

# THE APPLICATION OF INTEGRAL EQUATIONS TO THE NUMERICAL SOLUTION OF NONLINEAR SINGULAR PERTURBATION PROBLEMS\*

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## Abstract

The nonlinear singular perturbation problem is solved numerically on non-equidistant meshes which are dense in the boundary layers. The method presented is based on the numerical solution of integral equations [1]. The fourth order uniform accuracy of the scheme is proved. A numerical experiment demonstrates the effectiveness of the method.

## 1. A Continuous Problem

We consider the following singularly perturbed boundary value problem:

$$\varepsilon^2 \frac{d^2 u}{dx^2} = f(x, y), \quad x \in I = [0, 1], \quad u(0) = u(1) = 0, \quad (1)$$

where  $\varepsilon$  is a small positive parameter. We assume that

$$\begin{aligned} f \in C^4(I \times R), \quad g(x) \leq f_u(x, u) \leq G(x), \quad (x, u) \in I \times R, \\ \min\{5g(x) - 2G(x) : x \in I\} > 0, \quad 0 < r^2 < g(x), \quad |g'(x)| \leq L, \\ |G'(x)| \leq L, \quad x \in I. \end{aligned} \quad (2)$$

According to [2], we can prove

**Lemma 1.** Suppose that condition (2) is satisfied. There exists a unique solution  $u \in C^6(I)$  to problem (1), and the following representation holds:  $u(x) = u_0(x) + V_0(x) + V_1(x)$ , where  $V_0(x) = M \exp\left(-\gamma \frac{x}{\varepsilon}\right)$ ,  $V_1(x) = M \exp\left(-\gamma \frac{(1-x)}{\varepsilon}\right)$ , and  $|u_0^{(i)}(x)| \leq M$ ,  $i = 0, 1, \dots, 6$ ,  $x \in I$  (Throughout the paper  $M$  denotes any constant independent of  $\varepsilon$ ).

The proof of the following lemma is based on the monotonicity of (1), and can be found in [3,4].

**Lemma 2.** Let (2) be satisfied. Then, for the solution  $u \in C^6(I)$  to problem (1) there holds, for  $i = 0, 1, \dots, 6$ ,

$$|u^{(i)}(x)| \leq \begin{cases} M \left(1 + \varepsilon^{-i} \exp\left(\gamma \frac{x}{\varepsilon}\right)\right), & 0 \leq x \leq \frac{1}{2}, \\ M \left(1 + \varepsilon^{-i} \exp\left(\gamma \frac{(1-x)}{\varepsilon}\right)\right), & \frac{1}{2} \leq x \leq 1. \end{cases}$$

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## 2. An Equivalent Integral Equation Problem

Introduce the non-equidistant mesh  $I_h = \{x_i\}$ ,  $0 = x_0 < x_1 < \dots < x_n = 1$ ,  $h_i = x_i - x_{i-1}$ ,  $i = 1, 2, \dots, n$ . In the subinterval  $[x_{i-1}, x_{i+1}]$ , we consider

$$\epsilon^2 \frac{d^2 u}{dx^2} = f(x, u), \quad u(x_{i-1}) = A, \quad \frac{du(x_{i-1})}{dx} = B.$$

Integrating once, we have

$$\epsilon^2 u' = \epsilon^2 B + \int_{x_{i-1}}^x f(t, u(t))dt;$$

hence

$$\begin{aligned} \epsilon^2 u(x) &= \epsilon^2 A + \epsilon^2 B(x - x_{i-1}) + \int_{x_{i-1}}^x dt \int_{x_{i-1}}^t f(s, u(s))ds \\ &= \epsilon^2 A + \epsilon^2 B(x - x_{i-1}) + \int_{x_{i-1}}^x (x - t)f(t, u(t))dt. \end{aligned}$$

Now if we require that  $u(x_{i-1}) = u_{i-1}$ ,  $u(x_{i+1}) = u_{i+1}$ , we have  $u_{i-1} = u(x_{i-1}) = A$ ,  $u_{i+1} = u(x_{i+1}) = A + B(x_{i+1} - x_{i-1}) + \frac{1}{\epsilon^2} \int_{x_{i-1}}^{x_{i+1}} (x_{i+1} - t)f(t, u(t))dt$ . Solving for  $A$  and  $B$ , we find that  $u(x)$  satisfies the integral equation

