

§4.4 标准正交基和 Gram-Schmidt 正交化

回顾: $A \in M_{m \times n}(\mathbb{R})$, $A = [\vec{\alpha}_1 \vec{\alpha}_2 \cdots \vec{\alpha}_n]$ 列满秩 ($n \leq m$). $\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n$ 为 $C(A)$ 一维基.

$\forall \vec{b} \in \mathbb{R}^m$, 则 $A\vec{x} = \vec{b}$ 的最小二乘解为: $\vec{x} = (A^T A)^{-1} A^T \vec{b}$, (满足 $(A^T A)\vec{x} = A^T \vec{b}$)

\vec{b} 在 $C(A)$ 上的投影为 $\vec{p} = A\vec{x} = A(A^T A)^{-1} A^T \vec{b}$.

投影矩阵 $P = A(A^T A)^{-1} A^T \in M_{m \times m}(\mathbb{R})$

当 $\vec{\alpha}_i, i=1, 2, \dots, n$ 两两正交时, $A^T A$ 为对角阵, 易于计算.

1. 标准正交向量组

定理 设 $\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n \in \mathbb{R}^m$ 为两两正交的非零向量, 则 $\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n$ 线性无关.

pf. 设 $x_1 \vec{\alpha}_1 + x_2 \vec{\alpha}_2 + \cdots + x_n \vec{\alpha}_n = \vec{0}$

两边分别点乘 $\vec{\alpha}_i$, 得: $x_1(\vec{\alpha}_1 \cdot \vec{\alpha}_i) + x_2(\vec{\alpha}_2 \cdot \vec{\alpha}_i) + \cdots + x_n(\vec{\alpha}_n \cdot \vec{\alpha}_i) = \vec{0}$

两两正交: $\vec{\alpha}_i \cdot \vec{\alpha}_j = 0$. # 由于

$$\Rightarrow x_i \|\vec{\alpha}_i\|^2 = 0 \quad \text{又 } \vec{\alpha}_i \neq \vec{0}. \quad \therefore \| \vec{\alpha}_i \|^2 > 0$$

$$\Rightarrow x_i = 0, i=1, 2, \dots, n$$

$\therefore \vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n$ 线性无关. #

推论: \mathbb{R}^n 中两两正交的非零向量个数不超过 m .

定义. 若 $\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n$ 两两正交, 那 $\vec{\alpha}_i \cdot \vec{\alpha}_j = 0$. # 称向量组 $\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n$ 为正交 (orthogonal) 向量组.

进一步地, 若 $\|\vec{\alpha}_i\|=1, i=1, 2, \dots, n$, 称 $\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n$ 为标准正交 (orthonormal) 向量组.

定理 设 $Q = [\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n] \in M_{m \times n}(\mathbb{R})$, $n \leq m$. 若 $\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n$ 为 \mathbb{R}^m 中的标准正交向量组, 则 $Q^T Q = I_n$.

进一步地, 若 $m=n$, 则 Q 为正交阵, 那 $Q Q^T = Q^T Q = I_n$, 而 $Q^T = Q^{-1}$

pf. $(Q^T Q)_{ij} = \vec{q}_i^T \vec{q}_j = \vec{q}_i \cdot \vec{q}_j = \delta_{ij} \Rightarrow Q^T Q = I_n$. #

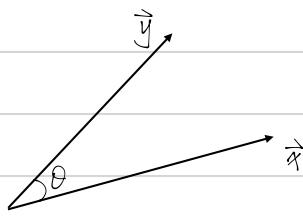
定理. $Q \in M_{m,n}(\mathbb{R})$ 定义了一个从 \mathbb{R}^n 到 \mathbb{R}^m 的线性映射 ($\vec{x} \in \mathbb{R}^n \rightarrow Q\vec{x} \in \mathbb{R}^m$). 若 $Q^T Q = I_n$, 则该线性映射 Q 保持长度.

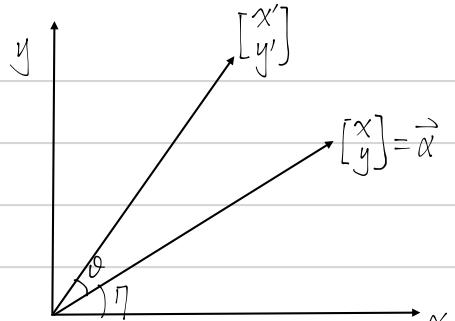
即 $(Q\vec{x}) \cdot (Q\vec{y}) = \vec{x} \cdot \vec{y} \quad \forall \vec{x}, \vec{y} \in \mathbb{R}^n$.

pf. $(Q\vec{x}) \cdot (Q\vec{y}) = (Q\vec{x})^T Q\vec{y} = \vec{x}^T Q^T Q\vec{y} = \vec{x}^T \vec{y} = \vec{x} \cdot \vec{y}$. #

推论. (1). Q 保持长度: $\|Q\vec{x}\|^2 = (Q\vec{x}) \cdot (Q\vec{x}) = \vec{x}^T Q^T Q \vec{x} = \vec{x}^T \vec{x} = \|\vec{x}\|^2 \quad \forall \vec{x} \in \mathbb{R}^n$.

