

Analysis in \mathbb{R}^n
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Geometric Linear Algebra

Definition 0.0.1 Let \mathbb{R} be the real numbers, that is, the unique ordered field that satisfies the least upper bound property.

Definition 0.0.2 Let $\mathbb{R}^n = \{(x_1, x_2, \dots, x_n) \mid x_i \in \mathbb{R}, i = 1, 2, \dots, n\}$ be the *points* in \mathbb{R}^n . Let $\mathbb{E}^n = \{v_1 \mathbf{e}_1 + v_2 \mathbf{e}_2 + \dots + v_n \mathbf{e}_n \mid v_i \in \mathbb{R}, i = 1, 2, \dots, n\}$, where $\mathbf{e}_1 = (1, 0, 0, \dots, 0)$, $\mathbf{e}_2 = (0, 1, 0, \dots, 0)$, \dots , $\mathbf{e}_n = (0, 0, 0, \dots, 0, 1)$, be the *vectors* in \mathbb{R}^n .

Definition 0.0.3 If $\mathbf{x} = (x_1, x_2, \dots, x_n), \mathbf{y} = (y_1, y_2, \dots, y_n) \in \mathbb{E}^n$, then we define *vector addition* as $\mathbf{x} + \mathbf{y} = (x_1 + y_1, x_2 + y_2, \dots, x_n + y_n)$.

Definition 0.0.4 If $\mathbf{x} = (x_1, x_2, \dots, x_n) \in \mathbb{E}^n$ and $c \in \mathbb{R}$, then we define *scalar multiplication* as $c \cdot \mathbf{x} = (c \cdot x_1, c \cdot x_2, \dots, c \cdot x_n)$.

Exercise 0.0.5 Show that \mathbb{E}^n is an n -dimensional vector space over \mathbb{R} .

Definition 0.0.6 Let \mathbf{p}_0 be a point in \mathbb{R}^n and \mathbf{v}_0 be a direction in \mathbb{E}^n . The *line* ℓ through \mathbf{p}_0 in the direction \mathbf{v}_0 is given by

$$\ell = \{\mathbf{p} \in \mathbb{R}^n \mid \mathbf{p} = \mathbf{p}_0 + t\mathbf{v}_0, t \in \mathbb{R}\}.$$

Definition 0.0.7 Suppose that $\mathbf{p}_0 \in \mathbb{R}^n$ and \mathbf{v}, \mathbf{w} are linearly independent vectors in \mathbb{E}^n . The *plane* through \mathbf{p}_0 , spanned by \mathbf{v} and \mathbf{w} , is

$$\mathcal{P} = \{\mathbf{p} \in \mathbb{R}^n \mid \mathbf{p} = \mathbf{p}_0 + t\mathbf{v} + s\mathbf{w}, t, s \in \mathbb{R}\}.$$

Definition 0.0.8 If $\mathbf{v}_1, \dots, \mathbf{v}_k$ are linearly independent vectors in \mathbb{E}^n , then we define the *k -dimensional affine subspace* through $\mathbf{p}_0 \in \mathbb{R}^n$, spanned by $\mathbf{v}_1, \dots, \mathbf{v}_k$, as

$$\mathbf{H} = \{\mathbf{p} \in \mathbb{R}^n \mid \mathbf{p} = \mathbf{p}_0 + t_1\mathbf{v}_1 + \dots + t_k\mathbf{v}_k, \text{ where } t_j \in \mathbb{R} \text{ and } 1 \leq j \leq k\}.$$

Definition 0.0.9 When $k = n - 1$ in the definition above, \mathbf{H} is called a *hyperplane* in \mathbb{R}^n .

Exercise 0.0.10 If $\mathbf{v}_1, \dots, \mathbf{v}_k$ are vectors in \mathbb{E}^n , then the collection of vectors $\{t_1\mathbf{v}_1 + \dots + t_k\mathbf{v}_k \mid t_j \in \mathbb{R}, j = 1, \dots, k\}$ is a subspace of \mathbb{E}^n .

