## Numerical Analysis HW5

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### **Theoretical Questions**

### **Question 1**

Proof:

(1) Positive Definite.

$$\forall u \in C[a, b], \ \langle u, v \rangle = \int_a^b p(x)u(x)\overline{u(x)}dx = \int_a^b p(x)u(x)^2dx > 0 \tag{0.1}$$

(2) Regularity.

$$\langle u, u \rangle = 0 \Leftrightarrow p(x)u(x)^2 = 0 \Leftrightarrow u = 0$$
 (0.2)

(3) Additivity.

$$\langle u + w, v \rangle = \int_{a}^{b} p(x)(u(x) + w(x))\overline{v(x)}dx$$

$$= \int_{a}^{b} p(x)u(x)\overline{v(x)}dx + \int_{a}^{b} p(x)w(x)\overline{v(x)}dx$$

$$= \langle u, v \rangle + \langle w, v \rangle$$

$$(0.3)$$

(4) Conjugate Symmetry.

$$\langle u, v \rangle = \int_{a}^{b} p(x)u(x)\overline{u(x)}dx$$

$$= \int_{a}^{b} \overline{p(x)u(x)}\overline{v(x)}dx$$

$$= \int_{a}^{b} p(x)\overline{u(x)}v(x)$$

$$= \overline{\langle v, u \rangle}$$

$$(0.4)$$

(5) Norm Positive Definite.

$$||u||_2 = \left[ \int_a^b \rho(x) |u(x)|^2 \, \mathrm{d}x \right]^{1/2} \ge 0$$
 (0.5)

(6) Norm Regularity.

$$||u||_2 = 0 \Leftrightarrow \rho(x)|u(x)|^2 = 0 \Leftrightarrow |u(x)|^2 = 0 \Leftrightarrow u = 0$$

$$(0.6)$$

(7) Norm Inequality.

$$\forall u, v \in \mathcal{C}[a, b], \quad \|u + v\|_{2} = \left[ \int_{a}^{b} \rho(x) |u(x) + v(x)|^{2} dx \right]^{1/2}$$

$$\leq \left[ \int_{a}^{b} \rho(x) |u(x)|^{2} dx \right]^{1/2} + \left[ \int_{a}^{b} \rho(x) |v(x)|^{2} dx \right]^{1/2}$$

$$= \|u\|_{2} + \|v\|_{2}$$

$$(0.7)$$

### **Question 2**

**(1)** 

We let  $p(x) = \frac{1}{\sqrt{1-x^2}}$  and the First Chebyshev Polynomial is,

$$T_n(x) = \cos(n \arccos x), \quad x \in [-1, 1], \quad n = 0, 1, 2, \dots$$
 (0.8)

therefore,

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

$$= \int_{-1}^1 \frac{1}{\sqrt{1 - x^2}} \cos(m \arccos x) \cos(n \arccos x) dx$$
Let  $\theta = \arccos x$ ,
$$= \int_0^{\pi} \cos(m\theta) \cos(n\theta) d\theta$$

$$= \frac{1}{2} \int_0^{\pi} [\cos(m + n)\theta + \cos(m - n)\theta] d\theta$$
(0.9)

Thus,

$$\langle T_m, T_n \rangle = \begin{cases} 0, & m \neq n \\ \pi, & m = n = 0 \\ \frac{\pi}{2}, & m = n \ge 1 \end{cases}$$
 (0.10)

Hence, the chebyshev polynomials are orthogonal.

**(2)** 

$$T_0^*(x) = \sqrt{\frac{1}{\pi}}, \quad T_1^*(x) = \sqrt{\frac{2}{\pi}}x, \quad T_2^*(x) = \sqrt{\frac{2}{\pi}}(2x^2 - 1).$$
 (0.11)

### **Question 3**

#### (1) Orthonormal Polynomials

According to the previous question, we have orthonormal polynomials,

$$T_0^*(x) = \sqrt{\frac{1}{\pi}}, \quad T_1^*(x) = \sqrt{\frac{2}{\pi}}x, \quad T_2^*(x) = \sqrt{\frac{2}{\pi}}(2x^2 - 1).$$
 (0.12)

Therefore, the approximation of the circular arc is given as follows,

$$\varphi(x) = \langle y, T_0^* \rangle T_0^* + \langle y, T_1^* \rangle T_1^* + \langle y, T_2^* \rangle T_2^* 
= \frac{2}{\pi} + 0 - \frac{4}{3\pi} (2x^2 - 1) 
= -\frac{8}{3\pi} x^2 + \frac{10}{3\pi}$$
(0.13)

#### (2) Normal Equation

Let  $\phi(x) = a_0 + a_1 x + a_2 x^2$ ,

$$\begin{bmatrix} \langle 1, 1 \rangle & \langle 1, x \rangle & \langle 1, x^2 \rangle \\ \langle x, 1 \rangle & \langle x, x \rangle & \langle x, x^2 \rangle \\ \langle x^2, 1 \rangle & \langle x^2, x \rangle & \langle x^2, x^2 \rangle \end{bmatrix}^T \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} = \begin{bmatrix} \langle y, 1 \rangle \\ \langle y, x \rangle \\ \langle y, x^2 \rangle \end{bmatrix}$$
(0.14)

that is,

$$\begin{bmatrix} \pi & 0 & \frac{\pi}{2} \\ 0 & \frac{\pi}{2} & 0 \\ \frac{\pi}{2} & 0 & \frac{3\pi}{2} \end{bmatrix}^{T} \begin{bmatrix} a_{0} \\ a_{1} \\ a_{2} \end{bmatrix} = \begin{bmatrix} 2 \\ 0 \\ \frac{2}{2} \end{bmatrix}$$
 (0.15)

