

1. 给出 Thm 5.7 的详细证明.

PF: 只需证  $\langle \cdot, \cdot \rangle$  在  $C[a, b]$  上的对称性, 正定性

(1) 正定性.

$$\langle u, u \rangle = \int_a^b \rho(t) u(t) \overline{u(t)} dt = \int_a^b \rho(t) |u(t)|^2 dt \geq 0.$$

(2) 绝对性.

$$\text{若 } \langle u, u \rangle = \int_a^b \rho(t) |u(t)|^2 dt = 0, \text{ 则因为 } \rho(t) > 0, \text{ 所以 } u(t) \equiv 0.$$

(3) 线性性.

$$\langle au + v, w \rangle = \int_a^b \rho(t) (au(t) + v(t)) \overline{w(t)} dt = a \int_a^b \rho(t) u(t) \overline{w(t)} dt + \int_a^b \rho(t) v(t) \overline{w(t)} dt = a \langle u, w \rangle + \langle v, w \rangle.$$

(4) 对称性.

$$\langle v, u \rangle = \int_a^b \rho(t) v(t) \overline{u(t)} dt = \int_a^b \overline{\rho(t) u(t) \overline{v(t)}} dt = \overline{\langle u, v \rangle}.$$

所以  $\langle \cdot, \cdot \rangle$  是内积运算,  $C[a, b]$  是内积空间.

2. 考虑第一类切比雪夫多项式.

(1) 证明它们在  $C[-1, 1]$  上正交. 内积如 Thm 5.7 定义,  $\rho(x) = \frac{1}{\sqrt{1-x^2}}$ .

$$\begin{aligned} \text{PF: } \langle T_n, T_m \rangle &= \int_{-1}^1 \rho(t) T_n(t) T_m(t) dt = \int_{-1}^1 \frac{1}{\sqrt{1-t^2}} \cos(n \arccos t) \cos(m \arccos t) dt \\ &= \int_{-1}^1 \cos(n \arccos t) \cos(m \arccos t) d \arccos t \\ &= \int_0^\pi \cos(n\theta) \cos(m\theta) d\theta = \int_0^\pi \frac{\cos[(n-m)\theta] + \cos[(n+m)\theta]}{2} d\theta \\ &= \left[ \frac{\sin[(n-m)\theta]}{2(n-m)} + \frac{\sin[(n+m)\theta]}{2(n+m)} \right]_0^\pi = \begin{cases} 0, & n \neq m \\ \frac{\pi}{2}, & n = m \end{cases} \\ &= \left[ \left( \frac{\theta}{2} + \frac{\sin 2n\theta}{4n} \right) \right]_0^\pi = \begin{cases} 0, & n \neq m \\ \frac{\pi}{2}, & n = m \end{cases} \end{aligned}$$

$\therefore \{T_n\}$  正交.

(2) 将前三个第一类切比雪夫多项式正交化, 使其成为标准正交组.

$$\text{Sol: } \langle T_m, T_n \rangle = \frac{\pi}{2}, \text{ 所以 } \|T_n\|_2 = \sqrt{\frac{\pi}{2}}.$$

$$\text{即 } T_1(x) = x, T_2(x) = 2x^2 - 1, T_3(x) = 4x^3 - 3x,$$

$$\text{所以 } T_1^*(x) = \sqrt{\frac{2}{\pi}} x, T_2^*(x) = \sqrt{\frac{2}{\pi}} (2x^2 - 1), T_3^*(x) = \sqrt{\frac{2}{\pi}} (4x^3 - 3x).$$

3. 连续函数的最小平方估计.

施在内积如 Thm 5.7 定义下, 用二次多项式估计函数  $y(x) = \sqrt{1-x^2}$ ,  $x \in [-1, 1]$ .

(1)  $\rho(x) = \frac{1}{\sqrt{1-x^2}}$ , 用傅里叶展开.

$$\text{Sol: } \langle y, 1 \rangle = \langle y, \sqrt{\frac{\pi}{2}} \rangle = \sqrt{\frac{\pi}{2}} \int_{-1}^1 \frac{1}{\sqrt{1-t^2}} \sqrt{1-t^2} dt = \sqrt{\frac{\pi}{2}} \int_{-1}^1 1 dt = \sqrt{\frac{\pi}{2}} \cdot 2 = \sqrt{2\pi}$$

$$\langle y, T_1^* \rangle = \langle y, \sqrt{\frac{2}{\pi}} x \rangle = \sqrt{\frac{2}{\pi}} \int_{-1}^1 \frac{1}{\sqrt{1-t^2}} t dt = 0.$$

$$\langle y, T_2^* \rangle = \langle y, \sqrt{\frac{2}{\pi}} (2x^2 - 1) \rangle = \sqrt{\frac{2}{\pi}} \int_{-1}^1 \frac{1}{\sqrt{1-t^2}} (2t^2 - 1) dt = \sqrt{\frac{2}{\pi}} \left( \frac{2}{3} t^3 - t \right) \Big|_{-1}^1 = -\frac{2}{3} \sqrt{\frac{2}{\pi}}.$$

