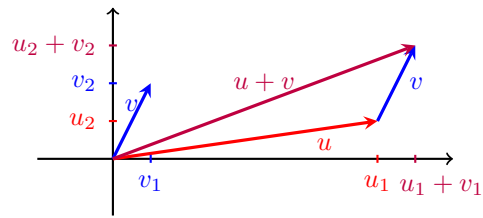


Figure 2: Vector Addition

### Scalar Multiplication.

- We can also multiply vectors by scalars. Suppose  $\lambda \in \mathbb{R}$  and  $v \in \mathbb{R}^n$ . Scalar multiplication gives a vector  $\lambda v \in \mathbb{R}^n$ , defined by

$$\lambda v = \begin{pmatrix} \lambda v_1 \\ \lambda v_2 \\ \vdots \\ \lambda v_n \end{pmatrix}$$

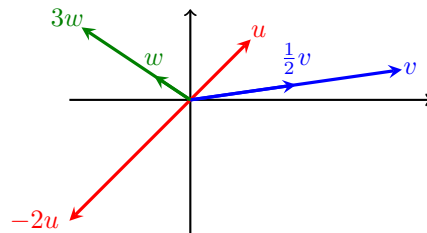


Figure 3: Scalar multiplication

### Subtraction.

- The difference of two vectors, say  $u - v$ , is the sum of the vector  $u$  with the vector  $-v = (-1)v$ .

$$u - v = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_n \end{pmatrix} + (-1) \begin{pmatrix} v_1 \\ v_2 \\ \vdots \\ v_n \end{pmatrix} = \begin{pmatrix} u_1 - v_1 \\ u_2 - v_2 \\ \vdots \\ u_n - v_n \end{pmatrix}.$$

### Laws of Vector Algebra.

- Let  $\lambda, \beta \in \mathbb{R}$  and  $u, v, w \in \mathbb{R}^n$ . Then the following algebraic properties of vectors hold.

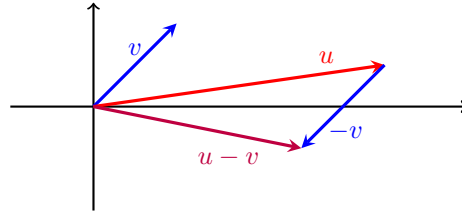


Figure 4: Vector Subtraction

– Associativity:

$$\begin{aligned}(u + v) + w &= u + (v + w) \\ (\lambda\beta)v &= \lambda(\beta v)\end{aligned}$$

– Commutativity:

$$u + v = v + u$$

– Distributivity:

$$\begin{aligned}\lambda(u + v) &= \lambda u + \lambda v \\ (\lambda + \beta)u &= \lambda u + \beta u\end{aligned}$$

**Definition.** Two vectors  $u, v \in \mathbb{R}^n$  are *parallel* if there exists a real number  $\lambda$  such that  $u = \lambda v$ . ▲

### The Inner Product.

**Definition.** Let  $u$  and  $v$  be two vectors in  $\mathbb{R}^n$ . The (*Euclidean*) *inner product* (also called *dot product* or *scalar product*) of  $u$  and  $v$  is the real number, denoted  $u \cdot v$ , given by

$$\begin{aligned}u \cdot v &= u_1 v_1 + u_2 v_2 + \cdots + u_n v_n \\ &= \sum_{i=1}^n u_i v_i\end{aligned}$$

- Note that if we think of  $u$  and  $v$  as  $n \times 1$  matrices, we have  $u \cdot v = u^T v = v^T u$ .
- If we think of them as  $1 \times n$  matrices, then  $u \cdot v = uv^T = vu^T$ .

