Orthonormal Sets



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Orthogonal sets

Definition

A set $\{\mathbf{v}_1,\ldots,\mathbf{v}_n\}$ of nonzero vectors in a vector space V forms an **orthogonal set** w.r.t. the inner product $\langle\cdot,\cdot\rangle$ if $\langle\mathbf{v}_i,\mathbf{v}_j\rangle=0$ whenever $i\neq j$.

Example 1. Let
$$A = \begin{bmatrix} 1 & 1 & 1 & 0 \\ 1 & 1 - 1 & 0 \\ 1 - 1 & 0 & 1 \\ 1 - 1 & 0 - 1 \end{bmatrix}$$

Do the columns a_j of A form an orthogonal set in Euclidean \mathbb{R}^4 w.r.t. the standard Euclidean inner product?

YES since $\langle \mathbf{a}_i, \mathbf{a}_i \rangle = \mathbf{a}_i^T \mathbf{a}_i = 0$ whenever $i \neq j$. This is equivalent to

$$A^{T}A = \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & 1 - 1 - 1 \\ 1 - 1 & 0 & 0 \\ 0 & 0 & 1 - 1 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 0 \\ 1 & 1 - 1 & 0 \\ 1 - 1 & 0 & 1 \\ 1 - 1 & 0 - 1 \end{bmatrix} = \begin{bmatrix} 4 & 0 & 0 & 0 \\ 0 & 4 & 0 & 0 \\ 0 & 0 & 2 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix}$$
 diagonal.

Property Orthogonal sets are linearly independent.

Proof: If
$$x_1 \cdot \mathbf{v}_1 + \cdots + x_n \cdot \mathbf{v}_n = \mathbf{0}$$
 then, for $j = 1, \dots, n$,

$$0 = \langle \mathbf{0}, \mathbf{v}_j \rangle = \langle x_1 \cdot \mathbf{v}_1 + \dots + x_n \cdot \mathbf{v}_n, \mathbf{v}_j \rangle = x_1 \langle \mathbf{v}_1, \mathbf{v}_j \rangle + \dots + x_n \langle \mathbf{v}_n, \mathbf{v}_j \rangle = x_j ||\mathbf{v}_j||^2 \Rightarrow x_j = 0 \checkmark$$

Orthonormal Sets

Definition

An **orthonormal set** (ON) of vectors is an **orthogonal** set of **unit/normalized** vectors (w.r.t. to the norm induced by the inner product)

Example 2. Let
$$A = \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 \end{bmatrix} = \begin{bmatrix} \cos \theta - \sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$
.

Do the columns $\{a_1,a_2\}$ of A form an ON set w.r.t. the Euclidean inner product in \mathbb{R}^2 ?

YES since

•
$$\mathbf{a}_{2}^{T}\mathbf{a}_{1} = \begin{bmatrix} -\sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} \cos\theta \\ \sin\theta \end{bmatrix} = -\sin\theta\cos\theta + \sin\theta\cos\theta = 0$$

The orthonormality of $\{a_1, a_2\}$ is equivalent to

$$A^{T}A = \begin{bmatrix} \cos \theta - \sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} \cos \theta - \sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} = \begin{bmatrix} \cos^{2} \theta + \sin^{2} \theta & 0 \\ 0 & \sin^{2} \theta + \cos^{2} \theta \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I_{2}$$

Example 3. Does $\{p_0, p_1, p_2\} = \left\{\sqrt{\frac{1}{2}}, \sqrt{\frac{3}{2}} t, \sqrt{\frac{5}{2}} \left(\frac{3}{2} t^2 - \frac{1}{2}\right)\right\}$ form an ON basis of \mathbb{P}_3 (equipped with standard operations), w.r.t.

$$\langle p, q \rangle = \int_{-1}^{1} p(t)q(t)dt$$
?