$$\begin{aligned} \epsilon^2 u(x) &= \epsilon^2 u_{i-1} + \epsilon^2 (x - x_{i-1}) \frac{u_{i+1} - u_{i-1}}{x_{i+1} - x_{i-1}} + \int_{x_{i-1}}^x (x - t)f(t, u(t))dt \\ &\quad - \frac{x - x_{i-1}}{x_{i+1} - x_{i-1}} \int_{x_{i-1}}^{x_{i+1}} (x_{i+1} - t)f(t, u(t))dt. \end{aligned}$$

which can be rewritten in the form

$$\epsilon^2 u(x) = \epsilon^2 \frac{x_{i+1} - x}{x_{i+1} - x_{i-1}} u(x_{i-1}) + \epsilon^2 \frac{x - x_{i-1}}{x_{i+1} - x_{i-1}} u(x_{i+1}) - \int_{x_{i-1}}^{x_{i+1}} K(x, t)f(t, u(t))dt, \quad (3)$$

where

$$K(x, t) = \begin{cases} (t - x_{i-1}) \frac{x_{i+1} - x}{x_{i+1} - x_{i-1}}, & x_{i-1} \leq t \leq x, \\ (x - x_{i-1}) \frac{x_{i+1} - t}{x_{i+1} - x_{i-1}}, & x \leq t \leq x_{i+1}. \end{cases}$$

The kernel is then Green's function for the problem, in the notation of classical mechanics.

## 3. Discretization

Letting  $x = x_i$  in (3), we obtain an exact three-point difference scheme:

$$\epsilon^2 u(x_i) + \epsilon^2 \frac{x_{i+1} - x_i}{x_{i+1} - x_{i-1}} u(x_{i-1}) + \epsilon^2 \frac{x_i - x_{i-1}}{x_{i+1} - x_{i-1}} u(x_{i+1}) - \int_{x_{i-1}}^{x_{i+1}} K(x_i, t)f(t, u(t))dt.$$

We denote

$$Nu(x_i) \equiv \epsilon^2 [A_{i-1}u(x_{i-1}) + A_i u(x_i) + A_{i+1}u(x_{i+1})] + \int_{x_{i-1}}^{x_{i+1}} K(x_i, t)f(t, u(t))dt = 0, \quad (4)$$

where

$$A_{i-1} = \frac{-h_{i+1}}{h_i + h_{i+1}}, \quad A_i = 1, \quad A_{i+1} = \frac{-h_i}{h_i + h_{i+1}}. \quad (5)$$

Let  $If = \int_{x_{i-1}}^{x_{i+1}} K(x_i, t) f(t, u(t)) dt$ , and  $P_2$  be the quadric algebra interpolation of  $f(t, u(t))$ ,

$$\begin{aligned} P_2(t) &= \frac{(t - x_i)(t - x_{i+1})}{(x_{i-1} - x_i)(x_{i-1} - x_{i+1})} f_{i-1} + \frac{(t - x_{i-1})(t - x_{i+1})}{(x_i - x_{i-1})(x_i - x_{i+1})} f_i \\ &\quad + \frac{(t - x_{i-1})(t - x_i)}{(x_{i+1} - x_{i-1})(x_{i+1} - x_i)} f_{i+1}, \end{aligned}$$

where  $f_j = f(x_j, u(x_j))$ ,  $j = i-1, i, i+1$ .

In  $If$ , approximating  $f(t, u(t))$  by  $P_2(t)$ , we get  $Qf$ , the approximation of  $If$ :

$$\begin{aligned} Qf &= f_{i-1} \int_{x_{i-1}}^{x_{i+1}} K(x_i, t) \frac{(t - x_i)(t - x_{i+1})}{(x_{i-1} - x_i)(x_{i-1} - x_{i+1})} dt \\ &\quad + f_i \int_{x_{i-1}}^{x_{i+1}} K(x_i, t) \frac{(t - x_{i-1})(t - x_{i+1})}{(x_i - x_{i-1})(x_i - x_{i+1})} dt \\ &\quad + f_{i+1} \int_{x_{i-1}}^{x_{i+1}} K(x_i, t) \frac{(t - x_{i-1})(t - x_i)}{(x_{i+1} - x_{i-1})(x_{i+1} - x_i)} dt \\ &= B_{i-1} f_{i-1} + B_i f_i + B_{i+1} f_{i+1}, \end{aligned}$$