### Laws of the Inner Product.

Let  $\lambda \in \mathbb{R}$  and  $u, v, w \in \mathbb{R}^n$ . Then:

- Associativity:

$$(\lambda u) \cdot v = \lambda(u \cdot v)$$

- Commutativity:

$$u \cdot v = v \cdot u$$

- Distributivity:

$$u \cdot (v + w) = u \cdot v + u \cdot w$$

### Length and Inner Product.

**Definition.** The *norm* or *length* of a vector  $u$  is the real number, denoted  $\|u\|$ , given by

$$\|u\| = \sqrt{u_1^2 + u_2^2 + \cdots + u_n^2}. \quad \blacktriangle$$

- Using our definition of the inner product we can also write this as

$$\|u\| = \sqrt{u \cdot u}$$

- The norm of a vector is always positive unless the vector is the zero vector, in which case the norm is zero.
- The distance between two vectors  $u, v \in \mathbb{R}^n$  is calculated as

$$\|u - v\| = \sqrt{(u_1 - v_1)^2 + \cdots + (u_n - v_n)^2}.$$

- Note that for any  $\lambda \in \mathbb{R}$  and  $u \in \mathbb{R}^n$ :

$$\|\lambda u\| = |\lambda| \|u\|.$$

- We can show that

$$u \cdot v = \|u\| \|v\| \cos \theta$$

where  $\theta$  is the angle between the vectors  $u$  and  $v$ .

- Using the properties of the cosine we get the following result.

**Theorem 1.** *The angle between vectors  $u$  and  $v$  in  $\mathbb{R}^n$  is*

1. *acute, if  $u \cdot v > 0$ ,*
2. *obtuse, if  $u \cdot v < 0$ ,*
3. *right, if  $u \cdot v = 0$ .*

**Definition.** Let  $v$  be a vector. The vector  $w$  which points in the same direction as  $v$ , but has length 1 is called the *unit vector* in the direction of  $v$  (or simply the *direction* of  $v$ ). It is given by

$$w = \frac{v}{\|v\|}. \quad \blacktriangle$$

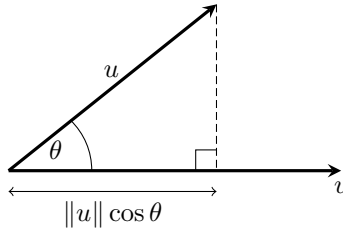


Figure 5: The angle between two vectors  $u$  and  $v$ .

**Definition.** Two vectors  $u, v \in \mathbb{R}^n$  are *orthogonal* if  $u \cdot v = 0$ . ▲

- This definition implies the zero vector is orthogonal to any vector.

**Definition.** Two vectors  $u, v \in \mathbb{R}^n$  are *orthonormal* if they are orthogonal and are unit vectors. ▲

**Theorem 2** (Triangle Inequality). For any two vectors  $u, v \in \mathbb{R}^n$ ,

$$\|u + v\| \leq \|u\| + \|v\|.$$

**Theorem 3** (Triangle Inequality Variant). For any two vectors  $u, v \in \mathbb{R}^n$ ,

$$\left| \|u\| - \|v\| \right| \leq \|u - v\|.$$

- There are three basic properties of Euclidean length for any vectors  $u$  and  $v$  and scalar  $\lambda$ :

1.  $\|u\| \geq 0$  and  $\|u\| = 0$  only when  $u = 0$ ,
2.  $\|\lambda u\| = |\lambda| \|u\|$ ,
3.  $\|u + v\| \leq \|u\| + \|v\|$ .

Any assignment of a real number to a vector satisfying these properties is called a norm (see sections 29.4 and 27 of S&B if interested).

### Projections.

- Let  $u, v \in \mathbb{R}^n$ . We want to find the *vector projection* of the vector  $u$  in the direction of  $v$ .
- Denote the projection of  $u$  on  $v$  by  $P_v(u)$ . We can see from the diagram that the length of  $P_v(u)$  (called the *scalar projection* of vector  $u$  on  $v$ ) is given by

$$\|P_v(u)\| = \|u\| \cos \theta.$$

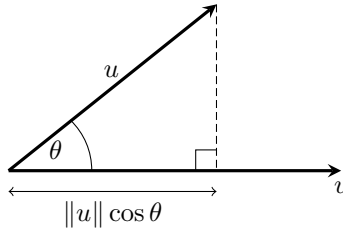


Figure 6: Projection of vector  $u$  in direction  $v$

- The unit vector in the direction of  $v$  is

$$\frac{1}{\|v\|}v.$$

So we have

$$P_v(u) = \frac{\|u\| \cos \theta}{\|v\|}v.$$

Using our expression for the inner product  $u \cdot v = \|u\|\|v\| \cos \theta$  we get

$$P_v(u) = \frac{u \cdot v}{\|v\|^2}v.$$

- Note that if the vectors are parallel, say  $u = \lambda v$  for some scalar  $\lambda$ , then  $P_v(u) = u$ . This says that projecting a vector  $u$  in its own direction gives the same vector  $u$ .