$$\|p_0\|^2 = \left\|\sqrt{\frac{1}{2}}\right\|^2 = \int_{-1}^1 \left(\sqrt{\frac{1}{2}}\right)^2 dt = 1 \quad \checkmark$$

$$\langle p_0, p_1 \rangle = \left\langle\sqrt{\frac{1}{2}}, \sqrt{\frac{3}{2}}t\right\rangle = \int_{-1}^1 \frac{\sqrt{3}}{2}t \, dt = 0 \quad \checkmark$$

$$\langle p_0, p_2 \rangle = \left\langle\sqrt{\frac{1}{2}}, \sqrt{\frac{5}{2}}\left(\frac{3}{2}t^2 - \frac{1}{2}\right)\right\rangle = \int_{-1}^1 \frac{\sqrt{5}}{2}\left(\frac{3}{2}t^2 - \frac{1}{2}\right) \, dt = 0 \quad \checkmark$$

$$\|p_1\|^2 = \left\|\sqrt{\frac{3}{2}}t\right\|^2 = \int_{-1}^{1} \frac{3}{2}t^2 dt = 1$$

$$\langle p_1, p_2 \rangle = \left\langle \sqrt{\frac{3}{2}} t, \sqrt{\frac{5}{2}} \left(\frac{3}{2} t^2 - \frac{1}{2} \right) \right\rangle = \int_{-1}^{1} \frac{\sqrt{15}}{2} t \left(\frac{3}{2} t^2 - \frac{1}{2} \right) dt = 0 \quad \checkmark$$

$$\|p_2\|^2 = \left\| \sqrt{\frac{5}{2}} \left(\frac{3}{2} t^2 - \frac{1}{2} \right) \right\|^2 = \int_{-1}^{1} \frac{5}{2} \left(\frac{3}{2} t^2 - \frac{1}{2} \right)^2 dt = 1 \quad \checkmark$$

$$\Rightarrow$$
 YES, $\{p_0, p_1, p_2\}$ **IS** an ON basis of \mathbb{P}_3 w.r.t. (\star)

Example 4. Does $\{\sin t, \cos t\}$ form an ON set in $\mathcal{C}([-\pi, \pi])$ w.r.t. the inner product

$$\langle f,g\rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} f(t)g(t) dt$$
?

$$\langle \sin t, \cos t \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \sin t \cos t \, dt = 0$$

$$\pi J_{-\pi}$$

$$\|\sin t\|^2 = \langle \sin t, \sin t \rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \sin^2 t \, dt = \frac{1}{2\pi} \int_{-\pi}^{\pi} (1 - \cos 2t) \, dt = 1$$

$$\|\cos t\|^2 = \langle\cos t, \cos t\rangle = \frac{1}{\pi} \int_{-\pi}^{\pi} \cos^2 t \, dt = \frac{1}{2\pi} \int_{-\pi}^{\pi} (1 + \cos 2t) \, dt = 1$$

$$\Rightarrow$$
 YES, $\{\sin t, \cos t\}$ IS an ON set in $\mathcal{C}([-\pi, \pi])$ w.r.t (\star)

Properties of ON bases

If $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is an orthonormal basis of V then

Property 1 The matrix representation C of $\langle \cdot, \cdot \rangle$ is the identity matrix

Proof:
$$C = \begin{bmatrix} \|\mathbf{v}_1\|^2 & \cdots & \langle \mathbf{v}_1, \mathbf{v}_n \rangle \\ \vdots & \vdots \\ \langle \mathbf{v}_1, \mathbf{v}_n \rangle & \cdots & \|\mathbf{v}_n\|^2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ \ddots & 0 \\ 0 & 1 \end{bmatrix} = I_n$$

Property 2 $\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{y}^T \mathbf{x}$ where $\mathbf{x} = [\mathbf{u}]_{\mathcal{B}}$ and $\mathbf{y} = [\mathbf{v}]_{\mathcal{B}}$

The inner product of two vectors is the Euclidean inner product of their representations

Proof: In general: $\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{y}^T C \mathbf{x}$, where C is the matrix representation of the inner product (see previous section). By Property 1, $C = I_n$ and the statement follows.

