

## Research Article

# The Constants in A Posteriori Error Indicator for State-Constrained Optimal Control Problems with Spectral Methods

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We employ Legendre-Galerkin spectral methods to solve state-constrained optimal control problems. The constraint on the state variable is an integration form. We choose one-dimensional case to illustrate the techniques. Meanwhile, we investigate the explicit formulae of constants within a posteriori error indicator.

## 1. Introduction

Spectral methods provide higher accurate approximations with a relatively small number of unknowns and play increasingly important roles in design optimization, engineering design, and other scientific and engineering computations. Gottlieb and Orszag [1] summarized the theories and applications of spectral methods. There have been extensive researches on finite element methods for optimal control problems, most of which focus on control-constrained problems; see [2–5]. The authors studied the optimal control problems with the control constraint with spectral methods in [6]. In applications of engineering, one cares more about how to control the average value or  $L^2$ -norms of the state variable. The authors [7] discussed state-constrained optimal control problems with finite element methods. However, there are few work on the state-constrained optimal control problems with spectral methods.

In order to get a numerical solution with acceptable accuracy, spectral methods only increase the degree of basis when the error indicator is larger than the a posteriori error indicator, while the finite element methods refine meshes (see [8, 9]). There have been lots of papers on the a posteriori error estimates for h-version finite element

methods but not for spectral methods. Guo [10] got a reliable and efficient error indicator for  $p$ -version finite element method in one dimension with a certain weight. The authors [11] deduced a simple error indicator for spectral Galerkin methods. In [12], the authors investigated Legendre-Galerkin spectral method for optimal control problems with integral constraint on state. It is difficult to obtain optimal a posteriori error indicators. Thus, if one gets the constants within upper bound a posteriori error estimates, it is easy to ensure the degree of polynomials to get an acceptable accuracy.

In this paper, we employ Legendre-Galerkin spectral methods to solve optimal control problems with state-constrained case and calculate constants in upper bound of the a posteriori error indicator, which can be used to decide the least unknowns for acceptable accuracy. With the help of auxiliary systems, we investigate explicit formulae of the constants in the a posteriori error indicator.

The outline of this paper is as follows. In Section 2, the model problem and its Legendre-Galerkin spectral approximations are listed. In Section 3, the constants within the a posteriori error indicator are investigated in detail and the explicit formulae are obtained. The conclusions are given in Section 4.

## 2. A Model Problem and Its Legendre-Galerkin Spectral Approximations

Throughout this paper we adopt the standard notations of Sobolev spaces [13]. Let  $H^m(I)$  be a Sobolev space on  $I = (-1, 1)$ ,  $L^2(I) = H^0(I)$  and  $H_0^1(I) = \{v \in H^1(I) : v = 0 \text{ on } \partial I\}$ , and the corresponding norms are denoted by  $\|\cdot\|_m$ ,  $\|\cdot\|_0$ , and  $\|\cdot\|_{0,1}$ , respectively. This work focuses on the Legendre polynomials, which are orthogonal polynomials on  $[-1, 1]$ .

We concern the following distributed convex optimal control problems with integral constraint on state:

$$(OCP) \quad \begin{cases} \min_{y \in K} J(u, y) = \frac{1}{2} \int_I (y - y_d)^2 + \frac{\alpha}{2} \int_I u^2, \\ \text{s.t.} \quad -y'' = u \quad \text{in } I, \quad y = 0 \quad \text{on } \partial I, \quad y \in K, \end{cases} \quad (1)$$

where  $u \in U = L^2(I)$  is the control variable,  $y \in K = \{w : \int_I w \geq \gamma\} \subset H_0^1(I) \triangleq V$  is the state, and  $y_d \in L^2(I)$  is the observation.

In order to assure the existence and regularity of the solution, we assume that  $\alpha$  is a given positive constant and  $y_d$  is an infinitely smooth function. It is well known that the problem (OCP) has a unique solution (see [3]).

We give some basic notations which will be used in the sequel. Let

$$\begin{aligned} (v, w) &= \int_I vw, \quad \forall v, w \in L^2(I), \\ a(v, w) &= \int_I v'w', \quad \forall v, w \in H_0^1(I). \end{aligned} \quad (2)$$

Hence, the state equation reduces to

$$a(y, w) = (u, w), \quad \forall w \in H_0^1(I). \quad (3)$$

Then (OCP) can be rewritten as finding  $(u, y)$  such that

$$(\mathcal{P}) \quad \begin{cases} \min_{y \in K} J(u, y) = \frac{1}{2} \int_I (y - y_d)^2 + \frac{\alpha}{2} \int_I u^2, \\ \text{s.t.} \quad a(y(u), w) = (u, w), \quad \forall w \in V. \end{cases} \quad (4)$$

We recall the following optimal conditions of  $(\mathcal{P})$  (for details please refer to [7]).