where

$$\left. \begin{aligned} B_{i-1} &= \int_{x_{i-1}}^{x_{i+1}} K(x_i, t) \frac{(t - x_i)(t - x_{i+1})}{(x_{i-1} - x_i)(x_{i-1} - x_{i+1})} dt \\ &= \int_{x_{i-1}}^{x_i} \frac{x_{i+1} - x_i}{x_{i+1} - x_{i-1}} (t - x_{i-1}) \frac{(t - x_i)(t - x_{i+1})}{(x_{i-1} - x_i)(x_{i-1} - x_{i+1})} dt \\ &\quad + \int_{x_i}^{x_{i+1}} \frac{x_i - x_{i-1}}{x_{i+1} - x_{i-1}} (x_{i+1} - t) \frac{(t - x_i)(t - x_{i+1})}{(x_{i-1} - x_i)(x_{i-1} - x_{i+1})} dt \\ &= \frac{h_{i+1}}{h_i(h_i + h_{i+1})} \left( \frac{h_i^4}{12} + \frac{h_i^3 h_{i+1}}{6} \right) - \frac{1}{12} \frac{h_{i+1}^4}{(h_i + h_{i+1})^2} \\ &= \frac{h_{i+1}(h_i^2 - h_{i+1}^2 + h_i h_{i+1})}{12(h_i + h_{i+1})}. \end{aligned} \right\} \quad (6)$$

Similarly,  $B_i = \frac{1}{12}(h_i^2 + h_{i+1}^2 + 3h_i h_{i+1})$ ,  $B_{i+1} = \frac{h_i(h_{i+1}^2 - h_i^2 + h_i h_{i+1})}{12(h_i + h_{i+1})}$ .

Thus, in mesh  $I_h$ , we get the difference scheme of problem (1):

$$\left\{ \begin{array}{l} N_h u_i = \varepsilon^2 (A_{i-1} u_{i-1} + A_i u_i + A_{i+1} u_{i+1} + B_{i-1} f_{i-1} + B_i f_i + B_{i+1} f_{i+1}) = 0, \\ \quad i = 1, 2, \dots, n-1, \\ N_h u_0 \equiv u_0 = 0, \quad N_h u_n \equiv 0. \end{array} \right. \quad (7)$$

In order to get the uniform fourth order difference scheme for problem (1), we take  $x_i = \lambda(t_i)$ ,  $t_i = ih$ ,  $i = 0, 1, \dots, n$ ,  $h = \frac{1}{n}$ ,  $n = 2m$ ,  $m \in N$ .

$$\lambda(t) = \begin{cases} \omega(t) = \frac{\alpha\epsilon t}{q-t}, & t \in [0, t_k], \\ \omega'(t_k)(t - t_k) + \omega(t_k), & t \in [t_k, \frac{1}{2}] \text{ for some } k \in \{1, 2, \dots, m-1\}, \\ 1 - \lambda(1-t), & t \in [\frac{1}{2}, 1], \end{cases} \quad (8)$$

where  $q = t_k + \sqrt{\epsilon}$ ,  $\alpha = \frac{1}{2}[q(\frac{1}{2} - t_k) + \sqrt{\epsilon}t_k]^{-1}$ . It is easy to see  $\lambda(\frac{1}{2}) = \frac{1}{2}$ .

We have  $\lambda : I \rightarrow I$ ,  $\lambda \in C(I)$ ,  $\lambda' \in C^1[1, \frac{1}{2}]$ ,  $\lambda \in C^\infty[0, t_k]$ ,  $\lambda \in C^\infty[t_k, \frac{1}{2}]$ .

Because of symmetry, we consider only  $\lambda(t)$  in  $[0, \frac{1}{2}]$ . When  $t \in [t_k, \frac{1}{2}]$ ,  $\lambda(t)$  is a linear function, so an equidistant mesh is developed, i.e.  $h_i = x_i - x_{i-1} = \omega'(t_k)h = \tilde{h}$ , and  $A_{i-1} = A_{i+1} = -\frac{1}{2}$ ,  $A_i = 1$ ,  $B_{i-1} = B_{i+1} = \frac{1}{24}\tilde{h}^2$ ,  $B_i = \frac{5}{12}\tilde{h}^2$ . When  $t \in [0, t_k]$ , we can prove that  $B_i \geq \frac{5h_i h_{i+1}}{12}$ ,  $B_{i+1} \geq B_{i-1}$ ,  $\frac{h_i h_{i+1}}{24} \leq B_{i+1} \leq \frac{h_i h_{i+1}}{12}$ ,  $B_{i-1} \geq -\frac{h_i h_{i+1}}{12}$ .