Property 3 If $\mathbf{u} = x_1 \cdot \mathbf{v}_1 + \cdots + x_n \cdot \mathbf{v}_n$, then $x_j = \langle \mathbf{u}, \mathbf{v}_j \rangle$ for $1 \le j \le n$. The coefficients of the linear combination are easily determined

Proof: Since
$$[\mathbf{v}_j]_{\mathcal{B}} = \mathbf{e}_j$$
, letting $\mathbf{v} = \mathbf{v}_j$ in Property 2 yields $\langle \mathbf{u}, \mathbf{v}_j \rangle = \mathbf{e}_j^\mathsf{T} \mathbf{x} = x_j$ for $1 \le j \le n$

Property 4 $\|\mathbf{u}\|^2 = \|\mathbf{x}\|_2^2$ where $\mathbf{x} = [\mathbf{u}]_{\mathcal{B}}$ (Parseval's Formula)

Proof: Letting $\mathbf{v} = \mathbf{u}$ in Property 2 yields $\|\mathbf{u}\|^2 = \langle \mathbf{u}, \mathbf{u} \rangle = \mathbf{x}^T \mathbf{x} = \|\mathbf{x}\|_2^2$

Property 5 In $V = \mathbb{R}^n$ with Euclidean inner product $\langle \mathbf{u}, \mathbf{v} \rangle = \mathbf{v}^T \mathbf{u}$:

Let
$$A = \begin{bmatrix} \mathbf{v}_1 & \mathbf{v}_n \\ \mathbf{v}_n \end{bmatrix}$$
. If $\mathbf{u} = x_1 \mathbf{v}_1 + \dots + x_n \mathbf{v}_n = A\mathbf{x}$
then $\mathbf{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} \mathbf{v}_1^T \mathbf{u} \\ \vdots \\ \mathbf{v}_n^T \mathbf{u} \end{bmatrix} = A^T \mathbf{u}$

Proof: The result follows from the fact that if $\mathcal{B} = \{\mathbf{v}_1, \dots, \mathbf{v}_n\}$ is an orthonormal basis of \mathbb{R}^n , then $A^T A = AA^T = I_n$

Example 5. Write
$$\begin{bmatrix} 1\\1\\1 \end{bmatrix}$$
 as a LC of $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\} = \left\{ \begin{bmatrix} \frac{2}{3}\\-\frac{2}{3}\\\frac{1}{3} \end{bmatrix}, \begin{bmatrix} \frac{1}{3}\\\frac{2}{3}\\\frac{2}{3} \end{bmatrix} \right\}$ in Euclidean \mathbb{R}^3

Verify that

$$A = \begin{bmatrix} \mathbf{v}_1 & \mathbf{v}_2 & \mathbf{v}_3 \end{bmatrix} = \begin{bmatrix} \frac{2}{3} & \frac{2}{3} & \frac{1}{3} \\ -\frac{2}{3} & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{3} & -\frac{2}{3} & \frac{2}{3} \end{bmatrix} \text{ satisfies } A^T A = I, \text{ i.e., } A^{-1} = A^T$$

i.e., $\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$ forms an ON basis w.r.t. $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{y}^T \mathbf{x}$ in Euclidean \mathbb{R}^3 . Then

i.e.,
$$\{\mathbf{v}_1, \mathbf{v}_2, \mathbf{v}_3\}$$
 forms an ON basis w.r.t. $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{y}^{\top} \mathbf{x}$ in Euclidean \mathbb{R}^2 . Then

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} - \mathbf{y}_1 \begin{bmatrix} \frac{2}{3} \\ -2 \end{bmatrix} + \mathbf{y}_2 \begin{bmatrix} \frac{2}{3} \\ 1 \end{bmatrix} + \mathbf{y}_2 \begin{bmatrix} \frac{1}{3} \\ 2 \end{bmatrix} - \mathbf{A} \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{y}_2 \end{bmatrix} \rightarrow \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{y}_2 \end{bmatrix} - \mathbf{A}^T \begin{bmatrix} 1 \\ 1 \end{bmatrix} - \begin{bmatrix} \frac{1}{3} \\ 1 \end{bmatrix}$$