**Lemma 1.** *The pair  $(u, y) \in U \times V$  is the optimal solution of  $(\mathcal{P})$  if and only if there exists a unique pair  $(p, \lambda) \in V \times \mathbb{R}_-^1$  ( $\mathbb{R}_-^1 \triangleq \{c \in \mathbb{R}^1; c \leq 0\}$ ) such that*

$$(OCP - OPT) \quad \begin{cases} a(y, w) = (u, w), \quad \forall w \in V, \\ a(q, p) \\ \quad = (y - y_d, q) + \lambda(1, q), \quad \forall q \in V, \\ \lambda(w - y) \leq 0, \quad \forall w \in K, \\ p + \alpha u = 0. \end{cases} \quad (5)$$

Let  $\mathcal{P}_N(I) = \{\text{polynomials of degree } \leq N \text{ on } I\}$  and let  $V_N = \mathcal{P}_N \cap H_0^1(I)$ . One prefers to choose appropriate bases

of  $V_N$  such that the resulting linear system is as simple as possible. We denote by  $\{L_j\}_{j=0}^N$  the Legendre polynomial and employ the following basis functions (see [14]):

$$\begin{aligned} U_N &= \text{span} \{L_0(x), L_1(x), \dots, L_N(x)\}, \\ V_N &= \text{span} \{\phi_0(x), \phi_1(x), \dots, \phi_{N-2}(x)\}, \end{aligned} \quad (6)$$

where

$$\phi_i(x) = c_i(L_i(x) - L_{i+2}(x)), \quad c_i = \frac{1}{\sqrt{4i+6}}. \quad (7)$$

For  $1 \leq j, k \leq N-2$ , we denote  $a_{jk} = a(\phi_k(x), \phi_j(x))$  and  $b_{jk} = (\phi_k(x), \phi_j(x))$ . By simple calculations, these coefficients satisfy

$$\begin{aligned} a_{jk} &= \begin{cases} 1, & k = j, \\ 0, & k \neq j, \end{cases} \\ b_{jk} = b_{kj} &= \begin{cases} c_k c_j \left( \frac{2}{2j+1} + \frac{2}{2j+5} \right), & k = j, \\ -c_k c_j \frac{2}{2k+1}, & k = j+2, \\ 0, & \text{otherwise.} \end{cases} \end{aligned} \quad (8)$$

Then Legendre-Galerkin spectral approximations of (OCP) can be read as finding  $(u_N, y_N)$  such that

$$(\mathcal{P}^N) \quad \begin{cases} \min_{y_N \in K} J(u_N, y_N) = \frac{1}{2} \int_I (y_N - y_d)^2 + \frac{\alpha}{2} \int_I u_N^2, \\ \text{s.t.} \quad a(y_N, w_N) = (u_N, w_N), \quad \forall w_N \in V_N. \end{cases} \quad (9)$$

The Legendre-Galerkin spectral approximations of (5) can be read as follows.

**Theorem 2.** *The pair  $(u_N, y_N) \in U_N \times V_N$  is the optimal solution of  $(\mathcal{P}^N)$  if and only if there exists a unique pair  $(p_N, \lambda_N) \in V_N \times \mathbb{R}_-^1$  such that*

$$(OCP - OPT)^N \quad \begin{cases} a(y_N, v_N) \\ \quad = (u_N, v_N), \quad \forall v_N \in V_N, \\ a(q_N, p_N) \\ \quad = (y_N - y_d, q_N) \\ \quad \quad + \lambda_N(1, q_N), \quad \forall q_N \in V_N, \\ \lambda_N(1, w_N - y_N) \leq 0, \quad \forall w_N \in K_N, \\ \alpha u_N + p_N = 0. \end{cases} \quad (10)$$

## 3. Constants within the A Posteriori Error Estimates

In this section, we calculate all constants within the a posteriori error estimates. Here, we analyze the constant in the Poincaré inequality.

For all  $v \in W_0^{1,p}(I)$ ,  $1 \leq p < \infty$ , we recall the Poincaré inequality with  $L^2$ -norm as (see [15])

$$\|v\|_0 \leq \frac{|I|}{2} \|v'\|_0. \quad (11)$$

Now, we are at the point to investigate all constants in detail. We introduce an auxiliary state  $y(u_N) \in H_0^1(I)$ , which satisfies

$$a(y(u_N), w) = (u_N, w), \quad \forall w \in H_0^1(I). \quad (12)$$

Subtracting (12) from (3), we get

$$a(y - y(u_N), w) = (u - u_N, w), \quad \forall w \in H_0^1(I). \quad (13)$$

Let  $w = y(u_N) - y \in H_0^1(I)$ . It is clear that

$$a(y(u_N) - y, y(u_N) - y) = (u_N - u, y(u_N) - y). \quad (14)$$

Then

$$\begin{aligned} \|(y(u_N) - y)'\|_0^2 &\leq \|u_N - u\|_0 \|y(u_N) - y\|_0 \\ &\leq \frac{|I|}{2} \|u_N - u\|_0 \|(y(u_N) - y)'\|_0, \end{aligned} \quad (15)$$

which means

$$\|(y(u_N) - y)'\|_0 \leq \frac{|I|}{2} \|u_N - u\|_0. \quad (16)$$

Hence

$$\begin{aligned} \|y(u_N) - y\|_1 &\leq \left( \|(y(u_N) - y)'\|_0^2 + \left(\frac{|I|}{2}\right)^2 \|(y(u_N) - y)'\|_0^2 \right)^{1/2} \\ &= \left( 1 + \left(\frac{|I|}{2}\right)^2 \right)^{1/2} \|(y(u_N) - y)'\|_0. \end{aligned} \quad (17)$$