(2) Q 保持角度: $\cos \theta = \frac{(Q\vec{x}) \cdot (Q\vec{y})}{\|Q\vec{x}\| \|Q\vec{y}\|} = \frac{\vec{x} \cdot \vec{y}}{\|\vec{x}\| \|\vec{y}\|}$





例. \mathbb{R}^2 中的旋转变换. ($m=n=2$)

将 $[\vec{x}, \vec{y}] \in \mathbb{R}^2$ 绕原点逆时针旋转角度角, 得 $[\vec{x}', \vec{y}']$.

$$\begin{cases} x' = |\vec{\alpha}| \cos(\theta + \eta) = |\vec{\alpha}| (\cos\theta \cos\eta - |\vec{\alpha}| \sin\theta \sin\eta) = x \cos\theta - y \sin\theta \\ y' = |\vec{\alpha}| \sin(\theta + \eta) = |\vec{\alpha}| (\sin\theta \cos\eta + |\vec{\alpha}| \cos\theta \sin\eta) = x \sin\theta + y \cos\theta \end{cases}$$

$$\Rightarrow \begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}. \quad Q = \begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} := [\vec{q}_1, \vec{q}_2]$$

$$\begin{aligned} \text{(1)} \vec{q}_1 \cdot \vec{q}_2 &= 0 \Rightarrow \vec{q}_1 \perp \vec{q}_2 \text{ 到向量正交} & \Rightarrow Q^T Q = I_2. \text{ 又 } Q \text{ 是方阵} \Rightarrow Q \text{ 是正交矩阵. } Q^{-1} = Q^T. \\ \text{(2)} \|\vec{q}_1\|^2 &= \|\vec{q}_2\|^2 = \sin^2\theta + \cos^2\theta = 1 \text{ 高度向量} \end{aligned}$$

\vec{q}_1, \vec{q}_2 构成 \mathbb{R}^2 中一组标准正交基, $\vec{q}_1 = Q \vec{e}_1$, $\vec{q}_2 = Q \vec{e}_2$. 分别由 \vec{e}_1, \vec{e}_2 转换而得. Fig 4.10

$$Q^{-1} = Q^T = \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} = \begin{bmatrix} \cos(-\theta) & -\sin(-\theta) \\ \sin(-\theta) & \cos(-\theta) \end{bmatrix}, \text{ 即逆时针旋转 } -\theta \text{ 角, 还原为 } \begin{bmatrix} x \\ y \end{bmatrix}.$$

例. \mathbb{R}^2 中的镜面反射 (reflection)

\vec{u} 为正交于 mirror 的单位向量. $\|\vec{u}\|=1$

$\forall \vec{\alpha} \in \mathbb{R}^2$, 其关于 mirror 的镜面反射: $Q\vec{\alpha} \in \mathbb{R}^2$

$\vec{\alpha}$ 向 \vec{u} 所在直线的投影: $P\vec{\alpha} = \frac{\vec{u}\vec{u}^T}{\vec{u}^T\vec{u}}\vec{\alpha} = (\vec{u}\vec{u}^T)\vec{\alpha}$

$Q\vec{\alpha} = \vec{\alpha} - 2P\vec{\alpha} = (\mathbb{I} - 2P)\vec{\alpha} = (\mathbb{I} - 2\vec{u}\vec{u}^T)\vec{\alpha}$

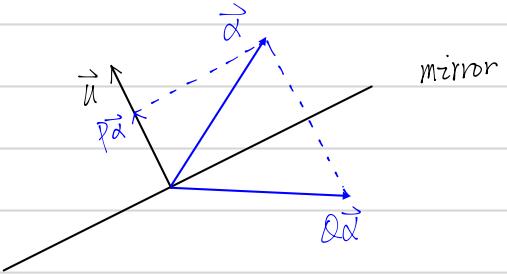
$\Rightarrow Q = \mathbb{I}_2 - 2\vec{u}\vec{u}^T = Q^T \text{ symmetric.}$

$$\|\vec{u}\|^2 = 1$$

$$Q^T Q = (\mathbb{I}_2 - 2\vec{u}\vec{u}^T)(\mathbb{I}_2 - 2\vec{u}\vec{u}^T) = \mathbb{I}_2 - 4\vec{u}\vec{u}^T + 4\vec{u}\vec{u}^T\vec{u}\vec{u}^T = \mathbb{I}_2.$$

$\Rightarrow Q$ 为正交矩阵. $Q^T Q = Q Q^T = \mathbb{I}_2 \Rightarrow Q^T = Q^{-1}$

Moreover, $Q^2 = Q^T Q = \mathbb{I}$. reflect twice through a mirror brings back the original.



