

Art's Commerce and Science College, Onde Tal:- Vikramgad, Dist:- Palghar USMT 402: Linear Algebra-II

My Inspiration Shri. V.G. Patil Saheb Dr. V. S. Sonawne

Practical No-4
Orthogonal, orthogonalisation

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Orthogonality

Let V be an inner product space. Two vectors $u, v \in V$ are said to be **orthogonal** if

$$\langle \boldsymbol{u}, \boldsymbol{v} \rangle = 0.$$

Example For inner product space $C[-\pi, \pi]$, the functions $\sin t$ and $\cos t$ are orthogonal as

$$\begin{split} \langle \sin t, \cos t \rangle &= \int_{-\pi}^{\pi} \sin t \, \cos t \, \mathrm{d}t \\ &= \left. \frac{1}{2} \sin^2 t \right|^{\pi} = 0 - 0 = 0. \end{split}$$

Example Let $u = [a_1, a_2, \dots, a_n]^T \in \mathbb{R}^n$. The set of all vector of the Euclidean n-space \mathbb{R}^n that are orthogonal to u is a subspace of \mathbb{R}^n . In fact, it is the solution space of the single linear equation

$$\langle u, x \rangle = a_1x_1 + a_2x_2 + \cdots + a_nx_n = 0.$$



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Santosh Shiv Dhamone **Example** Let $\boldsymbol{u} = [1, 2, 3, 4, 5]^T$, $\boldsymbol{v} = [2, 3, 4, 5, 6]^T$, and $\boldsymbol{w} = [1, 2, 3, 3, 2]^T \in \mathbb{R}^5$. The set of all vectors of \mathbb{R}^5 that are orthogonal to $\boldsymbol{u}, \boldsymbol{v}, \boldsymbol{w}$ is a subspace of \mathbb{R}^5 . In fact, it is the solution space of the linear system

$$\begin{cases} x_1 + 2x_2 + 3x_3 + 4x_4 + 5x_5 = 0\\ 2x_1 + 3x_2 + 4x_3 + 5x_4 + 6x_5 = 0\\ x_1 + 2x_2 + 3x_3 + 3x_4 + 2x_5 = 0 \end{cases}$$

Let S be a nonempty subset of an inner product space V. We denote by S^{\perp} the set of all vectors of V that are orthogonal to every vector of S, called the **orthogonal complement** of S in V. In notation,

$$S^{\perp} := \Big\{ \boldsymbol{v} \in V \ \big| \ \langle \boldsymbol{v}, \boldsymbol{u} \rangle = 0 \text{ for all } \boldsymbol{u} \in S \Big\}.$$

If S contains only one vector \boldsymbol{u} , we write

$$u^{\perp} = \{v \in V \mid \langle v, u \rangle = 0\}.$$



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Orthogonal sets and bases

Let V be an inner product space. A subset $S = \{u_1, u_2, \dots, u_k\}$ of nonzero vectors of V is called an **orthogonal set** if every pair of vectors are orthogonal, i.e.,

$$\langle \boldsymbol{u}_i, \boldsymbol{u}_j \rangle = 0, \quad 1 \le i < j \le k.$$

An orthogonal set $S = \{u_1, u_2, \dots, u_k\}$ is called an **orthonormal set** if we further have

$$\|u_i\| = 1, 1 \le i \le k.$$

An **orthonormal basis** of V is a basis which is also an orthonormal set.



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Theorem (Pythagoras). Let v_1, v_2, \dots, v_k be mutually orthogonal vectors. Then

$$\|\boldsymbol{v}_1 + \boldsymbol{v}_2 + \dots + \boldsymbol{v}_k\|^2 = \|\boldsymbol{v}_1\|^2 + \|\boldsymbol{v}_2\|^2 + \dots + \|\boldsymbol{v}_k\|^2.$$

Proof. For simplicity, we assume k=2. If \boldsymbol{u} and \boldsymbol{v} are orthogonal, i.e., $\langle \boldsymbol{u}, \boldsymbol{v} \rangle = 0$, then

$$||\mathbf{u} + \mathbf{v}||^2 = \langle \mathbf{u} + \mathbf{v}, \mathbf{u} + \mathbf{v} \rangle$$
$$= \langle \mathbf{u}, \mathbf{u} \rangle + \langle \mathbf{v}, \mathbf{v} \rangle$$
$$= ||\mathbf{u}||^2 + ||\mathbf{v}||^2.$$





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Example The three vectors

$$\mathbf{v}_1 = [1, 2, 1]^T, \quad \mathbf{v}_2 = [2, 1, -4]^T, \quad \mathbf{v}_3 = [3, -2, 1]^T$$

are mutually orthogonal. Express the the vector $v = [7, 1, 9]^T$ as a linear combination of v_1, v_2, v_3 . Set

$$x_1 v_1 + x_2 v_2 + x_3 v_3 = v.$$



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There are two ways to find x_1, x_2, x_3 .