Solving these linear equations, we have,

$$a_0 = \frac{10}{3\pi}, \ a_1 = 0, \ a_2 = \frac{-8}{3\pi}$$
 (0.16)

Therefore,

$$\phi(x) = -\frac{8}{3\pi}x^2 + \frac{10}{3\pi} \tag{0.17}$$

### **Question 4**

**(1)** 

Let the input  $u_1 = 1$ ,  $u_2 = x$ ,  $u_3 = x^2$ , by Gram-Schmidt,

$$v_{1} = u_{1}$$

$$u_{1}^{*} = v_{1} / \|v_{1}\|_{2} = v_{1} / \sqrt{12} = \frac{1}{2\sqrt{3}}$$

$$v_{2} = u_{2} - \langle u_{2}, u_{1}^{*} \rangle u_{1}^{*} = x - \frac{13}{2}$$

$$u_{2}^{*} = v_{2} / \|v_{2}\|_{2} = \frac{1}{\sqrt{143}} \left( x - \frac{13}{2} \right)$$

$$v_{3} = u_{3} - \langle u_{3}, u_{1}^{*} \rangle u_{1}^{*} - \langle u_{3}, u_{2}^{*} \rangle u_{2}^{*} = x^{2} - \frac{325}{6} - 13 \times \left( x - \frac{13}{2} \right) = x^{2} - 13x + \frac{91}{3}$$

$$u_{3}^{*} = v_{3} / \|v_{3}\|_{2} = \frac{\sqrt{3}}{\sqrt{4004}} \left( x^{2} - 13x + \frac{91}{3} \right)$$

$$(0.18)$$

**(2)** 

By corollary 5.25, the best approximation  $\varphi(x)$  is the projection on the orthonormal basis.

$$\hat{\varphi}(x) = \langle y, u_1^* \rangle u_1^* + \langle y, u_2^* \rangle u_2^* + \langle y, u_3^* \rangle u_3^*$$

$$= \frac{831}{\sqrt{3}} u_1^* + \frac{589}{\sqrt{143}} u_2^* + \frac{12068\sqrt{3}}{\sqrt{4004}} u_3^*$$

$$\approx 9.042 x^2 - 113.4266 x + 386.0013$$
(0.19)

**(3)** 

#### **Orthonormal Polynomials:**

For different  $y_i$ s, the orthonormal basis  $\{u_1^*, u_2^*, u_3^*\}$  can be reused, while the computation of the projection can not. So we only need to recompute  $\hat{\varphi}(x) = \langle y, u_1^* \rangle u_1^* + \langle y, u_2^* \rangle u_2^* + \langle y, u_3^* \rangle u_3^*$ .

#### **Normal Equation:**

For different  $y_i$ s, the coefficient matrix can be reused, but we have to recompute the vectors righthand, which requires us to compute n inner matrices and to solve the linear equations.

#### The advantage of orthonormal polynomials over normal equations:

- (1) The orthonormal basis can be reused in different approximation.
- (2) When using **ORTHONORMAL POLYNOMIALS**, we only need to recompute n inner products, while using **NORMAL EQUATION** requires us to additionally solve linear equation, which increases the amount of calculation and the condition number.

### **Programming Problems**

MATLAB is utilized for this part.

### Program 1

The input is x, y and we can find the coefficient in the output a.

$$a = [2.1757, 2.6704, -0.2384]^T (0.20)$$

Thus, the best approximation is,

$$\varphi(x) = 2.1752 + 2.6704x - 0.2384x^2 \tag{0.21}$$

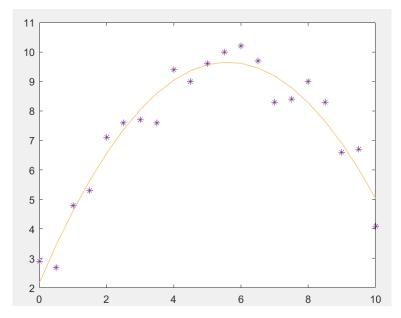


Figure 1: The approximation by Normal Equation

# Program 2

After QR Factorization for the Vandermonde Matrix and solving  $R_1 * x = c$ , we also have,

$$a = [2.1757, 2.6704, -0.2384]^T (0.22)$$

Thus, the best approximation is,

$$\hat{\varphi(x)} = 2.1752 + 2.6704x - 0.2384x^2 \tag{0.23}$$

which is the same as in Program 1, verifying the correctness.

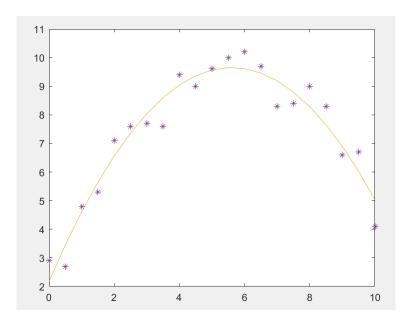


Figure 2: The approximation by QR

### Besides,

the condition number of G is 4.585153881696407e+04, the condition number of R1 is 2.141297242723767e+02.

Apparently, the condition number of the Gram matrix is larger than the condition number of the matrix  $R_1$  corresponding to the QR Factorization.