Then

$$\|y(u_N) - y\|_1 \leq \left( 1 + \left(\frac{|I|}{2}\right)^2 \right)^{1/2} \frac{|I|}{2} \|u_N - u\|_0. \quad (18)$$

We denote by  $c_1$  the constant in (18); that is,

$$c_1 = \left( 1 + \left(\frac{|I|}{2}\right)^2 \right)^{1/2} \frac{|I|}{2}. \quad (19)$$

Similarly, we introduce an auxiliary state  $p(u_N) \in H^1(I)$ , which satisfies

$$a(q, p(u_N)) = (y(u_N) - y_d, q) + \lambda(1, q), \quad \forall q \in H_0^1(I). \quad (20)$$

Subtracting (20) from the continuous systems (5), we get

$$\begin{aligned} a(p - p(u_N), w) &= (y - y(u_N), w) + (\lambda - \lambda_N)(1, w), \\ &\forall w \in H_0^1(I). \end{aligned} \quad (21)$$

We select  $\varphi \in C_0^\infty(I)$  which satisfies  $\bar{\varphi} = 1$ , where  $\bar{\varphi} \triangleq \int_I \varphi/|I| = 1$  denotes the integral average on  $I$  of the function

$\varphi$  and  $\|\varphi\|_1 \leq C_\varphi$ . Obviously,  $\overline{p - p(u_N)\varphi} \in C_0^\infty(I)$ . In fact,  $p - p(u_N) - \overline{p - p(u_N)\varphi} \in H_0^1(I)$ . Then there hold

$$\begin{aligned} &\|p - p(u_N)\|_1^2 \\ &\leq \left( 1 + \left(\frac{|I|}{2}\right)^2 \right) |p - p(u_N)|_1^2 \\ &= \left( 1 + \left(\frac{|I|}{2}\right)^2 \right) a(p - p(u_N), p - p(u_N)) \\ &= \left( 1 + \left(\frac{|I|}{2}\right)^2 \right) \\ &\quad \times \left\{ a(\overline{p - p(u_N)\varphi}, p - p(u_N)) \right. \\ &\quad \left. + (y - y(u_N), p - p(u_N) - \overline{p - p(u_N)\varphi}) \right\} \\ &\leq \left( 1 + \left(\frac{|I|}{2}\right)^2 \right) \\ &\quad \times \left\{ |\overline{p - p(u_N)\varphi}| \cdot \left( \frac{\epsilon_1}{2} |\varphi|_1^2 + \frac{1}{2\epsilon_1} |p - p(u_N)|_1^2 \right) \right. \\ &\quad \left. + \frac{\epsilon_2}{2} \|y - y(u_N)\|_0^2 \right. \\ &\quad \left. + \frac{1}{\epsilon_2} \left( \|p - p(u_N)\|_0^2 + |\overline{p - p(u_N)\varphi}|^2 \cdot \|\varphi\|_0^2 \right) \right\}, \end{aligned} \quad (22)$$

where we used the generalized Schwarz inequality, continuous systems (5), and auxiliary equation (20). Let

$$\left( 1 + \left(\frac{|I|}{2}\right)^2 \right) \frac{|\overline{p - p(u_N)\varphi}|}{2\epsilon_1} = \left( 1 + \left(\frac{|I|}{2}\right)^2 \right) \frac{1}{\epsilon_2} = \frac{1}{2}. \quad (23)$$

Then

$$\epsilon_1 = \left( 1 + \left(\frac{|I|}{2}\right)^2 \right) |\overline{p - p(u_N)\varphi}|, \quad \epsilon_2 = 2 \left( 1 + \left(\frac{|I|}{2}\right)^2 \right). \quad (24)$$

It is clear that (22) reduces to

$$\begin{aligned}
 & \|p - p(u_N)\|_1^2 \\
 & \leq \left(1 + \left(\frac{|I|}{2}\right)^2\right) \\
 & \times \left\{ \frac{1 + (|I|/2)^2}{2} \overline{|p - p(u_N)|^2} \cdot |\varphi|_1^2 \right. \\
 & \quad + \frac{\overline{|p - p(u_N)|^2}}{2(1 + (|I|/2)^2)} \|\varphi\|_0^2 \\
 & \quad \left. + \left(1 + \left(\frac{|I|}{2}\right)^2\right) \|y - y(u_N)\|_0^2 \right\} \\
 & + \frac{1}{2} \|p - p(u_N)\|_1^2.
 \end{aligned} \tag{25}$$

