Lecture 1.4: Inner products and orthogonality

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Basic Euclidean geometry

Definition

Let
$$V = \mathbb{R}^n$$
. The dot product of $\mathbf{v} = (a_1, \dots, a_n)$ and $\mathbf{w} = (b_1, \dots, b_n)$ is $\mathbf{v} \cdot \mathbf{w} = \sum_{i=1}^n a_i b_i$.

The length (or "norm") of $\mathbf{v} \in \mathbb{R}^n$, denoted $||\mathbf{v}||$, is the distance from \mathbf{v} to $\mathbf{0}$:

$$||\mathbf{v}|| = \sqrt{\mathbf{v} \cdot \mathbf{v}} = \sqrt{a_1^2 + \dots + a_n^2}.$$

To understand what $\mathbf{v} \cdot \mathbf{w}$ means geometrically, we can pick a "special" \mathbf{v} and \mathbf{w} .

- Pick **v** to be on the x-axis (i.e., $\mathbf{v} = a_1 \mathbf{e}_1$).
- Pick w to be in the xy-plane (i.e., $\mathbf{w} = b_1 \mathbf{e}_1 + b_2 \mathbf{e}_2$).

By basic trigonometry,

$$\mathbf{v} = \Big(||\mathbf{v}|| \ , \ 0 \ , \ 0 \ , \ldots \ , \ 0 \Big), \qquad \mathbf{w} = \Big(||\mathbf{w}||\cos\theta \ , \ ||\mathbf{w}||\sin\theta \ , \ 0 \ , \ldots \ , \ 0 \Big).$$

Proposition

The dot product of any two vectors $\mathbf{v}, \mathbf{w} \in \mathbb{R}^n$ satisfies $\mathbf{v} \cdot \mathbf{w} = ||\mathbf{v}|| ||\mathbf{w}|| \cos \theta$. Equivalently, the angle θ between them is

$$\cos\theta = \frac{\mathbf{v}\cdot\mathbf{w}}{||\mathbf{v}||\,||\mathbf{w}||}.$$

Basic Euclidean geometry

The following relations follow immediately:

$$(\mathbf{v} + \mathbf{w}) \cdot (\mathbf{v} + \mathbf{w}) = \mathbf{v} \cdot \mathbf{v} + 2 \mathbf{v} \cdot \mathbf{w} + \mathbf{w} \cdot \mathbf{w} = ||\mathbf{v} + \mathbf{w}||^2,$$

$$\mathbf{v} \cdot (\mathbf{v} - \mathbf{w}) \cdot (\mathbf{v} - \mathbf{w}) = \mathbf{v} \cdot \mathbf{v} - 2\mathbf{v} \cdot \mathbf{w} + \mathbf{w} \cdot \mathbf{w} = ||\mathbf{v} - \mathbf{w}||^2$$

Law of cosines

The last equation above says

$$||\mathbf{v}||^2 - 2 ||\mathbf{v}|| ||\mathbf{w}|| \cos \theta + ||\mathbf{w}||^2 = ||\mathbf{v} - \mathbf{w}||^2,$$

which is the law of cosines.

For any unit vector $\mathbf{n} \in \mathbb{R}^n$ ($||\mathbf{n}|| = 1$), the projection of $\mathbf{v} \in \mathbb{R}^n$ onto \mathbf{n} is $\text{proj}_{\mathbf{n}}(\mathbf{v}) = \mathbf{v} \cdot \mathbf{n}$.

For example, consider $\mathbf{v} = (4,3) = 4\mathbf{e}_1 + 3\mathbf{e}_2$ in \mathbb{R}^2 . Note that

$$\operatorname{proj}_{\mathbf{e}_1}(\mathbf{v}) = (4,3) \cdot (1,0) = 4, \quad \operatorname{proj}_{\mathbf{e}_2}(\mathbf{v}) = (4,3) \cdot (0,1) = 3.$$

Big idea

By defining the "dot product" in \mathbb{R}^n , we get for free a notion of geometry. That is, we get natural definitions of concepts such as length, angles, and projection.