**Lemma 3.** If  $x_i = \lambda(t_i)$ ,  $t_i = ih$ ,  $i = 0, 1, \dots, m$ ,  $h = \frac{1}{n}$ ,  $n = 2m$ ,  $A_{i-1}$ ,  $A_i$ ,  $A_{i+1}$  are given by (5),  $B_{i-1}$ ,  $B_i$ ,  $B_{i+1}$  are given by (6), then

- (i)  $A_{i-1} + A_i + A_{i+1} = 0$ ,  $B_{i-1} + B_i + B_{i+1} = \frac{h_i h_{i+1}}{2}$ ,
- (ii)  $B_i \geq \frac{5h_i h_{i+1}}{12}$ ,  $B_{i+1} \geq B_{i-1}$ ,  $\frac{h_i h_{i+1}}{24} \leq B_{i+1} \leq \frac{h_i h_{i+1}}{12}$ ,
- (iii)  $B_{i-1} \geq -\frac{h_i h_{i+1}}{12}$ .

*Proof.* (i) It is easy to verify the results directly. (ii) Since  $\lambda'(t) \geq 0$ ,  $h_{i+1} \geq h_i > 0$ , so  $B_{i+1} - B_{i-1} = \frac{1}{12(h_i + h_{i+1})}(h_{i+1}^3 - h_i^3) \geq 0$ . Thus,  $B_{i+1} \geq B_{i-1}$ . It is not hard to see that

$$\begin{aligned} B_i &= \frac{1}{12}(h_i^2 + h_{i+1}^2 + 3h_i h_{i+1}) = \frac{h_i h_{i+1}}{12} \left( \frac{h_i^2 + h_{i+1}^2}{h_i h_{i+1}} + 3 \right) \geq \frac{5}{12}h_i h_{i+1}, \\ B_{i+1} &= \frac{h_i(h_{i+1}^2 - h_i^2 + h_i h_{i+1})}{12(h_i + h_{i+1})} = \frac{h_i h_{i+1}}{12} \left[ 1 - \frac{h_i^2}{h_{i+1}(h_i + h_{i+1})} \right] \\ &= \frac{h_i h_{i+1}}{12} \left[ 1 - \frac{1}{h_{i+1} \left( 1 + \frac{h_{i+1}}{h_i} \right)} \right] \geq \frac{h_i h_{i+1}}{12} \left[ 1 - \frac{1}{2} \right] = \frac{h_i h_{i+1}}{12}, \end{aligned}$$

$$\text{and } B_{i-1} = \frac{h_i h_{i+1}}{12} \left[ 1 - \frac{h_i^2}{h_i(h_i + h_{i+1})} \right] \leq \frac{h_i h_{i+1}}{12}.$$

$$(iii) B_{i-1} = \frac{h_{i+1}(h_i^2 - h_{i+1}^2 + h_i h_{i+1})}{12(h_i + h_{i+1})} = \frac{h_i h_{i+1}}{12} \left[ 1 - \frac{h_{i+1}^2}{h_i(h_i + h_{i+1})} \right].$$

It is easy to see that  $B_{i-1} \geq \frac{h_i h_{i+1}}{12}$  is equivalent to  $1 - \frac{h_{i+1}^2}{h_i(h_i + h_{i+1})} \geq -1$ , i.e.

$\left(\frac{h_{i+1}}{h_i}\right)^2 - 2\frac{h_{i+1}}{h_i} - 2 \leq 0$ , i.e.  $\frac{h_{i+1}}{h_i} \leq 1 + \sqrt{3} = \beta^2$ . Let  $s = \frac{\beta - 1}{\beta}$ ,  $\beta = \sqrt{1 + \sqrt{3}} \approx 1.653$ ,  $s \approx 0.395$ . Let  $P = t_k + s\sqrt{\epsilon} - \frac{2h}{s} > 0$ . If  $h$  is sufficiently small, then  $P \geq t_k$ .

Let us now consider the case  $i < k+1$ . We have  $t_{i-1} \leq P$ ,  $t_{i+1} = t_{i-1} + 2h \leq P + 2h = t_k + s\sqrt{\epsilon} - \frac{2h}{s} + 2h = t_k + s\sqrt{\epsilon} - 2h\left(\frac{1}{s} - 1\right) < t_k + \sqrt{\epsilon} = q$ . In this case,  $t_{i-1} \leq t_k$ ,  $\frac{\lambda'(t_{i+1})}{\lambda'(t_{i-1})} \leq \frac{\omega'(t_{i+1})}{\omega'(t_{i-1})} = \left(\frac{q - t_{i-1}}{q - t_{i+1}}\right)^2 = \left(\frac{q - t_{i+1} + 2h}{q - t_{i+1}}\right)^2 = \left(1 + \frac{2h}{q - t_{i+1}}\right)^2$ . Because  $q - t_{i+1} = q - t_{i-1} - 2h \geq q - P - 2h = t_k + \sqrt{\epsilon} - t_k - s\sqrt{\epsilon} + \frac{2h}{s} - 2h = (1-s)\sqrt{\epsilon} + \frac{2h}{s}(1-s) \geq \frac{2h}{s}(1-s)$ , so  $\frac{\omega'(t_{i+1})}{\omega'(t_{i-1})} \leq \left[1 + \frac{2h}{(2h/s)(1-s)}\right]^2 = \left(1 - \frac{s}{1-s}\right)^2 = \left(\frac{1}{1-s}\right)^2 = \beta^2$ .

