

总分: 30 x 1.1 + 10 = 43

## I. Prove theorem 6.4 by assuming $\forall x \in (a, b)$ weight function $\rho(x) > 0$

### a. prove $L^2_\rho[a, b]$ is a vector space

+6

$\forall u, v, w \in L^2_\rho[a, b]$  and  $\forall a, b \in \mathbb{F}$  we can get following results easily

$$u + v = v + u \quad (1)$$

$$(u + v) + w = u + (v + w) \quad (2)$$

$$(ab)u = a(bu) \quad (3)$$

$$0 + u = u \quad (4)$$

$$1u = u \quad (5)$$

$$(a + b)u = au + bu \quad (6)$$

$$a(u + v) = au + av \quad (7)$$

where  $0 \in L^2_\rho[a, b]$  and  $1 \in \mathbb{F}$ . Besides,  $\forall u$ , there exists  $-u \in L^2_\rho[a, b]$  such that  $u + (-u) = 0$ , which satisfy all property of vector space. Hence proved.

### b. prove $L^2_\rho[a, b]$ is inner product space

$\forall u, v, w \in L^2_\rho[a, b]$  and  $\forall a \in \mathbb{F}$  assuming  $\rho(x) > 0$ , we can get following results easily

$$\langle v, v \rangle = \int_a^b \rho(x) v(x) \overline{v(x)} dx \geq 0 \quad (8)$$

$$(9)$$

Because  $\rho(x) > 0$ ,  $\langle v, v \rangle = 0$  iff  $v(x) = 0$ .

$$\langle u + v, w \rangle = \int_a^b \rho(x) (u(x) + v(x)) \overline{w(x)} dx = \int_a^b \rho(x) u(x) \overline{w(x)} dx + \int_a^b \rho(x) v(x) \overline{w(x)} dx = \langle u, w \rangle + \langle v, w \rangle \quad (10)$$

$$\langle av, w \rangle = \int_a^b a \rho(x) v(x) \overline{w(x)} dx = a \int_a^b \rho(x) v(x) \overline{w(x)} dx = a \langle v, w \rangle \quad (11)$$

$$\langle v, w \rangle = \int_a^b \rho(x) v(x) \overline{w(x)} dx = \int_a^b \rho(x) \overline{\overline{v(x)} w(x)} dx = \overline{\langle w, v \rangle} \quad (12)$$

Hence  $L^2_\rho[a, b]$  is an inner product space.

### c. prove $L^2_\rho[a, b]$ is norm space

From the definition of norm, we can know that  $\forall v \in L^2_\rho[a, b]$

$$\|v\|_2 = \left( \int_a^b \rho(x) |v(x)|^2 dx \right)^{\frac{1}{2}} = \left( \int_a^b \rho(x) v(x) \overline{v(x)} dx \right)^{\frac{1}{2}} = \sqrt{\langle v, v \rangle} \quad (13)$$

Hence proved.

## II. Consider Chebyshev polynomials of the first kind

### a. Show that they are orthogonal on $[-1, 1]$

+6

Because Chebyshev polynomials have form of  $T_n(x) = \cos(n \arccos x)$ , we can deduce

$$\langle T_n, T_m \rangle = \int_{-1}^1 \frac{\cos(n \arccos x) \cos(m \arccos x)}{\sqrt{1-x^2}} dx \quad (14)$$

Then we can take  $x = \cos \theta$  into the above equation as following, where  $\theta \in [0, \pi]$

$$\langle T_n, T_m \rangle = - \int_0^\pi \cos(m\theta) \cos(n\theta) d\theta = - \int_0^\pi \frac{\cos(m+n)\theta + \cos(m-n)\theta}{2} d\theta = 0 \quad (15)$$

Hence proved.

## b. Normalize the first three Chebyshev polynomials

The first three Chebyshev polynomials are as following

$$\begin{cases} u_1(x) = 1 \\ u_2(x) = x \\ u_3(x) = 2x^2 - 1 \end{cases}$$

We can deduce as following steps. Firstly,

$$v_1 = u_1 \quad (16)$$

$$u_1^* = \frac{v_1}{||v_1||} = \frac{1}{\sqrt{\pi}} \quad (17)$$

Secondly,

$$v_2 = u_2 = x \quad (18)$$

$$u_2^* = \frac{v_2}{||v_2||} = \sqrt{\frac{2}{\pi}}x \quad (19)$$

Thirdly,

$$v_3 = u_3 = 2x^2 - 1 \quad (20)$$

$$||v_3|| = \sqrt{\int_{-1}^1 \frac{(2x^2 - 1)^2}{\sqrt{1-x^2}} dx} = \sqrt{\int_{-\frac{\pi}{2}}^{\frac{\pi}{2}} (2\sin^2 \theta - 1)^2 d\theta} = \sqrt{\frac{\pi}{2}} \quad (21)$$