To do this in other vector spaces, we need a generalized notion of "dot product."

Inner products

Definition

Let V be an \mathbb{R} -vector space. A function $\langle -, - \rangle \colon V \times V \to \mathbb{R}$ is a (real) inner product if it satisfies (for all $\mathbf{u}, \mathbf{v}, \mathbf{w} \in V$, $c \in \mathbb{R}$):

(i)
$$\langle \mathbf{u} + \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{u}, \mathbf{v} \rangle + \langle \mathbf{v}, \mathbf{w} \rangle$$

(ii)
$$\langle c\mathbf{v}, \mathbf{w} \rangle = c \langle \mathbf{v}, \mathbf{w} \rangle$$

(iii)
$$\langle \mathbf{v}, \mathbf{w} \rangle = \langle \mathbf{w}, \mathbf{v} \rangle$$

(iv) $\langle \mathbf{v}, \mathbf{v} \rangle \geq 0$, with equaility if and only if $\mathbf{v} = 0$.

Conditions (i)-(ii) are called bilinearity, (iii) is symmetry, and (iv) is positivity.

Remark

Defining an inner product gives rise to a geometry, i.e., notions of length, angle, and projection.

 $\blacksquare \text{ length: } ||\mathbf{v}|| := \sqrt{\langle \mathbf{v}, \mathbf{v} \rangle}.$

angle:
$$\measuredangle(\mathbf{v}, \mathbf{w}) = \theta$$
, where $\cos \theta = \frac{\langle \mathbf{v}, \mathbf{w} \rangle}{||\mathbf{v}|| ||\mathbf{w}||}$.

projection: if $||\mathbf{n}|| = 1$, then we can project **v** onto **n** by defining

$$\operatorname{proj}_{\mathbf{n}}(\mathbf{v}) = \langle \mathbf{v}, \mathbf{n} \rangle, \qquad \operatorname{Proj}_{\mathbf{n}}(\mathbf{v}) = \langle \mathbf{v}, \mathbf{n} \rangle \mathbf{n}.$$

This is the length or magnitude, of v in the n-direction.

Orthogonality

Definition

Two vectors $\mathbf{v}, \mathbf{w} \in V$ are orthogonal if $\langle \mathbf{v}, \mathbf{w} \rangle = 0$. A set $\{\mathbf{v}_1, \dots, \mathbf{v}_n\} \subseteq V$ is orthonormal if $\langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0$ for all $i \neq j$ and $||\mathbf{v}_i|| = 1$ for all *i*.

Key idea

- Orthogonal is the abstract version of "*perpendicular*."
- Orthonormal means "perpendicular and unit length." An equivalent definition is

$$\langle \mathbf{v}_i, \mathbf{v}_j \rangle = \begin{cases} 0 & i \neq j \\ 1 & i = j. \end{cases}$$

Orthonormal bases are really desirable!

Examples

1. Let $V = \mathbb{R}^n$. The standard "dot product" $\langle \mathbf{v}, \mathbf{w} \rangle = \mathbf{v} \cdot \mathbf{w} = \sum_{i=1}^n v_i w_i$ is an inner product.

The set $\{e_1,\ldots,e_n\}$ is an orthonormal basis of $\mathbb{R}^n.$ We can write each $\textbf{v}\in\mathbb{R}^n$ uniquely as

$$\mathbf{v} = (a_1, \dots, a_n) := a_1 \mathbf{e}_1 + \dots + a_n \mathbf{e}_n, \quad \text{where } a_i = \operatorname{proj}_{\mathbf{e}_i}(\mathbf{v}) = \mathbf{v} \cdot \mathbf{e}_i.$$

Orthonormal bases

Examples

2. Consider $V = \operatorname{Per}_{2\pi}(\mathbb{C})$. We can define an inner product as

$$\langle f,g\rangle = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x)\overline{g(x)} \, dx.$$

The set

$$\left\{e^{inx}\mid n\in\mathbb{Z}\right\}=\left\{\ldots,\,e^{-2ix},\,e^{-ix},\,1,\,e^{ix},\,e^{2ix},\ldots\right\}$$

is an orthonormal basis w.r.t. to this inner product.