Thus,  $\frac{h_{i+1}}{h_i} = \frac{\lambda'(\xi_{i+1})h}{\lambda'(\xi_i)h} \leq \frac{\lambda'(t_{i+1})}{\lambda'(t_{i-1})}$ , where  $t_i < \xi_{i+1} < t_{i+1}$ ,  $t_{i-1} < \xi_i < t_i$ ,  $\lambda'(t_{i+1}) \geq \lambda'(\xi_{i+1})$ ,  $\lambda'(t_{i-1}) \leq \lambda'(\xi_i)$ . This implies that  $\frac{h_{i+1}}{h_i} \leq \beta^2 = 1 + \sqrt{3}$ .

In the case  $i \geq k+1$ , i.e.  $t_{i-1} \geq t_k$ ,  $x_i = \lambda(t_i)$  is an equidistant mesh. The result holds obviously.

**Theorem 1.** *Let condition (2) hold and let the discrete problem (7) be given on the mesh (8) with sufficiently large  $n$ . Then problem (7) has a unique solution  $V_h = [u_0, u_1, \dots, u_n]^T$ , which is a point of attraction of SOR-Newton and Newton-SOR method for the relaxation parameter in  $(0, 1]$ . Moreover, for any  $V_h^1, V_h^2 \in R^{n+1}$ , the following stability inequality holds:  $\|V_h^1 - V_h^2\|_\infty \leq \sigma^{-1} \|N_h V_h^1 - N_h V_h^2\|_\infty$ , where  $\sigma$  is a positive constant, independent of  $\epsilon$ .*

*Proof.* The Frechet derivative  $N_h^1(V)$  of  $N_h$  for arbitrary  $V = [V_0, V_1, \dots, V_n]^T$  is a tridiagonal matrix

$$N_h^1(V) = \begin{bmatrix} 1 & 0 & 0 & & \\ a_1 & b_1 & c_1 & 0 & \\ & \ddots & \ddots & \ddots & \\ 0 & a_n & b_n & c_n & \\ & 0 & 0 & 1 & \end{bmatrix}$$

where  $a_i = \epsilon^2 A_{i-1} + B_{i-1} f_u(x_{i-1}, V_{i-1})$ ,  $b_i = \epsilon^2 A_i + B_i f_u(x_i, V_i)$ ,  $c_i = \epsilon^2 A_{i+1} + B_{i+1} f_u(x_{i+1}, V_{i+1})$ .

Let  $\sigma = \min_{1 \leq i \leq n} \sigma_i$ ,  $\sigma_i = |b_i| - |a_i| - |c_i|$ . Note that  $B_{i-1} \geq -\frac{h_i h_{i+1}}{12}$ . We now consider two cases:  $B_{i-1} < 0$  and  $B_{i-1} \geq 0$ .

I.  $B_{i-1} < 0$ ,  $a_i < 0$ ,  $b_i > 0$ .

i)  $c_i \leq 0$ ,  $\sigma_i = |b_i| - |a_i| - |c_i| = b_i + a_i + c_i = \epsilon^2(A_{i-1} + A_i + A_{i+1}) + B_{i-1} f_u(x_{i-1}, V_{i-1}) + B_i f_u(x_i, V_i) + B_{i+1} f_u(x_{i+1}, V_{i+1}) \geq B_i g(x_i) + B_{i+1} g(x_{i+1}) + B_{i-1} G(x_{i-1}) = (B_i + B_{i+1})g(x_i) + B_{i+1}(g(x_{i+1}) - g(x_i)) + B_{i-1}G(x_{i-1}) - G(x_i))$ .