Definition 0.0.11 If $\mathbf{v}_1, \dots, \mathbf{v}_k$ are linearly independent vectors in \mathbb{E}^n with $k \leq n$ and $\mathbf{p}_0 \in \mathbb{R}^n$, we define the *k -dimensional parallelepiped with vertex \mathbf{p}_0 spanned by $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$* as

$$\mathbf{P} = \{\mathbf{p} \in \mathbb{R}^n \mid \mathbf{p} = \mathbf{p}_0 + t_1\mathbf{v}_1 + \dots + t_k\mathbf{v}_k, \text{ with } 0 \leq t_j \leq 1\}.$$

Definition 0.0.12 Let V be a vector space over a field F . A *bilinear form* $\langle \cdot, \cdot \rangle$ on V is a map

$$\langle \cdot, \cdot \rangle : V \times V \rightarrow F$$

which satisfies linearity in both variables. That is, for all $\mathbf{v}, \mathbf{w} \in V$, and all $\alpha \in F$,

$$\begin{aligned}\langle \mathbf{v}_1 + \mathbf{v}_2, \mathbf{w} \rangle &= \langle \mathbf{v}_1, \mathbf{w} \rangle + \langle \mathbf{v}_2, \mathbf{w} \rangle \\ \langle \alpha \mathbf{v}, \mathbf{w} \rangle &= \alpha \langle \mathbf{v}, \mathbf{w} \rangle \\ \langle \mathbf{v}, \mathbf{w}_1 + \mathbf{w}_2 \rangle &= \langle \mathbf{v}, \mathbf{w}_1 \rangle + \langle \mathbf{v}, \mathbf{w}_2 \rangle \\ \langle \mathbf{v}, \alpha \mathbf{w} \rangle &= \alpha \langle \mathbf{v}, \mathbf{w} \rangle.\end{aligned}$$

The form $\langle \cdot, \cdot \rangle$ is said to be *symmetric* if $\langle \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{w}, \mathbf{v} \rangle$ for all $\mathbf{v}, \mathbf{w} \in V$. When $F = \mathbb{R}$ we say the form is *positive definite* if $\langle \mathbf{v}, \mathbf{v} \rangle \geq 0$ for all $\mathbf{v} \in V$, and $\langle \mathbf{v}, \mathbf{v} \rangle = 0$ if and only if $\mathbf{v} = \mathbf{0}$. Henceforth we assume $F = \mathbb{R}$.

Definition 0.0.13 Suppose that $\mathbf{v} = (v_1, \dots, v_n)$ and $\mathbf{w} = (w_1, \dots, w_n)$ are vectors in \mathbb{E}^n . The *scalar product* of \mathbf{v} and \mathbf{w} is $\langle \mathbf{v}, \mathbf{w} \rangle = v_1 w_1 + \dots + v_n w_n$. The scalar product is sometimes called the *dot product* and is denoted by $\mathbf{v} \cdot \mathbf{w}$.

Exercise 0.0.14 Prove that the scalar product is a positive definite symmetric bilinear form on \mathbb{E}^n .

Definition 0.0.15 If $\mathbf{v} = (v_1, \dots, v_n) \in \mathbb{E}^n$, then the *length* or *norm* of \mathbf{v} is defined by

$$\|\mathbf{v}\| = \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle} = (v_1^2 + \dots + v_n^2)^{1/2}.$$

Exercise 0.0.16 Prove the following properties of the norm. If $\mathbf{v}, \mathbf{w} \in \mathbb{E}^n$, then:

- i. $\|\mathbf{v}\| \geq 0$;
- ii. $\|\mathbf{v}\| = 0$ iff $\mathbf{v} = \mathbf{0}$;
- iii. $\|\alpha \mathbf{v}\| = |\alpha| \|\mathbf{v}\|$, $\alpha \in \mathbb{R}$;
- iv. $\|\mathbf{v} + \mathbf{w}\| \leq \|\mathbf{v}\| + \|\mathbf{w}\|$;
- v. $\|\mathbf{v} + \mathbf{w}\|^2 + \|\mathbf{v} - \mathbf{w}\|^2 = 2(\|\mathbf{v}\|^2 + \|\mathbf{w}\|^2)$.

