

Lecture 21: Quick review from previous lecture

- Let V be any inner product space. The **Gram matrix** for vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ is given by

$$K = \begin{pmatrix} \langle \mathbf{v}_1, \mathbf{v}_1 \rangle & \langle \mathbf{v}_1, \mathbf{v}_2 \rangle & \dots & \langle \mathbf{v}_1, \mathbf{v}_n \rangle \\ \langle \mathbf{v}_2, \mathbf{v}_1 \rangle & \langle \mathbf{v}_2, \mathbf{v}_2 \rangle & \dots & \langle \mathbf{v}_2, \mathbf{v}_n \rangle \\ \vdots & \vdots & & \vdots \\ \langle \mathbf{v}_n, \mathbf{v}_1 \rangle & \langle \mathbf{v}_n, \mathbf{v}_2 \rangle & \dots & \langle \mathbf{v}_n, \mathbf{v}_n \rangle \end{pmatrix}$$

- Gram matrices are always **positive semidefinite**
- they are **positive definite** precisely when the vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly **independent**.

Today we will discuss

- Continue Sec. 4.1 Orthogonal(Orthonormal) bases.

- Lecture will be recorded -

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→ Gram matrix with $\langle x, y \rangle = x^T C y$, C posi

Fact 9: Let $K = A^T C A$, where A is an $m \times n$ matrix and C is an $m \times m$ positive definite matrix. Then

$$(1) \quad \ker K = \ker A$$

and, moreover, (2) $\text{rank}(K) = \text{rank} A$.

(1) Show $\ker A = \ker K$.

$$\textcircled{1} \quad \ker A \subset \ker K$$

$x \in \ker A$, $Ax = 0$. Then $Kx = A^T C Ax = 0$ yields $x \in \ker K$.

$$\textcircled{2} \quad \ker K \subset \ker A$$

$x \in \ker K$. $Kx = 0$ implies $A^T (Ax) = 0$.

$$\text{Then } x^T A^T C Ax = 0.$$

$$(Ax)^T C (Ax)$$

Since $C > 0$ (posi. defi.), one gets $Ax = 0$.

So $x \in \ker A$.

By $\textcircled{1}$ $\textcircled{2}$, we get $\ker K = \ker A$.

(2) We have learn that $A_{m \times n}$,

$$\text{rank}(A) + \dim(\ker A) = n.$$

$$\text{rank}(K) + \dim(\ker K) = n.$$

By (1), $\dim(\ker A) = \dim(\ker K)$ and

$$\text{thus, } \text{rank}(A) = \text{rank}(K). \quad \#$$

4 Orthogonality

4.1 Orthogonal and Orthonormal Bases

We've already seen that in an inner product space V , two vectors \mathbf{x} and \mathbf{y} are said to be **orthogonal** if $\langle \mathbf{x}, \mathbf{y} \rangle = 0$.

Definition: Let V be an inner product space. We say that **nonzero** vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ are **(mutually) orthogonal** if any two vectors are orthogonal, that is,

$$\langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0 \quad \text{if } i \neq j$$

If additionally $\|\mathbf{v}_i\| = 1$, we say $\mathbf{v}_1, \dots, \mathbf{v}_n$ are **orthonormal**.

Definition: Let V be an inner product space with $\dim V = n$.

- If $\mathbf{v}_1, \dots, \mathbf{v}_n$ are 1) **mutually orthogonal** vectors; 2) also a **basis**^a for V , then we say they are an **orthogonal basis**.
- If $\mathbf{v}_1, \dots, \mathbf{v}_n$ are 1) **orthonormal**; 2) a **basis** for V , then we say they are an **orthonormal basis**.

^aA **basis** for a vector space V is a linearly independent set of vectors that span V .

Example 1. (1) In \mathbb{R}^n equipped with the standard **dot product**, an orthonormal basis is the standard basis:

$$\mathbf{e}_1 = (1, 0, \dots, 0, 0)^T, \quad \mathbf{e}_2 = (0, 1, \dots, 0, 0)^T, \quad \mathbf{e}_n = (0, 0, \dots, 0, 1)^T$$

$\{\mathbf{e}_1, \dots, \mathbf{e}_n\}$ is a basis.