Then

$$\begin{aligned}
 & \|p - p(u_N)\|_1^2 \\
 & \leq 2 \left(1 + \left(\frac{|I|}{2}\right)^2\right) \\
 & \times \left\{ \left(1 + \left(\frac{|I|}{2}\right)^2\right) \overline{|p - p(u_N)|^2} \cdot \|\varphi\|_1^2 \right. \\
 & \quad \left. + \left(1 + \left(\frac{|I|}{2}\right)^2\right) \|y - y(u_N)\|_0^2 \right\} \\
 & = 2 \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2 \left\{ \overline{|p - p(u_N)|^2} \cdot \|\varphi\|_1^2 \right. \\
 & \quad \left. + \|y - y(u_N)\|_0^2 \right\},
 \end{aligned} \tag{26}$$

where we used  $(1 + (|I|/2)^2) \overline{|p - p(u_N)|^2} \geq (1/(1 + (|I|/2)^2)) \overline{|p - p(u_N)|^2}$ .  
Hence

$$\begin{aligned}
 & \|p - p(u_N)\|_1^2 \\
 & \leq 2 \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2
 \end{aligned}$$

$$\begin{aligned}
 & \times \left\{ \|\varphi\|_1^2 \cdot \frac{1}{|I|} (\alpha \|u - u_N\|_0 + \|p_N - p(u_N)\|_0)^2 \right. \\
 & \quad \left. + \left(1 + \left(\frac{|I|}{2}\right)^2\right) \left(\frac{|I|}{2}\right)^2 \|u - u_N\|_0^2 \right\} \\
 & = 2 \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2 \\
 & \times \left\{ \left[ \frac{2\|\varphi\|_1^2 \alpha^2}{|I|} + \left(1 + \left(\frac{|I|}{2}\right)^2\right) \left(\frac{|I|}{2}\right)^2 \right] \right. \\
 & \quad \left. \times \|u - u_N\|_0^2 + \frac{2\|\varphi\|_1^2}{|I|} \|p_N - p(u_N)\|_0^2 \right\} \\
 & = 2 \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2 \max \left\{ \frac{2\|\varphi\|_1^2 \alpha^2}{|I|} + c_1^2, \frac{2\|\varphi\|_1^2}{|I|} \right\} \\
 & \times \{ \|u - u_N\|_0^2 + \|p_N - p(u_N)\|_0^2 \},
 \end{aligned} \tag{27}$$

which means that

$$\begin{aligned}
 & \|p - p(u_N)\|_1 \\
 & \leq \left(2 \left(1 + \left(\frac{|I|}{2}\right)^2\right)\right)^2 \\
 & \times \max \left\{ \frac{2C_\varphi^2 \alpha^2}{|I|} + \left(1 + \left(\frac{|I|}{2}\right)^2\right) \left(\frac{|I|}{2}\right)^2, \frac{2C_\varphi^2}{|I|} \right\}^{1/2} \\
 & \times \{ \|u - u_N\|_0 + \|p_N - p(u_N)\|_0 \}.
 \end{aligned} \tag{28}$$

Denote by  $c_2$  the constant in (28). With simple calculation, we have

$$c_2 = \sqrt{2 \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2 \max \left\{ \frac{2C_\varphi^2 \alpha^2}{|I|} + c_1^2, \frac{2C_\varphi^2}{|I|} \right\}}. \tag{29}$$

We select  $\varphi \in C_0^\infty(I)$  which satisfies  $\bar{\varphi} = 1$  and  $\|\varphi\|_1 \leq C_\varphi$ . For instance,  $\varphi = (3/2)(1 - x^2)$ , which satisfies

$$\|\varphi\|_1 = \frac{3}{2} \sqrt{\frac{74}{15}} \triangleq C_\varphi. \tag{30}$$

Meanwhile,

$$\begin{aligned}
 & (\lambda - \lambda_N, (\lambda - \lambda_N) \varphi) \\
 & = (\lambda - \lambda_N)^2 \int_I \varphi = (\lambda - \lambda_N)^2 |I| = \|\lambda - \lambda_N\|_0^2.
 \end{aligned} \tag{31}$$