Theorem 0.0.17 (Cauchy-Schwarz Inequality) Let $\mathbf{v}, \mathbf{w} \in \mathbb{E}^n$. Then $|\langle \mathbf{v}, \mathbf{w} \rangle| \leq \|\mathbf{v}\| \|\mathbf{w}\|$.

Exercise 0.0.18 Prove that equality holds in the Cauchy-Schwarz Inequality iff one of the vectors is a scalar multiple of the other.

Definition 0.0.19 The *distance* between two points $\mathbf{p}_1, \mathbf{p}_2 \in \mathbb{R}^n$ is given by $d(\mathbf{p}_1, \mathbf{p}_2) = \|\mathbf{p}_1 - \mathbf{p}_2\|$.

Definition 0.0.20 Let $\mathbf{v}, \mathbf{w} \in \mathbb{E}^n$. Then \mathbf{v} and \mathbf{w} are said to be *orthogonal* (or *perpendicular*) if $\langle \mathbf{v}, \mathbf{w} \rangle = 0$.

Exercise 0.0.21

- i. Show that the $\mathbf{0}$ vector in \mathbb{E}^n is orthogonal to every vector in \mathbb{E}^n .
- ii. Show that the vectors in the set $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$ are pairwise orthogonal, that is $\langle \mathbf{e}_i, \mathbf{e}_j \rangle = 0$ if $i \neq j$, and further that $\langle \mathbf{e}_i, \mathbf{e}_i \rangle = 1$.
- iii. If \mathbf{v} is a nonzero vector in \mathbb{E}^n , show that the collection $W = \{\mathbf{w} \mid \langle \mathbf{w}, \mathbf{v} \rangle = 0\}$ is an $(n - 1)$ -dimensional subspace of \mathbb{E}^n .
- iv. If $\mathbf{v}_1, \dots, \mathbf{v}_k$ are pairwise orthogonal non-zero vectors in \mathbb{E}^n , show that they form a linearly independent set in \mathbb{E}^n .

Theorem 0.0.22 Let \mathbf{v} and \mathbf{w} be linearly independent vectors in \mathbb{E}^n . The angle between \mathbf{v} and \mathbf{w} is the unique solution θ to the equation

$$\cos \theta = \frac{\langle \mathbf{v}, \mathbf{w} \rangle}{\|\mathbf{v}\| \|\mathbf{w}\|}, \quad 0 < \theta < 180^\circ. \quad (1)$$

Definition 0.0.23 Let \mathbf{v} and \mathbf{w} be linearly independent vectors in \mathbb{E}^n . The *projection of \mathbf{v} onto \mathbf{w}* is defined by

$$\text{proj}_{\mathbf{w}}(\mathbf{v}) = \frac{\langle \mathbf{v}, \mathbf{w} \rangle}{\|\mathbf{w}\|^2} \mathbf{w}. \quad (2)$$

Exercise 0.0.24 Prove that if $\mathbf{w} \neq \mathbf{0}$, then $\mathbf{w}/\|\mathbf{w}\|$ is a unit vector in the direction of \mathbf{w} .

Exercise 0.0.25

i. Show that the projection of \mathbf{v} on \mathbf{w} has the same direction as \mathbf{w} if $\langle \mathbf{v}, \mathbf{w} \rangle > 0$, and the direction of $-\mathbf{w}$ if $\langle \mathbf{v}, \mathbf{w} \rangle < 0$.

ii. Show that if $\langle \mathbf{v}, \mathbf{w} \rangle = 0$, then \mathbf{v} and \mathbf{w} are orthogonal and the projection is just the zero vector.