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

$$\|\mathbf{e}_i\| = \sqrt{\langle \mathbf{e}_i, \mathbf{e}_i \rangle} = 1, \quad 1 \leq i \leq n.$$

(2) The set

$$\left\{ \underset{\substack{\uparrow \\ \mathbf{v}_1}}{(1, 2)^T}, \underset{\substack{\uparrow \\ \mathbf{v}_2}}{(2, -1)^T} \right\}$$

is an **orthogonal** basis for \mathbb{R}^2 with the standard dot product. Use this basis to get an orthonormal basis for \mathbb{R}^2 .

$$\left\langle \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 2 \\ -1 \end{pmatrix} \right\rangle = 0.$$

An orthonormal basis $\left\{ \frac{1}{\sqrt{5}} \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \frac{1}{\sqrt{5}} \begin{pmatrix} 2 \\ -1 \end{pmatrix} \right\}$.

$$\|\mathbf{v}_1\| = \sqrt{\langle \mathbf{v}_1, \mathbf{v}_1 \rangle} = \sqrt{\left\langle \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \begin{pmatrix} 1 \\ 2 \end{pmatrix} \right\rangle} = \sqrt{5}$$

Fact 1: Suppose that $\mathbf{v}_1, \dots, \mathbf{v}_n$ is an orthogonal basis. Then

$$\frac{\mathbf{v}_1}{\|\mathbf{v}_1\|}, \dots, \frac{\mathbf{v}_n}{\|\mathbf{v}_n\|}$$

form an orthonormal basis.

Fact 2: If nonzero vectors $\mathbf{v}_1, \dots, \mathbf{v}_n$ are mutually orthogonal, then $\mathbf{v}_1, \dots, \mathbf{v}_n$ are linearly independent. $\rightarrow \langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0, i \neq j.$

[To see this] Consider $a_1 \mathbf{v}_1 + \dots + a_n \mathbf{v}_n = \mathbf{0}$. Show $a_1 = \dots = a_n = 0$.

Inner product with \mathbf{v}_i :

$$\begin{aligned} 0 = \langle \mathbf{0}, \mathbf{v}_i \rangle &= \langle a_1 \mathbf{v}_1 + \dots + a_n \mathbf{v}_n, \mathbf{v}_i \rangle, \\ &= a_1 \langle \mathbf{v}_1, \mathbf{v}_i \rangle + \dots + a_n \langle \mathbf{v}_n, \mathbf{v}_i \rangle. \end{aligned}$$

(since $\langle \mathbf{v}_i, \mathbf{v}_j \rangle = 0$ $i \neq j$)

$$= a_i \|\mathbf{v}_i\|^2$$

Since $\|\mathbf{v}_i\| \neq 0$, one gets $a_i = 0$.

Similarly, inner product with \mathbf{v}_i , we get $a_i = 0$ $2 \leq i \leq n$.

Fact 2 directly implies the following result.

Fact 3: Suppose that $\mathbf{v}_1, \dots, \mathbf{v}_n$ are nonzero, mutually orthogonal (resp. orthonormal) vectors. Then $\mathbf{v}_1, \dots, \mathbf{v}_n$ form an orthogonal (resp. orthonormal) basis for their span $W = \text{span}\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$.

Example 2. $\mathbf{v}_1 = (1, 1, 0)^T$ and $\mathbf{v}_2 = (0, 0, 1)^T$ are vectors in \mathbb{R}^3 under the standard dot product. The space $W = \text{span}\{\mathbf{v}_1, \mathbf{v}_2\}$ is the plane $x - y = 0$ in \mathbb{R}^3 . Then $\{\mathbf{v}_1, \mathbf{v}_2\}$ is an orthogonal basis for W .

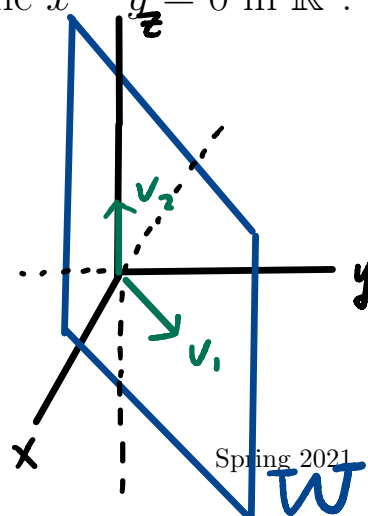
$$\langle \mathbf{v}_1, \mathbf{v}_2 \rangle = \left\langle \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \right\rangle = 0.$$

So $\mathbf{v}_1, \mathbf{v}_2$ are orthogonal.