According to Lemma 3, for the coefficients  $A_j$ ,  $B_j$  ( $j = i-1, i, i+1$ ) we have  $A_{i-1} + A_i + A_{i+1} = 0$ ,  $B_{i-1} + B_i + B_{i+1} = \frac{h_i h_{i+1}}{2}$ ,  $B_{i-1} \geq -\frac{h_i h_{i+1}}{12}$ ,  $\frac{h_i h_{i+1}}{24} \leq B_{i+1} \leq$

$\frac{h_i h_{i+1}}{12}$ . We also have  $h_{i+1} \geq h_i > 0$ ,  $|g'(x)| \leq L$ ,  $|G'(x)| \leq L$ ,  $5g(x) - 2G(x) \geq \delta > 0$ . So  $\frac{2}{h_i h_{i+1}} \sigma_i = \left(1 - \frac{2}{h_i h_{i+1}} B_{i-1}\right) g(x_i) + [B_{i+1}g'(\theta_{i+1})h_{i+1} + B_{i-1}G(x_i) + B_{i-1}G'(\theta_{i-1})h_i] \frac{2}{h_i h_{i+1}}$ , where,  $x_i < \theta_{i+1} < x_{i+1}$ ,  $x_{i-1} < \theta_i < x_i$ . Thus,  $\frac{12}{h_i h_{i+1}} \sigma_i \geq 7g(x_i) - G(x_i) - |g'(\theta_{i+1})|h_{i+1} - |G'(\theta_i)|h_i \geq 5g(x_i) - 2G(x_i) - 2Lh_{i+1} \geq \delta - 2Lh_{i+1}$ .

ii)  $c_i \geq 0$ ,  $\sigma_i = b_i + a_i - c_i = \epsilon^2(A_{i-1} + A_i + A_{i+1}) + B_{i-1}f_u(x_{i-1}, V_{i-1}) + B_i f_u(x_i, V_i) - B_{i+1}f_u(x_{i+1}, V_{i+1}) \geq B_i g(x_i) + B_{i-1}G(x_{i-1}) - B_{i+1}G(x_{i+1}) = B_i g(x_i) + (B_{i-1} - B_{i+1})G(x_i) + B_{i-1}[G(x_{i-1}) - G(x_i)] + B_{i+1}[G(x_i) - G(x_{i+1})] = B_i g(x_i) + (B_{i-1} - B_{i+1})G(x_i) - B_{i-1}|G'(\theta_i)|h_i - B_{i+1}|G'(\theta_{i+1})|h_{i+1}$ , where  $x_{i-1} < \theta_i < x_i$ ,  $x_i < \theta_{i+1} < x_{i+1}$ .

Thus  $\frac{12\sigma_i}{h_i h_{i+1}} \geq 5g(x_i) - 2G(x_i) - |G'(\theta_i)|h_i - |G'(\theta_{i+1})|h_{i+1} \geq \delta - 2Lh_{i+1}$ .

II.  $B_{i-1} \geq 0$ .

i)  $a_i \leq 0$ ,  $c_i \leq 0$ ,  $\sigma_i = b_i + a_i + c_i = \epsilon^2(A_{i-1} + A_i + A_{i+1}) + B_{i-1}f_u(x_{i-1}, V_{i-1}) + B_i f_u(x_i, V_i) - B_{i+1}f_u(x_{i+1}, V_{i+1}) \geq (B_{i-1} + B_i + B_{i+1})r^2 \geq \frac{h_i h_{i+1}}{2} r^2 > 0$ .

Similarly, we can prove that

ii)  $a_i \leq 0$ ,  $c_i \geq 0$ ,  $\frac{2}{h_i h_{i+1}} \sigma_i \geq \frac{\delta}{6} - \frac{Lh_{i+1}}{6}$ . iii)  $a_i \geq 0$ ,  $c_i \leq 0$ ,  $\frac{2}{h_i h_{i+1}} \sigma_i \geq \frac{\delta}{6} - \frac{Lh_{i+1}}{6}$ .

iv)  $a_i \geq 0$ ,  $c_i \geq 0$ ,  $\frac{2}{h_i h_{i+1}} \sigma_i \geq \frac{\delta}{6} - \frac{Lh_{i+1}}{3}$ .

Summarizing all the cases above, we have

$$\frac{2}{h_i h_{i+1}} \sigma_i \geq \min \left\{ r^2, \frac{\delta}{6} - \frac{Lh_{i+1}}{3} \right\}, \quad i = 1, 2, \dots, n.$$

For sufficiently large  $n$ , when  $t_i < t_k$ , we take  $n > \frac{4L\lambda'(t_k)}{\delta}$ . Then  $h_{i+1} = \lambda(t_{i+1}) - \lambda(t_i) = \lambda'(\theta)h$ ,  $\theta \in (t_i, t_{i+1})$ ,  $h = \frac{1}{n} < \frac{\delta}{4L\lambda'(t_k)}$ ,  $\frac{\delta}{6} - \frac{Lh_{i+1}}{3} \geq \frac{\delta}{6} - \frac{\lambda'(\theta)\delta}{12\lambda'(t_k)} \geq \frac{\delta}{12}$ . Thus  $\sigma_i \geq \min \left\{ \frac{h_i h_{i+1}}{2} \gamma^2, \frac{h_i h_{i+1}}{24} \delta \right\} > 0$ ,  $i = 1, 2, \dots, n$ . So  $\sigma = \min_{1 \leq i \leq n} \sigma_i > 0$ .