$$u_3^* = \frac{v_3}{||v_3||} = \sqrt{\frac{2}{\pi}}(2x^2 - 1) \quad (22)$$

## III. Least-square approximation of a continuous function

### a. $\rho(x) = \frac{1}{\sqrt{1-x^2}}$ with Fourier expansion

We select orthonormal polynomials in  $\mathbb{P}_2$  as  $u_1^* = \frac{1}{\sqrt{\pi}}$ ,  $u_2 = \sqrt{\frac{2}{\pi}}x$  and  $u_3 = \sqrt{\frac{2}{\pi}}(2x^2 - 1)$ , then we can deduce that

$$\langle y, u_1^* \rangle = \int_{-1}^1 \frac{1}{\sqrt{\pi}} 2 dx = \frac{2}{\sqrt{\pi}} \quad (23)$$

$$\langle y, u_2^* \rangle = \int_{-1}^1 \sqrt{\frac{2}{\pi}} x dx = 0 \quad (24)$$

$$\langle y, u_3^* \rangle = \int_{-1}^1 \sqrt{\frac{2}{\pi}} (2x^2 - 1) dx = -\frac{2}{3} \sqrt{\frac{2}{\pi}} \quad (25)$$

Therefore, the quadratic approximation of  $y$  is  $\varphi(x) = -\frac{8}{3\pi}x^2 + \frac{10}{3\pi}$ .

### b. $\rho(x) = \frac{1}{\sqrt{1-x^2}}$ with normal equation

We select linearly independent a set of basis as  $u_1 = 1$ ,  $u_2 = x$  and  $u_3 = x^2$ , then

$$G(u_1, u_2, u_3) = \begin{pmatrix} \langle u_1, u_1 \rangle & \langle u_1, u_2 \rangle & \langle u_1, u_3 \rangle \\ \langle u_2, u_1 \rangle & \langle u_2, u_2 \rangle & \langle u_2, u_3 \rangle \\ \langle u_3, u_1 \rangle & \langle u_3, u_2 \rangle & \langle u_3, u_3 \rangle \end{pmatrix} = \begin{pmatrix} \pi & 0 & \frac{\pi}{2} \\ 0 & \frac{\pi}{2} & 0 \\ \frac{\pi}{2} & 0 & \frac{3\pi}{8} \end{pmatrix}$$

$$c = (\langle y, u_1 \rangle \quad \langle y, u_2 \rangle \quad \langle y, u_3 \rangle) = (2 \quad 0 \quad \frac{2}{3})$$

As a result, we can get coefficients matrix by solving  $Ga^T = c^T$

$$a = \left( \frac{10}{3\pi} \quad 0 \quad -\frac{8}{3\pi} \right)$$

Therefore, the quadratic approximation of  $y$  is  $\varphi(x) = -\frac{8}{3\pi}x^2 + \frac{10}{3\pi}$ .

## IV. Discrete least square via orthonormal polynomials

### a. Construct orthonormal polynomials by the Gram-Schmidt process

The set of basis is as following

$$\begin{cases} u_1(x) = 1 \\ u_2(x) = x \\ u_3(x) = x^2 \end{cases} \quad +12$$

We can deduce as following steps. Firstly,

$$v_1 = u_1 = 1 \quad (26)$$

$$\|v_1\| = 2\sqrt{3} \approx 3.46 \quad (27)$$

$$u_1^* = \frac{v_1}{\|v_1\|} = \frac{\sqrt{3}}{6} \approx 0.29 \quad (28)$$

Secondly,

$$v_2 = u_2 - \langle u_2, u_1^* \rangle u_1^* = x - \frac{13}{2} \quad (29)$$

$$\|v_2\| = \sqrt{143} \approx 11.96 \quad (30)$$

$$u_2^* = \frac{v_2}{\|v_2\|} = \frac{x}{\sqrt{143}} - \frac{\sqrt{143}}{22} \approx \frac{x}{11.96} - 0.54 \quad (31)$$

Thirdly,

$$v_3 = u_3 - \langle u_3, u_1^* \rangle u_1^* - \langle u_3, u_2^* \rangle u_2^* \approx x^2 - 13x + 30.3 \quad (32)$$

$$\|v_3\| \approx 36.53 \quad (33)$$

$$u_3^* = \frac{v_3}{\|v_3\|} = \frac{1}{36.53}x^2 - \frac{13}{36.53}x + \frac{30.3}{36.53} \quad (34)$$