2. Projections Using Orthonormal Bases.

W 为 \mathbb{R}^m 的 n 维 ($n \leq m$) 子空间, 有 W 中一组标准正交基 $\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n \in \mathbb{R}^m$

$\forall W = C(Q)$, where $Q = [\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n] \in M_{m \times n}(\mathbb{R})$.

$\Rightarrow Q^T Q = I_n$

$\forall \vec{b} \in \mathbb{R}^m$. $\exists Q\vec{x} = \vec{b}$ 的最小二乘解满足 normal eqn. $Q^T Q \vec{x} = Q^T \vec{b} \Rightarrow \vec{x} = Q^T \vec{b}$

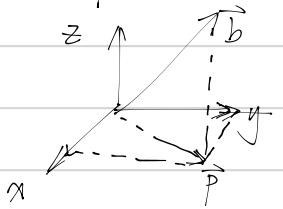
\vec{b} 在 $C(Q)$ 上的投影为 $\vec{p} = Q\vec{x} = Q Q^T \vec{b}$ ($m \neq n$ 时. $Q Q^T \neq I_m$!)

$$\text{投影矩阵 } P = Q(Q^T Q)^{-1} Q^T = Q Q^T$$

$$\vec{x} = Q^T \vec{b} \Rightarrow \begin{bmatrix} \hat{x}_1 \\ \hat{x}_2 \\ \vdots \\ \hat{x}_n \end{bmatrix} = \begin{bmatrix} \vec{q}_1^T \\ \vec{q}_2^T \\ \vdots \\ \vec{q}_n^T \end{bmatrix} \vec{b} \Rightarrow \hat{x}_1 = \vec{q}_1^T \vec{b}, \hat{x}_2 = \vec{q}_2^T \vec{b}, \dots, \hat{x}_n = \vec{q}_n^T \vec{b} \text{ decoupled.}$$

$$\vec{P} = Q \vec{x} = \hat{x}_1 \vec{q}_1 + \hat{x}_2 \vec{q}_2 + \dots + \hat{x}_n \vec{q}_n = \vec{q}_1 \vec{q}_1^T \vec{b} + \vec{q}_2 \vec{q}_2^T \vec{b} + \dots + \vec{q}_n \vec{q}_n^T \vec{b} \quad (*)$$

即将 \vec{b} 分割到一维子空间 $\text{span}\{\vec{q}_1\}$, $\text{span}\{\vec{q}_2\}$, ..., $\text{span}\{\vec{q}_n\}$ 上投影, 再相加!



特别地, 当 $m=n$ 时, Q 为正交矩阵. $W=C(Q)=\mathbb{R}^m$. — the subspace is the whole space. $Q^T=Q^{-1}$.

① $\vec{x} = Q^T \vec{b} = Q^{-1} \vec{b}$. — the solution is exact.

② $\vec{P} = Q \vec{x} = Q Q^T \vec{b} = \vec{b}$ — the projection of \vec{b} onto the whole space is \vec{b} itself.

③ $\vec{P} = Q Q^T = I_n$ 恒等变换.

$$(*) \Rightarrow \vec{b} = (\vec{q}_1^T \vec{b}) \vec{q}_1 + (\vec{q}_2^T \vec{b}) \vec{q}_2 + \dots + (\vec{q}_n^T \vec{b}) \vec{q}_n \text{ perpendicular pieces.}$$

\vec{b} 展开为 \mathbb{R}^m 中标准正交基的组合, 组合系数 $x_i = \vec{b} \cdot \vec{q}_i$.

