

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

⋮

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

即从 $\vec{\alpha}_k$ 中分别减去其在 $\vec{\beta}_1, \vec{\beta}_2, \dots, \vec{\beta}_{k-1}$ 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

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

4. QR分解

定理. 设 $A \in M_{m,n}(\mathbb{R})$, $\text{rank}(A)=n$, 则 A 可分解为 $A=QR$, 其中 $Q \in M_{m,n}$ 满足 $Q^T Q = I_n$ (Q 的列向量组标准正交), $R \in M_n$ 为主对角线元素为正的上三角阵, $A=QR$ 称为 A 的 QR 分解.

pf. 设 $A = [\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n]$.

$$\text{rank}(A)=n \Rightarrow \vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n \text{ 线性无关}$$

可由 Gram-Schmidt 过程化为标准正交向量组 $\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n$.

$$\text{Step 1. } \vec{\beta}_k = \vec{\alpha}_k - \frac{\vec{\beta}_1^T \vec{\alpha}_k}{\vec{\beta}_1^T \vec{\beta}_1} \vec{\beta}_1 - \dots - \frac{\vec{\beta}_{k-1}^T \vec{\alpha}_k}{\vec{\beta}_{k-1}^T \vec{\beta}_{k-1}} \vec{\beta}_{k-1}$$

$$\Rightarrow \vec{\alpha}_j = \sum_{i=1}^{j-1} \frac{\vec{\beta}_i^T \vec{\alpha}_j}{\vec{\beta}_i^T \vec{\beta}_i} \vec{\beta}_i + \vec{\beta}_j$$

\uparrow 数 \uparrow j 以前的向量

$$\Rightarrow [\vec{\alpha}_1, \vec{\alpha}_2, \dots, \vec{\alpha}_n] = [\vec{\beta}_1, \vec{\beta}_2, \dots, \vec{\beta}_n] \begin{bmatrix} 1 & & * \\ & 1 & \\ & & \ddots \\ & & & 1 \end{bmatrix}$$

$$\text{Step 2. } \vec{q}_i = \frac{1}{\|\vec{\beta}_i\|} \vec{\beta}_i \Rightarrow \vec{\beta}_i = \|\vec{\beta}_i\| \vec{q}_i, \quad i=1, 2, \dots, n$$

$$\Rightarrow [\vec{\beta}_1, \vec{\beta}_2, \dots, \vec{\beta}_n] = [\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n] \begin{bmatrix} \|\vec{\beta}_1\| & & \\ & \|\vec{\beta}_2\| & \\ & & \ddots \\ & & & \|\vec{\beta}_n\| \end{bmatrix}$$

综合 Steps 1, 2, 有:

$$A = [\vec{q}_1, \vec{q}_2, \dots, \vec{q}_n] \begin{bmatrix} \|\vec{\beta}_1\| & & \\ & \|\vec{\beta}_2\| & \\ & & \ddots \\ & & & \|\vec{\beta}_n\| \end{bmatrix} \begin{bmatrix} 1 & & * \\ & 1 & \\ & & \ddots \\ & & & 1 \end{bmatrix} = \underbrace{[\vec{q}_1, \dots, \vec{q}_n]}_Q \underbrace{\begin{bmatrix} \|\vec{\beta}_1\| & & * & * \\ & \|\vec{\beta}_2\| & & \\ & & \ddots & \\ & & & \|\vec{\beta}_n\| \end{bmatrix}}_R \neq$$

例. $A = \begin{bmatrix} 1 & 2 \\ 1 & 3 \end{bmatrix}$. 作 A 的 QR 分解.

解. 令 $\vec{\alpha}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$, $\vec{\alpha}_2 = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$. 施密特正交化:

$$\vec{\beta}_1 = \vec{\alpha}_1 = \begin{bmatrix} 1 \\ 1 \end{bmatrix} \Rightarrow \vec{\alpha}_1 = \vec{\beta}_1$$

$$\vec{\beta}_2 = \vec{\alpha}_2 - \frac{\vec{\beta}_1^T \vec{\alpha}_2}{\vec{\beta}_1^T \vec{\beta}_1} \vec{\beta}_1 = \vec{\alpha}_2 - \frac{5}{2} \vec{\beta}_1 = \frac{1}{2} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \Rightarrow \vec{\alpha}_2 = \frac{5}{2} \vec{\beta}_1 + \vec{\beta}_2$$

$$\text{单位化: } \vec{q}_1 = \frac{1}{\|\vec{\beta}_1\|} \vec{\beta}_1 = \frac{1}{\sqrt{2}} \vec{\beta}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \Rightarrow \vec{\beta}_1 = \sqrt{2} \vec{q}_1$$

$$\vec{q}_2 = \frac{1}{\|\vec{\beta}_2\|} \vec{\beta}_2 = \frac{1}{\frac{\sqrt{2}}{2}} \vec{\beta}_2 = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \Rightarrow \vec{\beta}_2 = \frac{\sqrt{2}}{2} \vec{q}_2$$

$$\Rightarrow [\vec{\alpha}_1 \ \vec{\alpha}_2] = [\vec{\beta}_1 \ \vec{\beta}_2] \begin{bmatrix} 1 & \frac{5}{2} \\ 0 & 1 \end{bmatrix} = [\vec{q}_1 \ \vec{q}_2] \underbrace{\begin{bmatrix} \sqrt{2} & \\ & \frac{\sqrt{2}}{2} \end{bmatrix}}_Q \begin{bmatrix} 1 & \frac{5}{2} \\ 0 & 1 \end{bmatrix} = \underbrace{\begin{bmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix}}_Q \underbrace{\begin{bmatrix} \sqrt{2} & \frac{5\sqrt{2}}{2} \\ 0 & \frac{\sqrt{2}}{2} \end{bmatrix}}_R$$