$$\hat{\rho}(x) = \sqrt{\frac{2}{\pi}} \left( 2 - \frac{2}{3} (2x^2 - 1) \right) = \sqrt{\frac{2}{\pi}} \left( -\frac{4}{3} x^2 + \frac{8}{3} \right).$$

$$\hat{\rho}(x) = 2\sqrt{\frac{2}{\pi}} - \frac{2}{3} \sqrt{\frac{2}{\pi}} (2x^2 - 1).$$

$$\hat{\rho}(x) = 2\sqrt{\frac{2}{\pi}} \cdot \sqrt{\frac{2}{\pi}} - \frac{2}{3} \sqrt{\frac{2}{\pi}} \cdot \sqrt{\frac{2}{\pi}} (2x^2 - 1) = \frac{10 - 8x^2}{3\pi}.$$

$$\text{Sol: } G(1, x, x^2) = \begin{bmatrix} \langle 1, 1 \rangle & \langle 1, x \rangle & \langle 1, x^2 \rangle \\ \langle x, 1 \rangle & \langle x, x \rangle & \langle x, x^2 \rangle \\ \langle x^2, 1 \rangle & \langle x^2, x \rangle & \langle x^2, x^2 \rangle \end{bmatrix} = \begin{bmatrix} \frac{\pi}{2} & 0 & \frac{\pi}{8} \\ 0 & \frac{\pi}{2} & 0 \\ \frac{\pi}{8} & 0 & \frac{3\pi}{8} \end{bmatrix}$$

$$\langle x, x \rangle = \langle 1, x^2 \rangle = \frac{\pi}{8}$$

$$\langle 1, 1 \rangle = \int_{-1}^1 \frac{1}{\sqrt{1-x^2}} dx = \arcsin x \Big|_{-1}^1 = \pi.$$

$$\langle 1, x \rangle = \int_{-1}^1 \frac{x}{\sqrt{1-x^2}} dx = 0.$$

$$\langle x, x \rangle = \langle 1, x^2 \rangle = \int_{-1}^1 \frac{x^2}{\sqrt{1-x^2}} dx \stackrel{x=\cos\theta}{=} \int_0^\pi \frac{\cos^2\theta}{\sin\theta} d\cos\theta = \int_0^\pi \cos^2\theta d\theta = \frac{\pi}{2}.$$

$$\langle x, x^2 \rangle = \int_{-1}^1 \frac{x^3}{\sqrt{1-x^2}} dx = 0.$$

$$\langle x^2, x^2 \rangle = \int_{-1}^1 \frac{x^4}{\sqrt{1-x^2}} dx \stackrel{x=\cos\theta}{=} \int_0^\pi \frac{\cos^4\theta}{\sin\theta} d\cos\theta = \int_0^\pi \cos^4\theta d\theta = \frac{\Gamma(\frac{5}{2})\Gamma(\frac{1}{2})}{\Gamma(3)} = \frac{3\pi}{8}.$$

$$\therefore G(1, x, x^2) = \begin{bmatrix} \frac{\pi}{2} & 0 & \frac{\pi}{8} \\ 0 & \frac{\pi}{2} & 0 \\ \frac{\pi}{8} & 0 & \frac{3\pi}{8} \end{bmatrix}.$$

$$C = [\langle y, 1 \rangle, \langle y, x \rangle, \langle y, x^2 \rangle]^T.$$

$$\langle y, 1 \rangle = \int_{-1}^1 dx = 2.$$

$$\langle y, x \rangle = \int_{-1}^1 x dx = 0.$$

$$\langle y, x^2 \rangle = \int_{-1}^1 x^2 dx = \frac{2}{3}.$$

$$\therefore \begin{bmatrix} \frac{\pi}{2} & 0 & \frac{\pi}{8} \\ 0 & \frac{\pi}{2} & 0 \\ \frac{\pi}{8} & 0 & \frac{3\pi}{8} \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ \frac{2}{3} \end{bmatrix} \Rightarrow \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} \frac{10}{3\pi} \\ 0 \\ -\frac{8}{3\pi} \end{bmatrix}$$

$$\hat{p}(x) = \frac{10 - 8x^2}{3\pi}.$$

14. 用正交多项式进行离散最小二乘估计

考虑 Ex 5.48 的销售记录表格.

(1) 从线性无关集  $\{1, x, x^2\}$  构造正交多项式. 内积定义为  $\langle u(t), v(t) \rangle = \sum_{i=1}^N p(t_i) u(t_i) v(t_i)$ .