也可如下推导:

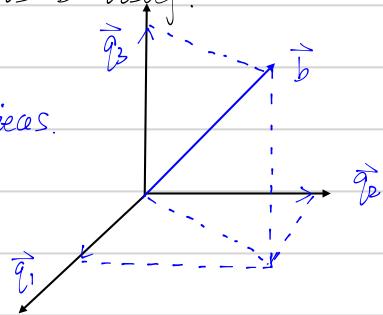
设 $\vec{b} = x_1 \vec{q}_1 + x_2 \vec{q}_2 + \dots + x_n \vec{q}_n$. \vec{x} 为 \vec{b} 的坐标.

$$\Rightarrow \vec{b} \cdot \vec{q}_1 = x_1 \|\vec{q}_1\|^2 \Rightarrow x_1 = \vec{b} \cdot \vec{q}_1$$

$$\vec{b} \cdot \vec{q}_2 = x_2 \|\vec{q}_2\|^2 \Rightarrow x_2 = \vec{b} \cdot \vec{q}_2$$

:

$$\vec{b} \cdot \vec{q}_n = x_n \|\vec{q}_n\|^2 \Rightarrow x_n = \vec{b} \cdot \vec{q}_n$$



3. Gram-Schmidt 过程

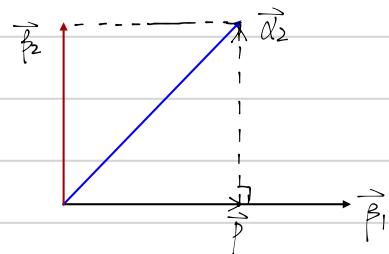
n 个线性无关的向量 $\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n \in \mathbb{R}^m$ ($n \leq m$) \rightarrow 标准正交向量组 $\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n \in \mathbb{R}^m$

Step 1 正交化.

1° 取 $\vec{\beta}_1 = \vec{\alpha}_1 \neq 0$

2° 从 $\vec{\alpha}_2$ 中减去其在 $\vec{\beta}_1$ 上的投影部分, 得 $\vec{\beta}_2$

$$\vec{\beta}_2 = \vec{\alpha}_2 - \frac{\vec{\beta}_1^T \vec{\alpha}_2}{\vec{\beta}_1^T \vec{\beta}_1} \vec{\beta}_1, \Rightarrow \vec{\beta}_2 \perp \vec{\beta}_1$$



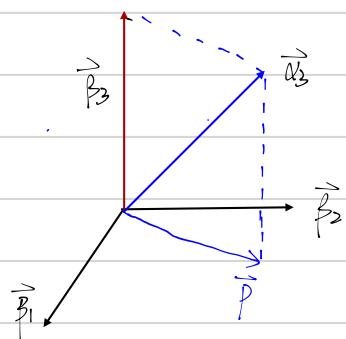
这里 $\vec{\beta}_2 \neq 0$, 否则 $\vec{\alpha}_1$ 与 $\vec{\alpha}_2$ 线性相关.

3° 从 $\vec{\alpha}_3$ 中减去其在 $\text{span}\{\vec{\beta}_1, \vec{\beta}_2\}$ 上的投影, 得 $\vec{\beta}_3$

$$\Rightarrow \vec{\beta}_3 \perp \text{span}\{\vec{\beta}_1, \vec{\beta}_2\} \Rightarrow \vec{\beta}_3 \perp \vec{\beta}_1 \text{ 且 } \vec{\beta}_3 \perp \vec{\beta}_2$$

又有 $\vec{\beta}_1 \perp \vec{\beta}_2$. $\therefore \vec{\alpha}_3$ 在 $\text{span}\{\vec{\beta}_1, \vec{\beta}_2\}$ 上的投影即为 $\vec{\alpha}_3$ 分别

在 $\vec{\beta}_1$ 及 $\vec{\beta}_2$ 上的投影之和.



$$\vec{\beta}_3 = \vec{\alpha}_3 - \frac{\vec{\beta}_1^T \vec{\alpha}_3}{\vec{\beta}_1^T \vec{\beta}_1} \vec{\beta}_1 - \frac{\vec{\beta}_2^T \vec{\alpha}_3}{\vec{\beta}_2^T \vec{\beta}_2} \vec{\beta}_2 \neq \vec{0}, \text{否则 } \vec{\alpha}_1, \vec{\alpha}_2, \vec{\alpha}_3 \text{ 线性相关.}$$

⋮

$$\text{一般地, } \vec{\beta}_k = \vec{\alpha}_k - \frac{\vec{\beta}_1^T \vec{\alpha}_k}{\vec{\beta}_1^T \vec{\beta}_1} \vec{\beta}_1 - \dots - \frac{\vec{\beta}_{k-1}^T \vec{\alpha}_k}{\vec{\beta}_{k-1}^T \vec{\beta}_{k-1}} \vec{\beta}_{k-1}$$

即从 $\vec{\alpha}_k$ 中减去其在 $\vec{\beta}_1, \vec{\beta}_2, \dots, \vec{\beta}_{k-1}$ 方向上的投影.

可得 $\vec{\beta}_1, \vec{\beta}_2, \dots, \vec{\beta}_n$ 正交

Step 2. 单位化

$$\vec{q}_i = \vec{\beta}_i / \|\vec{\beta}_i\|, \quad i=1, 2, \dots, n.$$