Hence

$$\begin{aligned}
 & \|\lambda - \lambda_N\|_0^2 \\
 &= (\lambda - \lambda_N, (\lambda - \lambda_N)\varphi) \\
 &= a((\lambda - \lambda_N)\varphi, p - p(u_N)) \\
 &\quad - (y - y(u_N), (\lambda - \lambda_N)\varphi) \\
 &\leq |\lambda - \lambda_N| \{ \|\varphi'\|_0 \cdot \|(p - p(u_N))'\|_0 \\
 &\quad + \|\varphi\|_0 \cdot \|y - y(u_N)\|_0 \} \\
 &\leq \frac{\epsilon_1}{2} |\lambda - \lambda_N|^2 \cdot \|\varphi'\|_0^2 + \frac{1}{2\epsilon_1} \|(p - p(u_N))'\|_0^2 \\
 &\quad + \frac{\epsilon_2}{2} |\lambda - \lambda_N|^2 \cdot \|\varphi\|_0^2 + \frac{1}{2\epsilon_2} \|y - y(u_N)\|_0^2 \\
 &= \frac{\epsilon_1}{2|I|} \|\lambda - \lambda_N\|_0^2 \|\varphi\|_1^2 \\
 &\quad + \frac{1}{2\epsilon_1} (\|(p - p(u_N))'\|_0^2 + \|y - y(u_N)\|_0^2),
 \end{aligned} \tag{32}$$

where  $\epsilon_1 = \epsilon_2 = |I|/\|\varphi\|_1^2$ .

Thus

$$\begin{aligned}
 \|\lambda - \lambda_N\|_0^2 &\leq \frac{\|\varphi\|_1^2}{|I|} \{ \|(p - p(u_N))'\|_0^2 + \|y - y(u_N)\|_0^2 \} \\
 &\leq \frac{\|\varphi\|_1^2}{|I|} \{ \|p - p(u_N)\|_1^2 + \|y - y(u_N)\|_0^2 \}.
 \end{aligned} \tag{33}$$

With the constant  $c_2$ , we infer that

$$\begin{aligned}
 & \|\lambda - \lambda_N\|_0^2 \\
 &\leq \frac{C_\varphi^2}{|I|} \left\{ 4 \left( 1 + \left( \frac{|I|}{2} \right)^2 \right)^2 \right. \\
 &\quad \times \max \left\{ \frac{2C_\varphi^2 \alpha^2}{|I|} + \left( 1 + \left( \frac{|I|}{2} \right)^2 \right) \left( \frac{|I|}{2} \right)^2, \frac{2C_\varphi^2}{|I|} \right\} \\
 &\quad \left. + \left( 1 + \left( \frac{|I|}{2} \right)^2 \right) \left( \frac{|I|}{2} \right)^2 \right\} \\
 &\quad \times \{ \|p_N - p(u_N)\|_0^2 + \|u - u_N\|_0^2 \} \\
 &= \frac{C_\varphi^2}{|I|} \{ 2c_2^2 + c_1^2 \} \{ \|p_N - p(u_N)\|_0^2 + \|u - u_N\|_0^2 \}.
 \end{aligned} \tag{34}$$

Then

$$\begin{aligned}
 & \|\lambda - \lambda_N\|_0 \\
 &\leq \frac{C_\varphi}{|I|} \{ 2c_2^2 + c_1^2 \} \{ \|p_N - p(u_N)\|_0 + \|u - u_N\|_0 \}.
 \end{aligned} \tag{35}$$

We denote by  $c_3$  the constant in (35); that is,

$$c_3 = \sqrt{\frac{C_\varphi^2}{|I|} \{ 2c_2^2 + c_1^2 \}}. \tag{36}$$

We calculate the error of  $u$  in  $L^2$ -norm as follows:

$$\begin{aligned}
 & \|u - u_N\|_0^2 \\
 &\leq (p_N - p(u_N), u - u_N) \\
 &\quad - (\lambda - \lambda_N, y_N - y(u_N)) \\
 &\leq \frac{\epsilon_1}{2} \|p_N - p(u_N)\|_0^2 + \frac{1}{2\epsilon_1} \|u - u_N\|_0^2 \\
 &\quad + \frac{1}{2\epsilon_2} \|y_N - y(u_N)\|_0^2 \\
 &\quad + \epsilon_2 \cdot c_3^2 \{ \|u - u_N\|_0^2 + \|p_N - p(u_N)\|_0^2 \} \\
 &= \left( \frac{\epsilon_1}{2} + \epsilon_2 \cdot c_3^2 \right) \|p_N - p(u_N)\|_0^2 \\
 &\quad + \left( \frac{1}{2\epsilon_1} + \epsilon_2 \cdot c_3^2 \right) \|u - u_N\|_0^2 \\
 &\quad + \frac{1}{2\epsilon_2} \|y_N - y(u_N)\|_0^2.
 \end{aligned} \tag{37}$$

Provided that  $1/2\epsilon_1 + \epsilon_2 \cdot c_3^2 = 1/2$ , we get

$$\begin{aligned}
 & \|u - u_N\|_0^2 \\
 &\leq 2 \left( \frac{\epsilon_1}{2} + \epsilon_2 \cdot c_3^2 \right) \|p_N - p(u_N)\|_0^2 \\
 &\quad + \frac{1}{\epsilon_2} \|y_N - y(u_N)\|_0^2 \\
 &= \max \left\{ \epsilon_1 + 2\epsilon_2 \cdot c_3^2, \frac{1}{\epsilon_2} \right\} \{ \|p_N - p(u_N)\|_0^2 \\
 &\quad + \|y_N - y(u_N)\|_0^2 \}.
 \end{aligned} \tag{38}$$