Exercise 0.0.26 Prove that $\|\text{proj}_{\mathbf{w}}(\mathbf{v})\| = \frac{|\langle \mathbf{v}, \mathbf{w} \rangle|}{\|\mathbf{w}\|}$.

Exercise 0.0.27 Show that $\mathbf{v} - \text{proj}_{\mathbf{w}}(\mathbf{v})$ is orthogonal to \mathbf{w} .

Definition 0.0.28 Given two linearly independent vectors $\mathbf{v} = (v_1, v_2, v_3)$ and $\mathbf{w} = (w_1, w_2, w_3)$ in \mathbb{E}^3 , we define the *cross-product* to be $\mathbf{v} \times \mathbf{w} = (v_2w_3 - v_3w_2, v_3w_1 - v_1w_3, v_1w_2 - v_2w_1)$.

Exercise 0.0.29

i. Show that $\mathbf{v} \times \mathbf{w} \neq \mathbf{0}$.

ii. Show that $\langle \mathbf{v}, \mathbf{v} \times \mathbf{w} \rangle = \langle \mathbf{w}, \mathbf{v} \times \mathbf{w} \rangle = 0$.

iii. Show that $\mathbf{w} \times \mathbf{v} = -(\mathbf{v} \times \mathbf{w})$.

Exercise 0.0.30

i. Show that $\|\mathbf{v} \times \mathbf{w}\| = \|\mathbf{v}\| \|\mathbf{w}\| \sin \theta$, where θ is the angle between \mathbf{v} and \mathbf{w} .

ii. Show that $\|\mathbf{v} \times \mathbf{w}\|$ is the area of the parallelogram spanned by \mathbf{v} and \mathbf{w} .

Definition 0.0.31 Let F be a subfield of the real numbers, and let $V = F^n$. If $\mathbf{v} = (v_1, \dots, v_n)$ and $\mathbf{w} = (w_1, \dots, w_n)$ are elements of V , we say that \mathbf{v} is *orthogonal* (or *perpendicular*) to \mathbf{w} if $\langle \mathbf{v} | \mathbf{w} \rangle = \mathbf{v} \cdot \mathbf{w} = v_1w_1 + v_2w_2 + \dots + v_nw_n = 0$. We say that \mathbf{v} is a *normalized vector* or *unit vector*, if $\langle \mathbf{v}, \mathbf{v} \rangle = v_1^2 + v_2^2 + \dots + v_n^2 = 1$. An *orthonormal set* in V is a collection $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k\}$ of linearly independent vectors in V such that $\langle \mathbf{v}_i | \mathbf{v}_j \rangle = 0$ if $i \neq j$, and $\langle \mathbf{v}_i | \mathbf{v}_i \rangle = 1$ for all $i, j = 1, \dots, k$. If $k = n$, the orthonormal set $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n\}$ is called an *orthonormal basis* for V .

Theorem 0.0.32 Let F be a subfield of \mathbb{R} , and let $V = F^n$. Suppose that $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}$ is a collection of linearly independent vectors in V , where $\mathbf{v}_j = (v_{j1}, v_{j2}, \dots, v_{jn})$. Consider the matrix

$$\begin{pmatrix} \mathbf{e}_1 & \mathbf{e}_2 & \cdots & \mathbf{e}_n \\ v_{11} & v_{12} & \cdots & v_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ v_{(n-1)1} & v_{(n-1)2} & \cdots & v_{(n-1)n} \end{pmatrix}.$$

Let \mathbf{v} be the vector obtained by taking the determinant of this matrix with respect to the first row. Then \mathbf{v} is nonzero and is orthogonal to each of the vectors $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}$. Moreover, the collection $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}, \mathbf{v}\}$ is a basis for V .

Exercise 0.0.33 If, as in the previous theorem, $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}\}$ forms an orthonormal set in V , show that $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}, \mathbf{v}\}$ is an orthonormal basis for V . (Hint: Consider $A^t A$ and AA^{-1} .)