 $\begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} = x_1 \begin{bmatrix} \frac{2}{3} \\ -\frac{2}{3} \\ 1 \end{bmatrix} + x_2 \begin{bmatrix} \frac{2}{3} \\ \frac{1}{3} \\ 2 \end{bmatrix} + x_3 \begin{bmatrix} \frac{1}{3} \\ \frac{2}{3} \\ 2 \end{bmatrix} = A \begin{bmatrix} x_1 \\ x_2 \\ x_2 \end{bmatrix} \Rightarrow \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = A^T \begin{bmatrix} 1 \\ 1 \\ \frac{1}{3} \end{bmatrix}$

$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} = x_1 \begin{bmatrix} -\frac{1}{3} \\ \frac{1}{3} \end{bmatrix} + x_2 \begin{bmatrix} \frac{1}{3} \\ -\frac{2}{3} \end{bmatrix} + x_3 \begin{bmatrix} \frac{1}{3} \\ \frac{2}{3} \end{bmatrix} = A \begin{bmatrix} x_2 \\ x_3 \end{bmatrix} \implies \begin{bmatrix} x_2 \\ x_3 \end{bmatrix} = A \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} \frac{1}{3} \\ \frac{5}{3} \end{bmatrix}$$
i.e.,
$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} \frac{2}{3} \\ -\frac{2}{3} \end{bmatrix} + \frac{1}{3} \begin{bmatrix} \frac{2}{3} \\ \frac{1}{3} \end{bmatrix} + \frac{5}{3} \begin{bmatrix} \frac{1}{3} \\ \frac{2}{3} \end{bmatrix}$$

$$\begin{bmatrix} \text{Check} \end{bmatrix} \begin{bmatrix} \frac{2}{9} + \frac{2}{9} + \frac{5}{9} \\ -\frac{2}{9} + \frac{1}{9} + \frac{10}{9} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \checkmark$$

i.e.,
$$\begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix} = \frac{1}{3} \begin{bmatrix} \frac{2}{3} \\ -\frac{2}{3} \\ \frac{1}{3} \end{bmatrix} + \frac{1}{3} \begin{bmatrix} \frac{2}{3} \\ \frac{1}{3} \\ -\frac{2}{3} \end{bmatrix} + \frac{5}{3} \begin{bmatrix} \frac{1}{3} \\ \frac{2}{3} \\ \frac{2}{3} \end{bmatrix}$$

$$\boxed{\textbf{Check}} \begin{bmatrix} \frac{2}{9} + \frac{2}{9} + \frac{5}{9} \\ -\frac{2}{9} + \frac{1}{9} + \frac{10}{9} \\ \frac{1}{9} - \frac{2}{9} + \frac{10}{9} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix} \checkmark$$

Example 6. Use the fact that $\{\sin t, \cos t\}$ forms an ON set in $C([-\pi, \pi])$ w.r.t.

$$\langle f,g\rangle=rac{1}{\pi}\int_{-\pi}^{\pi}f(t)g(t)\ dt$$

to determine $\langle f,g\rangle$, $\|f\|$, $\|g\|$ for $f(t)=3\cos t+2\sin t$ and $g(t)=\cos t-\sin t$

The representation vectors of f and g w.r.t. the given ON set are

$$\mathbf{x} = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$$
 and $\mathbf{y} = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$,

respectively.

By Property 2

$$\langle f, g \rangle = \mathbf{y}^T \mathbf{x} = \begin{bmatrix} -1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \end{bmatrix} = \boxed{1}$$

2 By Parseval formula

$$||f||^2 = \mathbf{x}^T \mathbf{x} = 2^2 + 3^2 = 13, \quad ||g||^2 = \mathbf{y}^T \mathbf{y} = (-1)^2 + 1^2 = 2$$

Orthogonality and Least Squares

Theorem

Let
$$A = \begin{vmatrix} \mathbf{a}_1 & \cdots & \mathbf{a}_n \end{vmatrix} \in \mathbb{R}^{m \times n}$$
 s.t. $A^T A = I_n$. Then

- $m \ge n$ and $\{\mathbf{a}_1, \ldots, \mathbf{a}_n\}$ is an ON set w.r.t. $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{y}^T \mathbf{x}$ in Euclidean \mathbb{R}^m
- The least squares solution to Ax = b is

$$\mathbf{x} = (A^T A)^{-1} A^T \mathbf{b} = A^T \mathbf{b} = \begin{bmatrix} \mathbf{a}_1^T \mathbf{b} \\ \vdots \\ \mathbf{a}_n^T \mathbf{b} \end{bmatrix}$$