Thus, we can write each $f(x) \in \operatorname{Per}_{2\pi}$ uniquely as

$$f(x) = \sum_{n = -\infty}^{\infty} c_n e^{inx} = c_0 + \sum_{n=1}^{\infty} c_n e^{inx} + c_{-n} e^{-inx}$$

where

$$c_n = \operatorname{proj}_{e^{inx}}(f) = \langle f, e^{inx} \rangle = \frac{1}{2\pi} \int_{-\pi}^{\pi} f(x) e^{-inx} dx.$$

Orthonormal bases

Examples

3. Consider $V = \operatorname{Per}_{2\pi}(\mathbb{R})$. We can define an inner product as

$$\langle f,g\rangle = \frac{1}{\pi}\int_{-\pi}^{\pi}f(x)g(x)\,dx.$$

The set

$$\left\{\frac{1}{\sqrt{2}}, \cos x, \cos 2x, \ldots\right\} \cup \left\{\sin x, \sin 2x, \ldots\right\}.$$

is an orthonormal basis w.r.t. to this inner product.

Thus, we can write each $f(x) \in \operatorname{Per}_{2\pi}$ uniquely as

$$f(x) = \frac{a_0}{2} + \sum_{n=1}^{\infty} a_n \cos nx + b_n \sin nx,$$

where

$$a_n = \operatorname{proj}_{\cos nx} (f) = \langle f, \cos nx \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \cos nx \, dx$$
$$b_n = \operatorname{proj}_{\sin nx} (f) = \langle f, \sin nx \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f(x) \sin nx \, dx$$

Orthogonal bases

Important remark

Sometimes we have an orthogonal (but *not* orthonormal) basis, $\mathbf{v}_1, \ldots, \mathbf{v}_n$.

There is still a simple way to decompose a vector $\mathbf{v} \in V$ into this basis.

Note that
$$\left\{ \frac{\mathbf{v}_1}{||\mathbf{v}_1||}, \dots, \frac{\mathbf{v}_n}{||\mathbf{v}_n||} \right\}$$
 is an orthonormal basis, so
 $\mathbf{v} = \mathbf{a}_1 \frac{\mathbf{v}_1}{||\mathbf{v}_1||} + \dots + \mathbf{a}_n \frac{\mathbf{v}_n}{||\mathbf{v}_n||}$
 $\mathbf{a}_i = \left\langle \mathbf{v}, \frac{\mathbf{v}_i}{||\mathbf{v}_i||} \right\rangle = \frac{1}{||\mathbf{v}_i||} \left\langle \mathbf{v}, \mathbf{v}_i \right\rangle = \frac{\left\langle \mathbf{v}, \mathbf{v}_i \right\rangle}{\sqrt{\left\langle \mathbf{v}_i, \mathbf{v}_i \right\rangle}}$
 $= \frac{\mathbf{a}_1}{||\mathbf{v}_1||} \mathbf{v}_1 + \dots + \frac{\mathbf{a}_n}{||\mathbf{v}_n||} \mathbf{v}_n$
 $= c_1 \mathbf{v}_1 + \dots + c_n \mathbf{v}_n,$
 $c_i = \frac{\mathbf{a}_i}{||\mathbf{v}_i||} = \frac{\left\langle \mathbf{v}, \mathbf{v}_i \right\rangle}{\left\langle \mathbf{v}_i, \mathbf{v}_i \right\rangle} = \frac{\left\langle \mathbf{v}, \mathbf{v}_i \right\rangle}{||\mathbf{v}_i||^2}$