### b. Find the best approximation $\hat{\varphi} = \sum_{i=0}^2 a_i x^i$

$$G(u_1^*, u_2^*, u_3^*) = \begin{pmatrix} \langle u_1^*, u_1^* \rangle & \langle u_1^*, u_2^* \rangle & \langle u_1^*, u_3^* \rangle \\ \langle u_2^*, u_1^* \rangle & \langle u_2^*, u_2^* \rangle & \langle u_2^*, u_3^* \rangle \\ \langle u_3^*, u_1^* \rangle & \langle u_3^*, u_2^* \rangle & \langle u_3^*, u_3^* \rangle \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$c = (\langle y, u_1^* \rangle \quad \langle y, u_2^* \rangle \quad \langle y, u_3^* \rangle) = (481.98 \quad 55.03 \quad 328.84)$$

Then the normal equation yield

$$a = c = (481.98 \quad 55.03 \quad 328.84)$$

Hence,

$$\varphi(x) = 328.84u_3^* + 55.03u_2^* + 481.98u_1^* = 9.01x^2 - 112.42x + 382.82 \quad (35)$$

which is very similar to the answer in the note.

### c. Suppose there are other tables of sales. Which calculations can be reused?

The orthonormal polynomials and Gram matrix can be reused. But we need to recalculate

$$c = (\langle y, u_1^* \rangle \quad \langle y, u_2^* \rangle \quad \langle y, u_3^* \rangle)$$

The biggest advantage of orthonormal polynomials is that you can get the coefficients  $a$  as soon as you can get matrix  $c$  without solving equation  $a = G^{-1}c$ , because we have already known that  $G^{-1}$  is an identity matrix.

## Programming

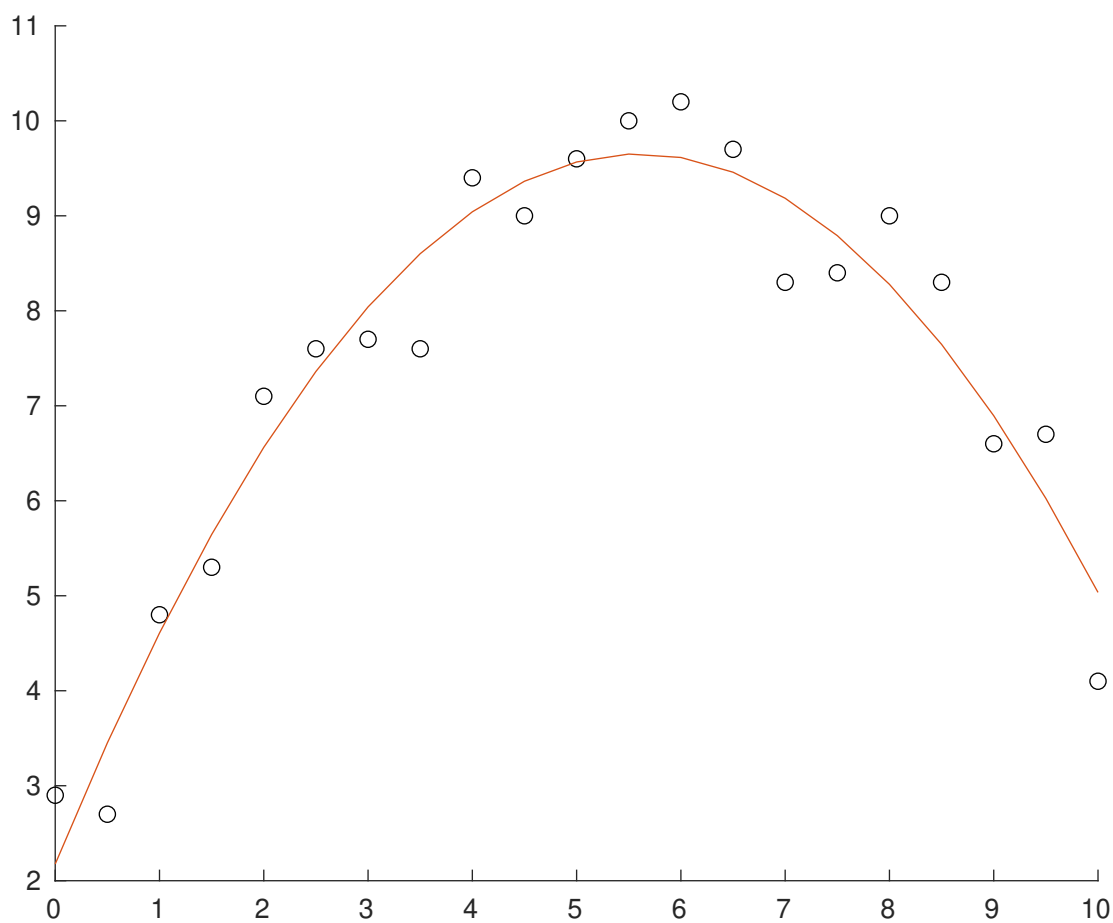


Figure 1: Discrete Least Square via normal equations

The best approximation I find is  $\varphi(x) = -0.238444 * x^2 + 2.67041 * x + 2.17572$  .