### Lines.

- A straight line is completely determined by two things:
  - a point  $p \in \mathbb{R}^n$  on the line, and
  - a direction  $v \in \mathbb{R}^n$ , with  $v \neq 0$ , in which to move from  $p$ .
- A straight line  $\ell$  in  $\mathbb{R}^n$  can be defined in its parametric form as

$$\ell = \{x \in \mathbb{R}^n \mid x = p + tv, t \in \mathbb{R}\}$$

- So, a point  $x \in \mathbb{R}^n$  belongs to the line  $\ell$  iff there exists a real number  $t$  such that  $x = p + tv$ .
- A straight line  $\ell$  can also be defined by two distinct points  $P$  and  $Q$ .
  - The vector  $v = Q - P$  gives the direction of the line.
  - So then the line is just the set of points  $P + tv$ , with  $t \in \mathbb{R}$ .
- Now,  $P + tv = P + t(Q - P) = (1 - t)P + tQ$ , so the line is

$$\ell = \{(1 - t)P + tQ \in \mathbb{R}^n \mid t \in \mathbb{R}\}$$

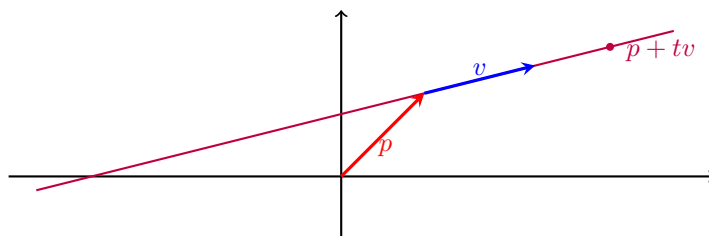


Figure 7: Parametric line  $\ell$  in  $\mathbb{R}^2$

### Two-dimensional Planes.

- We saw that a line – a one-dimensional object – can be described using only one parameter.
- A plane is two-dimensional, so we need two parameters.
- Consider a plane  $\mathcal{P}$  in  $\mathbb{R}^3$  and let  $u$  and  $v$  be two vectors in  $\mathcal{P}$  that point in different directions:

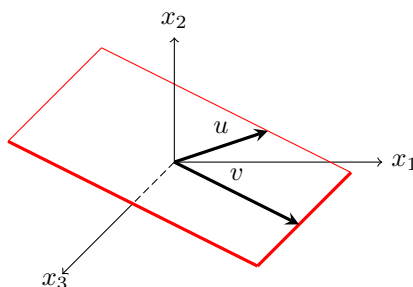


Figure 8: A plane  $\mathcal{P}$  through the origin in  $\mathbb{R}^3$

- We can move from the origin in direction  $u$ ,  $v$  or any combination of the two. So, for any scalars  $s$  and  $t$ , the vector  $su + tv$  also lies in the plane  $\mathcal{P}$ .
- Thus any plane  $\mathcal{P}$  through the origin, in a vector space  $\mathbb{R}^n$  ( $n > 2$ ), can be defined in its parametric form as

$$\mathcal{P} = \{x \in \mathbb{R}^n \mid x = su + tv, s, t \in \mathbb{R}\}$$

- But what if the plane does not pass through the origin?
- Suppose the plane does not pass through the origin.
- We can move from point  $p$  in the plane in direction  $u$ ,  $v$  or any combination of the two. Thus the vector  $p + su + tv$  also lies in the plane  $\mathcal{P}$ .