Considering the item  $\max\{\epsilon_1 + 2\epsilon_2 \cdot c_3^2, 1/\epsilon_2\}$  with the constraint  $1/2\epsilon_1 + \epsilon_2 \cdot c_3^2 = 1/2$ , we get

$$F(\epsilon_1) = \max \left\{ \epsilon_1 + 1 - \frac{1}{\epsilon_1}, \frac{2c_3^2 \epsilon_1}{\epsilon_1 - 1} \right\}. \tag{39}$$

In fact, for  $\forall \epsilon_1 > 1$ , the derivation of the following function

$$f(\epsilon_1) = \epsilon_1 + 1 - \frac{1}{\epsilon_1} - \frac{2c_3^2 \epsilon_1}{\epsilon_1 - 1} \tag{40}$$

is

$$f'(\epsilon_1) = 1 + \frac{1}{\epsilon_1^2} + \frac{2c_3^2}{(\epsilon_1 - 1)^2} \geq 0. \tag{41}$$

Then we have  $\epsilon_1^0 = 2c_3^2 + \epsilon > 1$  and

$$\lim_{\epsilon \rightarrow 0} f(\epsilon_1^0) = 0. \quad (42)$$

Now, we are at the point to investigate

$$\min F(\epsilon_1). \quad (43)$$

If  $\epsilon_1 = \epsilon_1^0$ , we get

$$\min F(\epsilon_1) = \epsilon_1^0 + 1 - \frac{1}{\epsilon_1^0} \leq 2c_3^2 + \epsilon + 1. \quad (44)$$

If  $1 < \epsilon_1 < \epsilon_1^0$ ,  $f(\epsilon_1) < 0$ , we obtain

$$F(\epsilon_1) = \frac{2c_3^2 \epsilon_1}{\epsilon_1 - 1}, \quad F'(\epsilon_1) = -\frac{2c_3^2}{(\epsilon_1 - 1)^2} < 0. \quad (45)$$

Then

$$\lim_{\epsilon' \rightarrow 0} \min F(\epsilon_1) = 2c_3^2 \left( 1 + \frac{1}{2c_3^2 + \epsilon - \epsilon' - 1} \right) < 4c_3^2. \quad (46)$$

If  $\epsilon_1 > \epsilon_1^0$ ,  $f(\epsilon_1) > 0$ , we infer that

$$F(\epsilon_1) = \epsilon_1 + 1 - \frac{1}{\epsilon_1}, \quad F'(\epsilon_1) = 1 + \frac{1}{\epsilon_1^2} > 0. \quad (47)$$

Then there hold

$$\lim_{\epsilon'' \rightarrow 0} \min F(\epsilon_1) = 2c_3^2 + \epsilon + \epsilon'' + 1 - \frac{1}{2c_3^2 + \epsilon + \epsilon''} < 2c_3^2 + 1. \quad (48)$$

Combining the above discussions, we deduce that

$$\min \left\{ \max \left\{ \epsilon_1 + 2\epsilon_2 \cdot c_3^2, \frac{1}{\epsilon_2} \right\} \right\} < 2c_3^2 + 1. \quad (49)$$

Obviously,

$$\begin{aligned} & \|u - u_N\|_0 \\ & \leq \sqrt{2c_3^2 + 1} \{ \|p_N - p(u_N)\|_0 + \|y_N - y(u_N)\|_0 \}. \end{aligned} \quad (50)$$

We denote by  $c_4$  the constant in (50); that is,

$$c_4 = \sqrt{2c_3^2 + 1}. \quad (51)$$

For any  $v \in L^2(I)$ , we define a projection operator  $\mathbb{P}_N : L^2(I) \rightarrow V_N$ , which satisfies

$$(\mathbb{P}_N v - v, w_N) = 0, \quad \forall w_N \in U_N. \quad (52)$$

**Lemma 3.** For all  $v \in H^\sigma(I)$  ( $\sigma \geq 0$ ), one has

$$\|\mathbb{P}_N v - v\|_0 \leq c_5 N^{-\sigma} \|v\|_\sigma, \quad (53)$$

where  $c_5 = 2\sqrt{2}$ .