• The orthogonal projection of **b** onto R(A) is

$$\mathbf{p} = A\mathbf{x} = A(A^T\mathbf{b}) = (\mathbf{a}_1^T\mathbf{b})\mathbf{a}_1 + \ldots + (\mathbf{a}_n^T\mathbf{b})\mathbf{a}_n$$

• p = Pb where

$$P = A(A^{T}A)^{-1}A^{T} = AA^{T} = \mathbf{a}_{1}\mathbf{a}_{1}^{T} + \cdots + \mathbf{a}_{n}\mathbf{a}_{n}^{T}$$

is the matrix of the orthogonal projection onto R(A)

Orthogonal projection on subspaces of general vector spaces

Theorem

Let $\mathbf{b} \in V$ and $\{\mathbf{u}_1, \dots, \mathbf{u}_n\}$ be an ON basis for a subspace S of $(V, +, \cdot)$.

Then the orthogonal projection **p** of **b** onto S is given by

$$\label{eq:posterior} \boldsymbol{p} = \langle \boldsymbol{b}, \boldsymbol{u}_1 \rangle \cdot \boldsymbol{u}_1 + \dots + \langle \boldsymbol{b}, \boldsymbol{u}_n \rangle \cdot \boldsymbol{u}_n.$$

Properties:

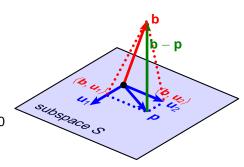
- $\mathbf{0}$ $\mathbf{p} \in S$
- $\langle \mathbf{p}, \mathbf{u} \rangle = \langle \mathbf{b}, \mathbf{u} \rangle$ for all $\mathbf{u} \in \mathcal{S}$

This is equivalent to

$$\langle \mathbf{b} - \mathbf{p}, \mathbf{u} \rangle = \langle \mathbf{b}, \mathbf{u} \rangle - \langle \mathbf{p}, \mathbf{u} \rangle = 0$$

for all $\mathbf{u} \in S$, i.e., $\mathbf{b} - \mathbf{p} \perp S$.

3 The induced norm $\|\mathbf{b} - \mathbf{u}\|$ is minimum for \mathbf{u} in S when $\mathbf{u} = \mathbf{p}$



Example 7. Use the fact that $\{\mathbf{u}_1, \mathbf{u}_2\} = \{1, \sqrt{3}(2t-1)\}$ is an ON basis of

$$\mathbb{P}_2$$
 w.r.t. $\langle p,q \rangle = \int_0^1 p(t)q(t) \ dt$ to find the \perp projection p of t^2 onto \mathbb{P}_2

• Verify $\{\mathbf{u}_1, \mathbf{u}_2\}$ is an ON set:

$$\langle \mathbf{u}_1, \mathbf{u}_2 \rangle = \int_0^1 \sqrt{3}(2t-1)dt = 0, \ \|\mathbf{u}_1\|^2 = \int_0^1 dt = 1, \ \|\mathbf{u}_2\|^2 = \int_0^1 3(2t-1)^2 dt = 1$$

2
$$p(t) = \langle t^2, \mathbf{u}_1 \rangle \mathbf{u}_1 + \langle t^2, \mathbf{u}_2 \rangle \mathbf{u}_2$$
 with

$$\langle t^2, \mathbf{u}_1 \rangle = \langle t^2, 1 \rangle = \int_0^1 t^2 \ dt = \frac{1}{3}$$
$$\langle t^2, \mathbf{u}_2 \rangle = \langle t^2, \sqrt{3}(2t - 1) \rangle$$

$$= \int_0^1 t^2 \sqrt{3}(2t - 1) dt = \frac{1}{6}\sqrt{3}$$

$$\Rightarrow p(t) = \frac{1}{3} \cdot 1 + \frac{1}{6} \sqrt{3} \cdot \sqrt{3} (2t - 1)$$
$$= \frac{1}{3} + t - \frac{1}{3} = t - \frac{1}{6}$$