Remarks

$$A = QR, \quad Q^T Q = I, \quad R \text{ 可逆上三角}$$

$$(1) \quad Q^T A = \underbrace{Q^T Q}_I R = R \Rightarrow (R)_{ij} = \vec{q}_i \cdot \vec{\alpha}_j = 0 \text{ when } j < i \quad (\vec{\alpha}_j \perp \vec{q}_i) \Rightarrow R \text{ 为上三角}$$

$$(2) \text{ normal equation: } A^T A \vec{x} = A^T \vec{b}$$

$$A^T A = R^T \underbrace{Q^T Q}_I R = R^T R$$

$$A^T A \vec{x} = A^T \vec{b} \Rightarrow R^T R \vec{x} = R^T Q^T \vec{b} \xrightarrow{R^T \text{ 可逆}} R \vec{x} = Q^T \vec{b}, \quad R \text{ 为上三角, 易求解.}$$

$$\Rightarrow \vec{x} = R^{-1} Q^T \vec{b}. \quad \text{back substitution. fast.}$$

Chapter 5. 行列式

n 阶行列式:

\mathbb{R}^n 上 n 个向量所构成的平行 $2n$ 面体的有向体积.

§5.1 几何向量的数量积, 向量积与混合积

1. 数量积

定义: 两个几何向量 $\vec{\alpha}$ 与 $\vec{\beta}$ 的数量积/点积/内积 (记作 $\vec{\alpha} \cdot \vec{\beta}$) 为一个数:

$$\vec{\alpha} \cdot \vec{\beta} := \|\vec{\alpha}\| \|\vec{\beta}\| \cos \langle \vec{\alpha}, \vec{\beta} \rangle$$

其中 $\langle \vec{\alpha}, \vec{\beta} \rangle$ 为 $\vec{\alpha}$ 与 $\vec{\beta}$ 的夹角 (不大于 π). 若 $\vec{\alpha}$ 或 $\vec{\beta}$ 为 $\vec{0}$, 规定 $\vec{\alpha} \cdot \vec{\beta} = 0$.

Remark: $\|\vec{\alpha}\|^2 = \vec{\alpha} \cdot \vec{\alpha}$ 长度
 $\cos \langle \vec{\alpha}, \vec{\beta} \rangle = \frac{\vec{\alpha} \cdot \vec{\beta}}{\|\vec{\alpha}\| \|\vec{\beta}\|}$ 夹角

性质: (1) $\vec{\alpha} \cdot \vec{\beta} = \vec{\beta} \cdot \vec{\alpha}$ 对称性
 (2) $(k\vec{\alpha}) \cdot \vec{\beta} = k(\vec{\alpha} \cdot \vec{\beta})$ 线性
 (3) $(\vec{\alpha} + \vec{\beta}) \cdot \vec{\gamma} = \vec{\alpha} \cdot \vec{\gamma} + \vec{\beta} \cdot \vec{\gamma}$
 (4) $\vec{\alpha} \cdot \vec{\alpha} \geq 0$, 等号成立 $\Leftrightarrow \vec{\alpha} = \vec{0}$ 正定性

\Rightarrow 对第 2 个向量也是线性的

定理: $\vec{\alpha} \perp \vec{\beta} \Leftrightarrow \vec{\alpha} \cdot \vec{\beta} = 0$

用坐标计算数量积

定理: $\sum_{i=1}^n \sum_{j=1}^n b_{ij} x_i y_j = [x_1 \ x_2 \ \dots \ x_n] \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{m1} & b_{m2} & \dots & b_{mn} \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = \vec{x}^T B \vec{y}$

定义 矩阵 $A = \begin{bmatrix} \vec{e}_1 \cdot \vec{e}_1 & \vec{e}_1 \cdot \vec{e}_2 & \vec{e}_1 \cdot \vec{e}_3 \\ \vec{e}_2 \cdot \vec{e}_1 & \vec{e}_2 \cdot \vec{e}_2 & \vec{e}_2 \cdot \vec{e}_3 \\ \vec{e}_3 \cdot \vec{e}_1 & \vec{e}_3 \cdot \vec{e}_2 & \vec{e}_3 \cdot \vec{e}_3 \end{bmatrix}$ 称为仿射坐标系 $\{O; \vec{e}_1, \vec{e}_2, \vec{e}_3\}$ 的度量矩阵.

定理 给定仿射坐标系 $\{O; \vec{e}_1, \vec{e}_2, \vec{e}_3\}$, A 为度量矩阵, 向量 $\vec{\alpha} = x_1 \vec{e}_1 + x_2 \vec{e}_2 + x_3 \vec{e}_3$, $\vec{\beta} = y_1 \vec{e}_1 + y_2 \vec{e}_2 + y_3 \vec{e}_3$,
 则内积 $\vec{\alpha} \cdot \vec{\beta} = \vec{x}^T A \vec{y}$, 其中 $\vec{x}^T = [x_1 \ x_2 \ x_3]$, $\vec{y} = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}$