*Proof.* Firstly, assuming that  $\sigma = 2p$  ( $p \geq 1$ ) is integer, we define a differential operator as

$$A = \frac{d}{dx} \left( (1-x^2) \frac{d}{dx} \right). \quad (54)$$

From the fact that

$$\frac{d}{dx} \left( (1-x^2) \frac{dL_k}{dx} \right) + k(k+1)L_k = 0, \quad (55)$$

it is easy to get

$$\begin{aligned} \hat{v}_k &= \left( k + \frac{1}{2} \right) (v, L_k) \\ &= \frac{k+1/2}{k(k+1)} \int_{-1}^1 AL_k(x) v(x) dx \\ &= -\frac{k+1/2}{k(k+1)} \int_{-1}^1 Av(x) L_k(x) dx \\ &= -\frac{k+1/2}{k(k+1)} (Av(x), L_k(x)). \end{aligned} \quad (56)$$

By iterations, we obtain

$$\hat{v}_k = \left( \frac{-1}{k(k+1)} \right)^p \left( k + \frac{1}{2} \right) (A^p v, L_k). \quad (57)$$

Secondly, for all  $v \in H^{2p}(I)$ , we note that  $A^p v = \sum_{i=0}^{\infty} \alpha_i L_i(x)$  and

$$\begin{aligned} (A^p v(x), L_k(x)) &= \alpha_i \left( k + \frac{1}{2} \right)^{-1}, \\ \|A^p v\|_0^2 &= \sum_{k=0}^{\infty} |\alpha_k|^2 \left( k + \frac{1}{2} \right)^{-1}. \end{aligned} \quad (58)$$

Hence

$$\begin{aligned} \|\mathbb{P}_N v - v\|_0^2 &= \sum_{k=N+1}^{\infty} \left( k + \frac{1}{2} \right)^{-1} |\hat{v}_k|^2 \\ &= \sum_{k=N+1}^{\infty} \left( \frac{1}{k(k+1)} \right)^{2p} \left( k + \frac{1}{2} \right) |A^p v, L_k|^2 \\ &\leq N^{-4p} \sum_{k=N+1}^{\infty} \left( k + \frac{1}{2} \right) |\alpha_k|^2 \left( k + \frac{1}{2} \right)^{-2} \\ &\leq N^{-4p} \|A^p v\|_0^2. \end{aligned} \quad (59)$$

Finally, there hold

$$\begin{aligned} |Av|^2 &= \left| \frac{d}{dx} \left( (1-x^2) \frac{dv}{dx} \right) \right|^2 \\ &= \left| (1-x^2)v'' - 2xv' \right|^2 \\ &\leq (|v''| + 2|v'|)^2 \\ &\leq 2v''^2 + 8v'^2 \\ &\leq 8 \{ v''^2 + v'^2 + v^2 \}, \end{aligned} \quad (60)$$

which means that

$$\|Av\|_0^2 \leq 8\|v\|_2^2. \quad (61)$$

Let  $p = 1, \sigma = 2$ . It is clear that

$$\|\mathbb{P}_N v - v\|_0 \leq \sqrt{8} N^{-2} \|v\|_2. \quad (62)$$

This completes the proof.  $\square$

Now, we are at the point to calculate the constant for  $\|y_N - y(u_N)\|_1 + \|p_N - p(u_N)\|_1$ . Similarly, let  $E^p = p_N - p(u_N)$  and let  $E_I^p = \mathbb{P}_N E^p \in V_N$ . Then

$$\begin{aligned} & \|p_N - p(u_N)\|_1^2 \\ & \leq \left(1 + \left(\frac{|I|}{2}\right)^2\right) (a(E^p, E^p - E_I^p) + (y(u_N) - y_N, E_I^p)) \\ & = \left(1 + \left(\frac{|I|}{2}\right)^2\right) (a(p(u_N) - p_N, E^p - E_I^p) \\ & \quad + (y(u_N) - y_N, E_I^p)) \\ & = \left(1 + \left(\frac{|I|}{2}\right)^2\right) \\ & \quad \times ((-p''(u_N), E^p - E_I^p) \\ & \quad + (p_N'', E^p - E_I^p) + (y(u_N) - y_N, E_I^p)) \\ & = \left(1 + \left(\frac{|I|}{2}\right)^2\right) ((y_N - y_d + \lambda_N + p_N'', E^p - E_I^p) \\ & \quad + (y(u_N) - y_N, E^p)) \\ & \leq \left(1 + \left(\frac{|I|}{2}\right)^2\right) \|E^p\|_1 \{c_5 N^{-1} \|y_N - y_d + \lambda_N + p_N''\|_0 \\ & \quad + \|y_N - y(u_N)\|_0\}, \end{aligned} \quad (63)$$

which means that

$$\begin{aligned} & \|p_N - p(u_N)\|_1 \\ & \leq \left(1 + \left(\frac{|I|}{2}\right)^2\right) \{c_5 N^{-1} \|y_N - y_d + \lambda_N + p_N''\|_0 \\ & \quad + \|y_N - y(u_N)\|_0\}. \end{aligned} \quad (64)$$