其中  $N=12, p(x)=1$ .

$$\text{Sol: } \langle u_0, u_0 \rangle = \langle 1, 1 \rangle = \sum_{i=1}^{12} 1 = 12. \quad u_0^* = \frac{u_0}{\|u_0\|_2} = \frac{1}{\sqrt{12}}.$$

$$\langle u_0^*, u_1 \rangle = \langle \frac{1}{\sqrt{12}}, x \rangle = \frac{1}{\sqrt{12}} \sum_{i=1}^{12} x_i = \frac{78}{\sqrt{12}}.$$

$$v_1 = u_1 - \langle u_0^*, u_1 \rangle u_0^* = x - \frac{13}{2}$$

$$\langle v_1, v_1 \rangle = \langle x - \frac{13}{2}, x - \frac{13}{2} \rangle = \sum_{i=1}^{12} (x_i - \frac{13}{2})^2 = 143. \quad u_1^* = \frac{v_1}{\|v_1\|_2} = \frac{x - \frac{13}{2}}{\sqrt{143}}.$$

$$\langle u_0^*, u_2 \rangle = \langle \frac{1}{\sqrt{12}}, x^2 \rangle = \frac{1}{\sqrt{12}} \sum_{i=1}^{12} x_i^2 = \frac{650}{\sqrt{12}}.$$

$$\langle u_0^*, u_2 \rangle = \langle \frac{x - \frac{13}{2}}{\sqrt{143}}, x^2 \rangle = \frac{1}{\sqrt{143}} \sum_{i=1}^{12} (x_i - \frac{13}{2}) x_i^2 = \frac{1859}{\sqrt{143}}.$$

$$v_2 = u_2 - \langle u_0^*, u_2 \rangle u_0^* - \langle u_1^*, u_2 \rangle u_1^* = x^2 - 13x + \frac{91}{3}.$$

$$\langle v_2, v_2 \rangle = \langle x^2 - 13x + \frac{91}{3}, x^2 - 13x + \frac{91}{3} \rangle = \sum_{i=1}^{12} (x_i^2 - 13x_i + \frac{91}{3})^2 = \frac{4004}{3}.$$

$$u_2 = \frac{v_2}{\|v_2\|_2} = \frac{\sqrt{3}}{\sqrt{4004}} (x^2 - 13x + \frac{91}{3}).$$

(2) 求最佳线性估计  $\hat{p} = \sum_{i=0}^2 a_i x^i$ , 使得  $\|y - \hat{p}\| \leq \|y - \sum_{i=0}^2 b_i x^i\|$ ,  $\forall b_i \in \mathbb{R}$ , 并验证  $\hat{p}$  和讲义中的  $\hat{p}$  相同.

Sol:  $\langle u_0, y \rangle = \langle \frac{1}{\sqrt{13}}, y \rangle = \frac{1}{\sqrt{13}} \sum_{i=1}^{12} y_i = \frac{1662}{\sqrt{13}}$ .

$\langle u_1, y \rangle = \langle \frac{x - \frac{13}{2}}{\sqrt{143}}, y \rangle = \frac{1}{\sqrt{143}} \sum_{i=1}^{12} (x_i - \frac{13}{2}) y_i = \frac{589}{\sqrt{143}}$ .

$\langle u_2, y \rangle = \langle \frac{\sqrt{3}}{\sqrt{4004}} (x^2 - 13x + \frac{91}{2}), y \rangle = \frac{\sqrt{3}}{\sqrt{4004}} \sum_{i=1}^{12} (x_i^2 - 13x_i + \frac{91}{2}) y_i = 12068 \frac{\sqrt{3}}{\sqrt{4004}}$ .

$\hat{p} = \frac{1662}{\sqrt{13}} \cdot \frac{1}{\sqrt{13}} + \frac{589}{\sqrt{143}} \cdot \frac{x - \frac{13}{2}}{\sqrt{143}} + 12068 \frac{\sqrt{3}}{\sqrt{4004}} \cdot \frac{\sqrt{3}}{\sqrt{4004}} (x^2 - 13x + \frac{91}{2})$

$= 9.04196 x^2 - 113.427 x + 386.$

(3) 假设另一张销售表格的格式与本题相同,  $N$  和  $x_i$  相同, 但  $y_i$  不同. 则哪些计算结果可重复使用 哪些不能?

这种可重用性体现了正交多项式相对于正则方程组的哪些优点?

Sol: 正交多项式结果  $u_0, u_1, u_2, \dots$  可重用.  $\langle u_0, y \rangle, \langle u_1, y \rangle, \langle u_2, y \rangle$  的值不可重用.

假设要用  $P$  次多项式拟合, 则对于正交多项式, 在用  $O(nP)$  次运算计算出  $u_0, u_1, \dots, u_P$  后, 对每个不同的  $y$ , 只需用  $O(nP)$  次运算计算出  $\langle u_0, y \rangle, \langle u_1, y \rangle, \dots, \langle u_P, y \rangle$  即可. 而对于正则化方法, 在用  $O(nP^2)$  次运算计算出  $G(1, x, \dots, x^P)$  后, 对于每个不同的  $y$ , 还需用  $O(nP)$  次运算计算出  $C$  后, 还需再用  $O(P^3)$  次运算解方程组才能得到  $\hat{p}$ . 在  $P$  较大时正则方程组方法的运算量远大于正交多项式方法.