Fact 3 \rightarrow

$W = \text{span}\{\mathbf{v}_1, \mathbf{v}_2\}$ has

an orthogonal basis $\{\mathbf{v}_1, \mathbf{v}_2\}$.



Example 3. Explain why vectors $\mathbf{x} = (1, 1, 0)^T$, $\mathbf{y} = (1, -1, 1)^T$, and $\mathbf{z} = (1, -1, -2)^T$ form an orthogonal basis in \mathbb{R}^3 under the standard dot product? Turn them into an orthonormal basis.

$$(1) \quad \langle \mathbf{x}, \mathbf{y} \rangle = \left\langle \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix} \right\rangle = 0$$

$$\langle \mathbf{x}, \mathbf{z} \rangle = \left\langle \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \\ -2 \end{pmatrix} \right\rangle = 0$$

$$\langle \mathbf{y}, \mathbf{z} \rangle = \left\langle \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \\ -2 \end{pmatrix} \right\rangle = 1 + 1 - 2 = 0.$$

$$\begin{aligned} \|\mathbf{x}\|^2 &= \langle \mathbf{x}, \mathbf{x} \rangle \\ &= 2 \\ \|\mathbf{x}\| &= \sqrt{2} \end{aligned}$$

So $\{\mathbf{x}, \mathbf{y}, \mathbf{z}\}$ are orthogonal. By Fact 2, $\{\mathbf{x}, \mathbf{y}, \mathbf{z}\}$ are l. indep. Combining $\dim(\mathbb{R}^3) = 3$, $\{\mathbf{x}, \mathbf{y}, \mathbf{z}\}$ is an orthogonal basis ~~##~~.

(2) To get orthonormal basis.

$$\left\{ \frac{\mathbf{x}}{\|\mathbf{x}\|}, \frac{\mathbf{y}}{\|\mathbf{y}\|}, \frac{\mathbf{z}}{\|\mathbf{z}\|} \right\} = \left\{ \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \frac{1}{\sqrt{3}} \begin{pmatrix} 1 \\ -1 \\ 1 \end{pmatrix}, \frac{1}{\sqrt{6}} \begin{pmatrix} 1 \\ -1 \\ -2 \end{pmatrix} \right\} \quad \#$$

§ Computations in Orthogonal Bases

Q: What are the advantages of orthogonal (orthonormal) bases?

It is simple to find the *coordinates of a vector* in the orthogonal (orthonormal) basis. However, in general this is not so easy if it is not in such basis.

Fact 4: If $\mathbf{v}_1, \dots, \mathbf{v}_n$ is an orthogonal basis in any inner product space V , then for any vector $\mathbf{v} \in V$, we have

$$\mathbf{v} = \underbrace{\frac{\langle \mathbf{v}, \mathbf{v}_1 \rangle}{\|\mathbf{v}_1\|^2}}_{a_1} \mathbf{v}_1 + \dots + \underbrace{\frac{\langle \mathbf{v}, \mathbf{v}_n \rangle}{\|\mathbf{v}_n\|^2}}_{a_n} \mathbf{v}_n.$$

Moreover, we have

$$\|\mathbf{v}\|^2 = \sum_{i=1}^n \left(\frac{\langle \mathbf{v}, \mathbf{v}_i \rangle}{\|\mathbf{v}_i\|} \right)^2.$$

Let $a_i = \frac{\langle \mathbf{v}, \mathbf{v}_i \rangle}{\|\mathbf{v}_i\|^2}$ for $i = 1, \dots, n$. We call $(a_1, \dots, a_n)^T$ the coordinates of \mathbf{v} in the basis $\{\mathbf{v}_1, \dots, \mathbf{v}_n\}$.