In the case  $t_k \leq t_i \leq \frac{1}{2}$ ,  $\lambda(t)$  is a linear function.  $\{x_i\}$  is an equidistant mesh.  $h_i = \omega'(t_k)h = \tilde{h}$ .  $B_{i-1} = \frac{\tilde{h}^2}{24}$ ,  $B_i = \frac{5\tilde{h}^2}{12}$ ,  $B_{i+1} = \frac{\tilde{h}^2}{24}$ ,  $A_{i-1} = A_{i+1} = -\frac{1}{2}$ ,  $A_i = 1$ . According to case I, it is easy to see that

$$\sigma = \min_{1 \leq i \leq n} \sigma_i \geq \min \left\{ \gamma^2 \frac{\tilde{h}^2}{2}, \delta \frac{\tilde{h}^2}{24} \right\} > 0.$$

From the results above, we have

$$\|N'_h(V)^{-1}\|_\infty \leq \frac{1}{\sigma}. \tag{9}$$

Now by Hadamard's theorem [5], (7) has a unique solution  $V_h = [u_0, u_1, \dots, u_n]^T$ . The matrix  $N_h^1(V_h)$  is a strictly diagonally dominant matrix and the convergence

of Newton-SOR and SOR-Newton iterative methods for the relaxation parameter in  $(0, 1]$  follows by the well known theorem from [5].

Since  $N_h V_h^1 - N_h V_h^2 = N_h^1(V)(V_h^1 - V_h^2)$ , from (9) it follows that

$$\|V_h^1 - V_h^2\|_\infty \leq \sigma^{-1} \|N_h V_h^1 - N_h V_h^2\|_\infty.$$

Thus, we have proven the theorem.

#### 4. Error Analysis

Let us consider the consistency error

$$r_h = N_h u_h - N_h V_h = N_h u_h$$

where  $V_h = [u_0, u_1, \dots, u_n]^T$  is the solution of problem (7), and  $u_h = [u(x_0), u(x_1), \dots, u(x_n)]^T \in R^{n+1}$  is the restriction of solution  $u(x)$  to problem (1) on the mesh  $I_h$ . The components of the vector  $r_h$  are  $r_0 = r_n = 0$ , and for  $i = 1, 2, \dots, n-1$ ,

$$\begin{aligned} r_i &= N_h u(x) - N_h u_i = N_h u(x_i) = \epsilon^2 (A_{i-1}u(x_{i-1}) + A_i u(x_i) + A_{i+1}u(x_{i+1})) \\ &\quad + B_{i-1}f(x_{i-1}, u(x_{i-1})) + B_i f(x_i, u(x_i)) + B_{i+1}f(x_{i+1}, u(x_{i+1})) = \epsilon^2 (A_{i-1}u(x_{i-1}) \\ &\quad + A_i u(x_i) + A_{i+1}u(x_{i+1})) + \epsilon^2 (B_{i-1}u''(x_i) + B_{i+1}u''(x_{i+1})). \end{aligned} \quad (10)$$

By using the Taylor expansion we have

$$\begin{aligned} u(x_{i-1}) &= u(x_i) - u'(x_i)h_i + \frac{1}{2}u''(x_i)h_i^2 - \frac{1}{3!}u'''(x_i)h_i^3 + \frac{1}{4!}u^{(4)}(x_i)h_i^4 \\ &\quad - \frac{1}{5!}u^{(5)}(x_i)h_i^5 + \frac{1}{6!}u^{(6)}(\xi_i^1)h_i^6, \\ u(x_{i+1}) &= u(x_i) + u'(x_i)h_{i+1} + \frac{1}{2}u''(x_i)h_{i+1}^2 + \frac{1}{3!}u'''(x_i)h_{i+1}^3 + \frac{1}{4!}u^{(4)}(x_i)h_{i+1}^4 \\ &\quad + \frac{1}{5!}u^{(5)}(x_i)h_{i+1}^5 + \frac{1}{6!}u^{(6)}(\xi_i^2)h_{i+1}^6, \\ u''(x_{i-1}) &= u''(x_i) - u'''(x_i)h_i + \frac{1}{2}u^{(4)}(x_i)h_i^2 - \frac{1}{3!}u^{(5)}(x_i)h_i^3 + \frac{1}{4!}u^{(6)}(\xi_i^3)h_i^4, \\ u''(x_{i+1}) &= u''(x_i) + u'''(x_i)h_{i+1} + \frac{1}{2}u^{(4)}(x_i)h_{i+1}^2 + \frac{1}{3!}u^{(5)}(x_i)h_{i+1}^3 + \frac{1}{4!}u^{(6)}(\xi_i^4)h_{i+1}^4. \end{aligned}$$