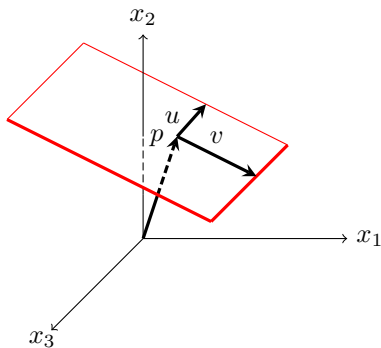


Figure 9: A plane  $\mathcal{P}$  not through the origin in  $\mathbb{R}^3$

- So any plane  $\mathcal{P}$  through the point  $p$ , in a vector space  $\mathbb{R}^n$  ( $n > 2$ ), can be defined in its parametric form as

$$\mathcal{P} = \{x \in \mathbb{R}^n \mid x = p + su + tv, s, t \in \mathbb{R}\}$$

- A point  $x \in \mathbb{R}^n$  belongs to the plane  $\mathcal{P}$  iff there exist two scalars  $s$  and  $t$  such that  $x = p + su + tv$ . Equivalently, the vector  $x - p$  must be a linear combination of the vectors  $u$  and  $v$ .
- As two points uniquely determine a line, three distinct (non-collinear) points  $P$ ,  $Q$  and  $R$  uniquely determine a plane.
  - Let  $u = Q - P$  and  $v = R - P$ . We can picture these as displacement vectors from  $P$ :

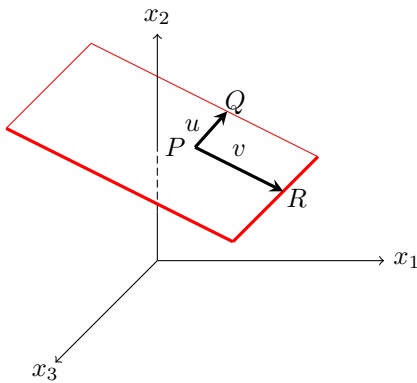


Figure 10: A plane  $\mathcal{P}$  not through the origin in  $\mathbb{R}^3$

- Remember, we need  $u$  and  $v$  to be nonparallel to for them to uniquely determine a plane.



- If  $u$  and  $v$  are parallel, then there exists a scalar  $\lambda$  such that  $v = \lambda u$ . But by definition of  $u$  and  $v$ , it would then be the case that  $R = (1 - \lambda)P + \lambda Q$ .
- But then the points lie on a line  $\ell$  and so are collinear.

- The plane containing the three points  $P, Q$  and  $R$  is

$$\mathcal{P} = \{(1 - s - t)P + sQ + tR \mid s, t \in \mathbb{R}\},$$

or equivalently

$$\mathcal{P} = \{t_P P + t_Q Q + t_R R \mid t_P, t_Q, t_R \in \mathbb{R} \text{ and } t_P + t_Q + t_R = 1\},$$

- We can also completely describe a plane using
  - a point  $p = (x_0, y_0, z_0) \in \mathbb{R}^3$  in the plane, and
  - an inclination, specified by a vector  $n = (a, b, c) \in \mathbb{R}^3$ , called a *normal vector*, which is perpendicular to the plane.

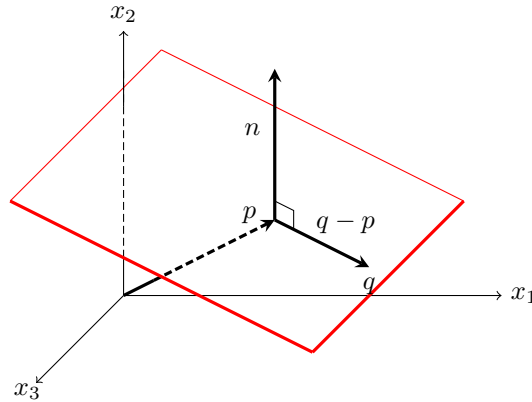


Figure 11: A plane  $\mathcal{P}$  through  $p$  with normal  $n$

- Let  $q = (x, y, z)$  be an arbitrary point on the plane  $\mathcal{P}$ . Then  $q - p$  is a vector in the plane and will thus be perpendicular to  $n$ .
- Two vectors are orthogonal iff their dot product is zero, so the plane is defined as