[To see this]

continue next time

Fact 5: If $\mathbf{v}_1, \dots, \mathbf{v}_n$ is an **orthonormal basis** in any inner product space V , then for any vector $\mathbf{v} \in V$, we have $\|\mathbf{v}_i\| = 1$.

$$\|\mathbf{v}_i\|=1 \quad \text{Fact 4} \Rightarrow \mathbf{v} = \langle \mathbf{v}, \mathbf{v}_1 \rangle \mathbf{v}_1 + \dots + \langle \mathbf{v}, \mathbf{v}_n \rangle \mathbf{v}_n.$$

Moreover, we have

$$\|\mathbf{v}\|^2 = \sum_{i=1}^n (\langle \mathbf{v}, \mathbf{v}_i \rangle)^2.$$

Example 4. Consider the same orthogonal basis as in Example 3:

$$\mathbf{x} = (1, 1, 0)^T, \quad \mathbf{y} = (1, -1, 1)^T, \quad \mathbf{z} = (1, -1, -2)^T.$$

Write $\mathbf{v} = (1, 0, 3)^T$ as the linear combination of \mathbf{x} , \mathbf{y} and \mathbf{z} in \mathbb{R}^3 under the **standard dot product**. (that is, find the coordinates of \mathbf{v} in this orthogonal basis.)

$$\mathbf{v} = a_1 \mathbf{x} + a_2 \mathbf{y} + a_3 \mathbf{z}.$$
$$a_1 = \frac{\langle \mathbf{v}, \mathbf{x} \rangle}{\|\mathbf{x}\|^2} = \frac{\langle \begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \rangle}{\langle \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 1 \\ 1 \\ 0 \end{pmatrix} \rangle} = \frac{1}{2}$$

$$a_2 = \frac{\langle v, y \rangle}{\|y\|^2} = \frac{\langle \begin{pmatrix} 1 \\ 0 \\ 3 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \rangle}{\langle \begin{pmatrix} 1 \\ -1 \end{pmatrix}, \begin{pmatrix} 1 \\ -1 \end{pmatrix} \rangle} = \frac{4}{3}$$

$$a_3 = \frac{\langle v, z \rangle}{\|z\|^2} = \frac{-5}{6}$$

$$\text{Then } v = \frac{1}{2}x + \frac{4}{3}y - \frac{5}{6}z$$

Example 5. We consider the inner product space $\mathcal{P}^{(2)}([0, 1])$ (the set of polynomials of degree ≤ 2) equipped with the L^2 inner product $\langle p, q \rangle = \int_0^1 p(x)q(x)dx$ in the following problems.

(1) The basis $1, x, x^2$ do NOT form an orthogonal basis in $\mathcal{P}^{(2)}([0, 1])$.

$$\langle 1, x \rangle = \int_0^1 1 \cdot x \, dx = \frac{1}{2}x^2 \Big|_0^1 = \frac{1}{2} \neq 0$$

(2) $\{p_1(x) = 1, p_2(x) = x - \frac{1}{2}, p_3(x) = x^2 - x + \frac{1}{6}\}$ is an orthogonal basis of $\mathcal{P}^{(2)}([0, 1])$.

$$\langle p_1, p_2 \rangle = \int_0^1 1 \cdot (x - \frac{1}{2}) \, dx = \frac{1}{2}x^2 - \frac{1}{2}x \Big|_0^1 = 0$$

$$\langle p_1, p_3 \rangle = \int_0^1 1 \cdot (x^2 - x + \frac{1}{6}) \, dx = \frac{1}{3}x^3 - \frac{1}{2}x^2 + \frac{1}{6}x \Big|_0^1 = 0$$

$$\langle p_2, p_3 \rangle = \sim = 0$$

Since $\{p_1, p_2, p_3\}$ are orthogonal, Fact 2 implies

$\{p_1, p_2, p_3\}$ are l. indep. Together with

$\dim(\mathcal{P}^{(2)}([0, 1])) = 3$, $\{p_1, p_2, p_3\}$ is an orthogonal basis for $\mathcal{P}^{(2)}([0, 1])$

Poll Question 1: Are $\{(1, 0, 1)^T, (1, 0, -1)^T\}$ orthogonal with respect to the **standard dot product**

- A) Yes
- B) No

Poll Question 2: Are $\{(1, 0, 1)^T, (1, 0, -1)^T\}$ orthogonal with respect to the inner product $\langle \mathbf{x}, \mathbf{y} \rangle = \mathbf{x}^T \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix} \mathbf{y}$

$$\mathbf{x}^T \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{pmatrix} \mathbf{y}$$

- A) Yes
- B) No

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