# **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 ∈ ℝ<sup>n</sup>. We define this vector by its n coordinates, v<sub>1</sub>, v<sub>2</sub>,..., v<sub>n</sub>. It is common to write v = (v<sub>1</sub>, v<sub>2</sub>,..., v<sub>n</sub>) 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 dispacements 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  $P = (p_1, \ldots, p_n)$  to the point  $Q = (q_1, \ldots, q_n)$ in  $\mathbb{R}^n$ . Then v = Q - P, i.e.  $v_i = p_i - q_i, \forall i \in \{1, 2, \ldots, n\}$ .



# 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



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:

$$(u+v)+w = u+(v+w)$$
$$(\lambda\beta)v = \lambda(\beta v)$$

- Commutativity:

u+v = v+u

- Distributivity:

**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

$$u \cdot v = u_1 v_1 + u_2 v_2 + \dots + u_n v_n$$
$$= \sum_{i=1}^n u_i v_i$$

▲

- 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

$$||u|| = \sqrt{u_1^2 + u_2^2 + \dots + u_n^2}.$$

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

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

- 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

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

• 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

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



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

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

$$||u+v|| \le ||u|| + ||v||.$$

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

$$|||u|| - ||v||| \le ||u - v||.$$

- There are three basic properties of Euclidean length for any vectors u and v and scalar λ:
  - 1.  $||u|| \ge 0$  and ||u|| = 0 only when u = 0,
  - 2.  $\|\lambda u\| = |\lambda| \|u\|$ ,
  - 3.  $||u+v|| \le ||u|| + ||v||$ .

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}\}\$$



## **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 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 throught 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 P.



• 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

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

• Now, for any point q in the plane

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

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

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

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

- The set of vectors in the hyperplane have tail at  $(0, \ldots, 0, d/a_n)$  and are perpindicular 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

$$\{(x_1,\ldots,x_n) \mid x_1 \ge 0,\ldots,x_n \ge 0\}$$

and is called a commodity space.

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

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

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*.

The budget set is bounded above by the hyperplane p ⋅ x = I, whose normal vector is the price vector p.





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

$$P_n = \{(p_1, \dots, p_2) \mid p_1 \ge 0, \dots, p_n \ge 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, ..., 1).