Method 1: Solving the linear system by performing row operations to its augmented matrix

$$[\boldsymbol{v}_1, \boldsymbol{v}_2, \boldsymbol{v}_3 \mid \boldsymbol{v}],$$

we obtain $x_1 = 3$, $x_2 = -1$, $x_3 = 2$. So $\mathbf{v} = 3\mathbf{v}_1 - \mathbf{v}_2 + 2\mathbf{v}_3$. Method 2: Since $\mathbf{v}_i \perp \mathbf{v}_i$ for $i \neq j$, we have

$$\langle \boldsymbol{v}, \boldsymbol{v}_i \rangle = \langle x_1 \boldsymbol{v}_1 + x_2 \boldsymbol{v}_2 + x_3 \boldsymbol{v}_3, \boldsymbol{v}_i \rangle = x_i \langle \boldsymbol{v}_i, \boldsymbol{v}_i \rangle,$$

where i = 1, 2, 3. Then

$$x_i = \frac{\langle \boldsymbol{v}, \boldsymbol{v}_i \rangle}{\langle \boldsymbol{v}_i, \boldsymbol{v}_i \rangle}, \quad i = 1, 2, 3.$$

We then have

$$\begin{array}{rcl} x_1 & = & \displaystyle \frac{7+2+9}{1+4+1} = \frac{18}{6} = 3, \\ x_2 & = & \displaystyle \frac{14+1-36}{4+1+16} = \frac{-21}{21} = -1, \\ x_3 & = & \displaystyle \frac{21-2+9}{9+4+1} = \frac{28}{14} = 2. \end{array}$$



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Gram-Schmidt process

Let W be a subspace of an inner product space V. Let $\mathcal{B} = \{v_1, v_2, \dots, v_k\}$ be a basis of W, not necessarily orthogonal. An orthogonal basis $\mathcal{B}' = \{w_1, w_2, \dots, w_k\}$ may be constructed from \mathcal{B} as follows:

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 \begin{aligned} & \boldsymbol{w}_1 &= \boldsymbol{v}_1, & W_1 = \operatorname{Span} \left\{ \boldsymbol{w}_1 \right\}, \\ & \boldsymbol{w}_2 &= \boldsymbol{v}_2 - \operatorname{Proj}_{W_1}(\boldsymbol{v}_2), & W_2 = \operatorname{Span} \left\{ \boldsymbol{w}_1, \boldsymbol{w}_2 \right\}, \\ & \boldsymbol{w}_3 &= \boldsymbol{v}_3 - \operatorname{Proj}_{W_2}(\boldsymbol{v}_3), & W_3 = \operatorname{Span} \left\{ \boldsymbol{w}_1, \boldsymbol{w}_2, \boldsymbol{w}_3 \right\}, \\ & \vdots \\ & \boldsymbol{w}_{k-1} &= \boldsymbol{v}_{k-1} - \operatorname{Proj}_{W_{k-1}}(\boldsymbol{v}_{k-1}), & W_{k-1} = \operatorname{Span} \left\{ \boldsymbol{w}_1, \dots, \boldsymbol{w}_{k-1} \right\}, \\ & \boldsymbol{w}_k &= \boldsymbol{v}_k - \operatorname{Proj}_{W_{k-1}}(\boldsymbol{v}_k). \end{aligned}
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More precisely,

$$\begin{array}{lll} w_1 & = & v_1, \\ w_2 & = & v_2 - \frac{\langle w_1, v_2 \rangle}{\langle w_1, w_1 \rangle} w_1, \\ w_3 & = & v_3 - \frac{\langle w_1, v_3 \rangle}{\langle w_1, w_1 \rangle} w_1 - \frac{\langle w_2, v_3 \rangle}{\langle w_2, w_2 \rangle} w_2, \\ & \vdots & \\ w_k & = & v_k - \frac{\langle w_1, v_k \rangle}{\langle w_1, w_1 \rangle} w_1 - \frac{\langle w_2, v_k \rangle}{\langle w_2, w_2 \rangle} w_2 - \dots - \frac{\langle w_{k-1}, v_k \rangle}{\langle w_{k-1}, w_{k-1} \rangle} w_{k-1}. \end{array}$$

The method of constructing the orthogonal vector w_1, w_2, \dots, w_k is known as the **Gram-Schmidt proces**



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Santosh Shiv Dhamone Clearly, the vector w_1, w_2, \ldots, w_k are linear combinations of v_1, v_2, \ldots, v_k . Conversely, the vectors v_1, v_2, \ldots, v_k are also linear combinations of w_1, w_2, \ldots, w_k :

Hence

$$\operatorname{Span} \{ v_1, v_2, \dots, v_k \} = \operatorname{Span} \{ w_1, w_2, \dots, w_k \}.$$

Since $\mathcal{B} = \{v_1, v_2, \dots, v_k\}$ is a basis for W, so is the set $\mathcal{B}' = \{w_1, w_2, \dots, w_k\}$.