Likewise, let  $E^y = y_N - y(u_N)$  and let  $E_I^y = \mathbb{P}_{1,N}^0 E^y \in V_N$ . Then

$$\begin{aligned} & \|y_N - y(u_N)\|_1^2 \\ & = \|E^y\|_1^2 \leq \left(1 + \left(\frac{|I|}{2}\right)^2\right) a(E^y, E^y) \\ & = \left(1 + \left(\frac{|I|}{2}\right)^2\right) a(E^y - E_I^y, E^y) \\ & = \left(1 + \left(\frac{|I|}{2}\right)^2\right) (- (u_N + y_N''), E^y - E_I^y) \\ & \leq \left(1 + \left(\frac{|I|}{2}\right)^2\right) c_5 N^{-1} \|u_N + y_N''\|_0 \cdot \|E^y\|_1, \end{aligned} \quad (65)$$

which is equivalent to

$$\|y_N - y(u_N)\|_1 \leq \left(1 + \left(\frac{|I|}{2}\right)^2\right) c_5 N^{-1} \|u_N + y_N''\|_0. \quad (66)$$

Hence

$$\|y_N - y(u_N)\|_1 + \|p_N - p(u_N)\|_1 \leq c_6 \eta, \quad (67)$$

where

$$c_6 = \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2 c_5, \quad (68)$$

$$\eta = N^{-1} \|y_N - y_d + \lambda_N + p_N''\|_0 + N^{-1} \|u_N + y_N''\|_0.$$

Combining the above analyses, we get that

$$\begin{aligned} & \|u - u_N\|_0 + \|y - y_N\|_1 + \|p - p_N\|_1 + \|\lambda - \lambda_N\|_0 \\ & \leq \|u - u_N\|_0 + \|y - y(u_N)\|_1 + \|y_N - y(u_N)\|_1 \\ & \quad + \|p - p(u_N)\|_1 + \|p_N - p(u_N)\|_1 + \|\lambda - \lambda_N\|_0 \\ & \leq \|y_N - y(u_N)\|_1 + \|p_N - p(u_N)\|_1 \\ & \quad + c_4 \{\|p_N - p(u_N)\|_0 + \|y_N - y(u_N)\|_0\} \\ & \quad + c_1 c_4 \{\|p_N - p(u_N)\|_0 + \|y_N - y(u_N)\|_0\} \\ & \quad + c_2 \{\|p_N - p(u_N)\|_0 \\ & \quad \quad + c_4 (\|p_N - p(u_N)\|_0 + \|y_N - y(u_N)\|_0)\} \\ & \quad + c_3 \{\|p_N - p(u_N)\|_{0,I} \\ & \quad \quad + c_4 (\|p_N - p(u_N)\|_0 + \|y_N - y(u_N)\|_0)\} \\ & \leq \|p_N - p(u_N)\|_1 \{c_4 + 1 + c_1 c_4 + c_2 (c_4 + 1) + c_3 (c_4 + 1)\} \\ & \quad + \|y_N - y(u_N)\|_1 \{c_4 + 1 + c_1 c_4 + c_2 c_4 + c_3 c_4\} \\ & \leq \{c_4 + 1 + c_1 c_4 + c_2 (c_4 + 1) + c_3 (c_4 + 1)\} \\ & \quad \times \{\|p_N - p(u_N)\|_1 + \|y_N - y(u_N)\|_1\} \\ & \leq \{c_4 + 1 + c_1 c_4 + c_2 (c_4 + 1) + c_3 (c_4 + 1)\} c_6 \eta, \end{aligned} \quad (69)$$

which means that

$$\|u - u_N\|_0 + \|y - y_N\|_1 + \|p - p_N\|_1 + \|\lambda - \lambda_N\|_0 \leq C\eta, \quad (70)$$

where

$$\begin{aligned} C &= \{1 + c_4 + c_1 c_4 + c_2 (c_4 + 1) + c_3 (c_4 + 1)\} c_6, \\ c_1 &= \left(1 + \left(\frac{|I|}{2}\right)^2\right)^{1/2} \frac{|I|}{2}, \\ c_2 &= \sqrt{2 \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2 \max \left\{ \frac{2C_\varphi^2 \alpha^2}{|I|} + c_1^2, \frac{2C_\varphi^2}{|I|} \right\}}, \\ c_3 &= \sqrt{\frac{C_\varphi^2}{|I|} \{2c_2^2 + c_1^2\}}, \\ c_4 &= \sqrt{2c_3^2 + 1}, \\ c_5 &= \sqrt{2}, \\ c_6 &= \left(1 + \left(\frac{|I|}{2}\right)^2\right)^2. \end{aligned} \quad (71)$$

#### 4. Conclusions

This paper discusses the explicit formulae of constants within upper bound of the a posteriori error estimate for optimal control problems with Legendre-Galerkin spectral methods in one dimension. Thus, with those formulae, it is easy to choose a suitable degree of polynomials to obtain an acceptable accuracy. In the future, we will study the corresponding constants in lower bound of the a posteriori error indicator. Meanwhile, the corresponding constants in a two-dimensional domain will be investigated.

#### Conflict of Interests

The author declares that there is no conflict of interests regarding the publication of this paper.

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