$$\mathcal{P} = \{q \in \mathbb{R}^3 \mid n \cdot (q - p) = 0\}.$$

- Now, for any point  $q$  in the plane

$$0 = n \cdot (q - p) = (a, b, c) \cdot (x - x_0, y - y_0, z - z_0),$$

or

$$a(x - x_0) + b(y - y_0) + c(z - z_0) = 0.$$

- So another way to write the nonparametric equation of a plane, is

$$ax + by + cz = d,$$

where  $d = ax_0 + by_0 + cz_0$ . This is called the *point-normal equation* of the plane.

## Hyperplanes.

- We saw a line in  $\mathbb{R}^2$  can be written as

$$a_1x_1 + a_2x_2 = d$$

and a plane in  $\mathbb{R}^3$  can be written in point-normal form as

$$a_1x_1 + a_2x_2 + a_3x_3 = d.$$

- Generalizing, a *hyperplane* in  $\mathbb{R}^n$  can be written in point-normal form as

$$a_1x_1 + a_2x_2 + \cdots + a_nx_n = d,$$

where  $(a_1, a_2, \dots, a_n)$  is a normal.

- The set of vectors in the hyperplane have tail at  $(0, \dots, 0, d/a_n)$  and are perpendicular to the normal vector to the hyperplane.

### Example 1.

1. An economic application you have probably seen deals with commodity spaces.

- The vector

$$x = (x_1, x_2, \dots, x_n)$$

of nonnegative quantities of  $n$  commodities is called a *commodity bundle*. The set of all commodity bundles is the set

$$\{(x_1, \dots, x_n) \mid x_1 \geq 0, \dots, x_n \geq 0\}$$

and is called a *commodity space*.

- Let  $p_i > 0$  be the price of commodity  $i$ . The cost of buying bundle  $x$  is

$$p_1x_1 + p_2x_2 + \cdots + p_nx_n = p \cdot x.$$

A consumer with income  $I$  can purchase only bundles  $x$  for which  $p \cdot x \leq I$ . This subset of the commodity space is the consumer's *budget set*.

1.
  - The budget set is bounded above by the hyperplane  $p \cdot x = I$ , whose normal vector is the price vector  $p$ .

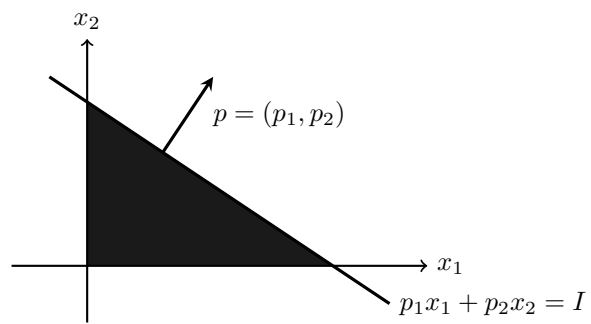


Figure 12: A consumer's budget set,  $p \cdot x \leq I$ , in commodity space.

2. • Another hyperplane you will see is the space of *probability vectors*

$$P_n = \{(p_1, \dots, p_n) \mid p_1 \geq 0, \dots, p_n \geq 0 \text{ and } p_1 + \dots + p_n = 1\},$$

which is called the *probability simplex*. The probability simplex  $P_n$  is part of a hyperplane in  $\mathbb{R}^n$  with normal vector  $(1, 1, \dots, 1)$ . ♦

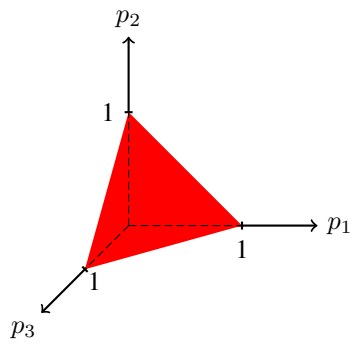


Figure 13: The probability simplex for  $n = 3$ .