Exercise 0.0.34 Suppose that $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}$ are linearly independent vectors in \mathbb{E}^n . Take a point \mathbf{p}_0 in \mathbb{R}^n and consider the hyperplane \mathbf{H} through \mathbf{p}_0 spanned by $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}$. If \mathbf{v} is the vector determined in Theorem 0.0.32, show that $\mathbf{H} = \{\mathbf{p} \in \mathbb{R}^n \mid \langle \mathbf{p} - \mathbf{p}_0, \mathbf{v} \rangle = 0\}$. Specialize this to obtain formulas for a line in \mathbb{R}^2 and for a plane in \mathbb{R}^3 .

Exercise 0.0.35 Let $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}$ be linearly independent vectors in \mathbb{E}^n . Let \mathbf{p}_0 be a point of \mathbb{R}^n . Let \mathbf{H} be the hyperplane through \mathbf{p}_0 spanned by $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_{n-1}$. If \mathbf{p} is any point in \mathbb{R}^n , show that the distance from \mathbf{p} to \mathbf{H} , that is $\inf\{\|\mathbf{p} - \mathbf{q}\| \mid \mathbf{q} \in \mathbf{H}\}$, is given by the length of the vector $\text{proj}_{\mathbf{v}}(\mathbf{p} - \mathbf{p}_0)$ where \mathbf{v} is the vector obtained in Theorem 0.0.32. Specialize this to obtain formulas for the distance from a point to a line in \mathbb{R}^2 and from a point to a plane in \mathbb{R}^3 .

Exercise 0.0.36 Find a formula for the distance from a point to a line in \mathbb{R}^n .

Definition 0.0.37 The *Gram-Schmidt Orthogonalization Process* is the following algorithm.

Given a set $\{\mathbf{v}_1, \dots, \mathbf{v}_k\}$ of linearly independent vectors, we proceed as follows. Let $\tilde{\mathbf{v}}_1 = \mathbf{v}_1$. We continue to find vectors $\tilde{\mathbf{v}}_k$ by taking \mathbf{v}_k and subtracting the projections on the vectors already constructed. More explicitly, we let

$$\tilde{\mathbf{v}}_2 = \mathbf{v}_2 - \text{proj}_{\tilde{\mathbf{v}}_1}(\mathbf{v}_2). \quad (3)$$

$$\tilde{\mathbf{v}}_3 = \mathbf{v}_3 - \text{proj}_{\tilde{\mathbf{v}}_1}(\mathbf{v}_3) - \text{proj}_{\tilde{\mathbf{v}}_2}(\mathbf{v}_3). \quad (4)$$

$$\vdots \quad (5)$$

$$\tilde{\mathbf{v}}_k = \mathbf{v}_k - \sum_{i=1}^{k-1} \text{proj}_{\tilde{\mathbf{v}}_i}(\mathbf{v}_k). \quad (6)$$

Exercise 0.0.38 Check that the set of vectors above is pairwise orthogonal, and, in addition, show that $\tilde{\mathbf{v}}_1, \tilde{\mathbf{v}}_2, \dots, \tilde{\mathbf{v}}_k$ span the same subspace as $\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_k$.

Exercise 0.0.39 Consider the vectors $\mathbf{v}_1 = (1, 1, -1, 0)$ and $\mathbf{v}_2 = (1, 0, 0, -1)$ in \mathbb{E}^4 , and complete this pair to an orthogonal basis as above.

Exercise 0.0.40 Show that the collection $\{\mathbf{e}_1, \mathbf{e}_2, \dots, \mathbf{e}_n\}$ is an orthonormal basis for \mathbb{E}^n .

Exercise 0.0.41 Show that, given linearly independent vectors $\mathbf{v}_1, \dots, \mathbf{v}_k$ in \mathbb{E}^n , we can transform this collection into an orthonormal set $\{\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_k\}$, which spans the same subspace. In addition, the set $\{\tilde{\mathbf{v}}_1, \dots, \tilde{\mathbf{v}}_k\}$ can be completed to an orthonormal basis for \mathbb{E}^n .