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Example Let W be the subspace of \mathbb{R}^4 spanned by

$$oldsymbol{v}_1 = egin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}, \ oldsymbol{v}_2 = egin{bmatrix} 1 \\ 1 \\ 1 \\ 0 \end{bmatrix}, \ oldsymbol{v}_2 = egin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}.$$

Construct an orthogonal basis for W.



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Santosh Shivla Dhamone Set $w_1 = v_1$. Let $W_1 = \text{Span}\{w_1\}$. To find a vector w_2 in W that is orthogonal to W_1 , set

$$\begin{aligned} \boldsymbol{w}_2 &= \boldsymbol{v}_2 - \operatorname{Proj}_{W_1} \boldsymbol{v}_2 = \boldsymbol{v}_2 - \frac{\langle \boldsymbol{w}_1, \boldsymbol{v}_2 \rangle}{\langle \boldsymbol{w}_1, \boldsymbol{w}_1 \rangle} \boldsymbol{w}_1 \\ &= \begin{bmatrix} 1\\1\\1\\0 \end{bmatrix} - \frac{3}{4} \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix} = \frac{1}{4} \begin{bmatrix} 1\\1\\1\\-3 \end{bmatrix}. \end{aligned}$$

Let $W_2 = \operatorname{Span} \{ \boldsymbol{w}_1, \boldsymbol{w}_2 \}$. To find a vector \boldsymbol{w}_3 in W that is orthogonal to W_2 , set

$$egin{array}{lll} oldsymbol{w}_3 &=& oldsymbol{v}_3 - \operatorname{Proj}_{W_2} oldsymbol{v}_3 \ \\ &=& oldsymbol{v}_3 - rac{\langle oldsymbol{w}_1, oldsymbol{v}_3
angle}{\langle oldsymbol{w}_1, oldsymbol{w}_1
angle} oldsymbol{w}_1 - rac{\langle oldsymbol{w}_2, oldsymbol{v}_3
angle}{\langle oldsymbol{w}_2, oldsymbol{w}_2
angle} oldsymbol{w}_2 \end{array}$$



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$$= v_3 - \frac{\langle w_1, v_3 \rangle}{\langle w_1, w_1 \rangle} w_1 - \frac{\langle w_2, v_3 \rangle}{\langle w_2, w_2 \rangle} w_2$$

$$= \begin{bmatrix} 1\\1\\0\\0 \end{bmatrix} - \frac{1}{2} \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix} - \frac{\frac{2}{4}}{\frac{12}{16}} \cdot \frac{1}{4} \begin{bmatrix} 1\\1\\1\\-3 \end{bmatrix}$$

$$= \begin{bmatrix} 1/3\\1/3\\-2/3\\0 \end{bmatrix}.$$

Then the set $\{w_1, w_2, w_3\}$ is an orthogonal basis for W.



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We want to convert an arbitrary basis $\{x_1, x_2, \dots, x_n\}$ of $\mathcal V$ to an orthonormal basis $\{u_1, u_2, \dots, u_n\}$.

Idea: construct u_1, u_2, \dots, u_n successively so that $\{u_1, u_2, \dots, u_k\}$ is an ON basis for span $\{x_1, x_2, \dots, x_k\}$, $k = 1, \dots, n$.



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Construction

$$k = 1$$
:

$$u_1 = \frac{x_1}{\|x_1\|}.$$

k = 2: Consider the projection of x_2 onto u_1 , i.e.,

$$\langle u_1, x_2 \rangle u_1$$
.

Then

$$\mathbf{v}_2 = \mathbf{x}_2 - \langle \mathbf{u}_1, \mathbf{x}_2 \rangle \mathbf{u}_1$$

and

$$\mathbf{u}_2 = \frac{\mathbf{v}_2}{\|\mathbf{v}_2\|}.$$



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Santosh Shiv Dhamone In general, consider $\{u_1, \dots, u_k\}$ as a given ON basis for $span\{x_1, \dots, x_k\}$.

Use the Fourier expansion to express x_{k+1} with respect to $\{u_1, \dots, u_{k+1}\}$:

$$\begin{aligned} & \pmb{x}_{k+1} = \sum_{i=1}^{k+1} \langle \pmb{u}_i, \pmb{x}_{k+1} \rangle \pmb{u}_i \\ & \iff & \pmb{x}_{k+1} = \sum_{i=1}^{k} \langle \pmb{u}_i, \pmb{x}_{k+1} \rangle \pmb{u}_i + \langle \pmb{u}_{k+1}, \pmb{x}_{k+1} \rangle \pmb{u}_{k+1} \\ & \iff & \pmb{u}_{k+1} = \frac{\pmb{x}_{k+1} - \sum_{i=1}^{k} \langle \pmb{u}_i, \pmb{x}_{k+1} \rangle \pmb{u}_i}{\langle \pmb{u}_{k+1}, \pmb{x}_{k+1} \rangle} = \frac{\pmb{v}_{k+1}}{\langle \pmb{u}_{k+1}, \pmb{x}_{k+1} \rangle} \end{aligned}$$

This vector, however is not vet normalized.