Let  $\alpha_s$  denote the coefficient of  $u^{(5)}(x_i)$  in (10),  $s = 0, 1, 2, 3, 4, 5$ , and let  $\alpha_6$  denote the sum of the terms of  $u^{(6)}(\xi_i^j)$ ,  $j = 1, 2, 3, 4$ . We get

$$\alpha_0 = A_{i-1} + A_i + A_{i+1} = 0,$$

$$\alpha_1 = -h_i A_{i-1} + h_{i+1} A_{i+1} = \frac{h_i h_{i+1}}{h_i + h_{i+1}} - \frac{h_i h_{i+1}}{h_i + h_{i+1}} = 0,$$

$$\begin{aligned} \alpha_2 &= \frac{1}{2}h_i^2 A_{i-1} + \frac{1}{2}h_{i+1}^2 A_{i+1} + B_{i-1} + B_i + B_{i+1} = \frac{-h_i^2 h_{i+1}}{2(h_i + h_{i+1})} + \frac{-h_i h_{i+1}^2}{2(h_i + h_{i+1})} \\ &\quad + \frac{h_i h_{i+1}}{2} = 0, \end{aligned}$$

$$\begin{aligned}
\alpha_3 &= \frac{-1}{3!} h_i^3 A_{i-1} + \frac{1}{3!} h_{i+1}^3 A_{i+1} - h_i B_{i-1} + h_{i+1} B_{i+1} = 0, \\
\alpha_4 &= \frac{1}{4!} h_i^4 A_{i-1} + \frac{1}{4!} h_{i+1}^4 A_{i+1} + \frac{1}{2} h_i^2 B_{i-1} + \frac{1}{2} h_{i+1}^2 B_{i+1} = 0, \\
\alpha_5 &= \frac{-1}{5!} h_i^5 A_{i-1} + \frac{1}{5!} h_{i+1}^5 A_{i+1} - \frac{1}{3!} h_i^3 B_{i-1} + \frac{1}{3!} h_{i+1}^3 B_{i+1} = 0, \\
\alpha_6 &= \frac{1}{6!} u^{(6)}(\xi_i^1) h_i^6 A_{i-1} + \frac{1}{6!} u^{(6)}(\xi_i^2) h_{i+1}^6 A_{i+1} + \frac{1}{4!} u^{(6)}(\xi_i^3) h_i^4 B_{i-1} + \frac{1}{4!} u^{(6)}(\xi_i^4) h_{i+1}^4 B_{i+1} \\
&= \frac{-h_i h_{i+1}}{6!(h_i + h_{i+1})} [u^{(6)}(\xi_i^1) h_i^5 + u^{(6)}(\xi_i^2) h_{i+1}^5] + \frac{u^{(6)}(\xi_i^3) h_i^4 h_{i+1} (h_i^2 - h_{i+1}^2 + h_i h_{i+1})}{12 \times 4!(h_i + h_{i+1})} \\
&\quad + \frac{u^{(6)}(\xi_i^4) h_i h_{i+1}^4 (h_{i+1}^2 - h_i^2 + h_i h_{i+1})}{12 \times 4!(h_i + h_{i+1})},
\end{aligned}$$

where  $\xi_i^j \in (x_{i-1}, x_{i+1})$ ,  $j = 1, 2, 3, 4$ .

Thus, we obtain

$$r_i = \varepsilon^2 \alpha_6.$$

According to Lemma 2,

$$|u^{(i)}(x)| \leq \begin{cases} M[1 + \varepsilon^{-i} \exp(-\gamma \frac{x}{\varepsilon})], & 0 \leq x \leq \frac{1}{2}, \\ M[1 + \varepsilon^{-i} \exp(-\gamma \frac{(1-x)}{\varepsilon})], & \frac{1}{2} \leq x \leq 1. \end{cases}$$

If  $t \in [0, t_k]$ ,  $\exp(-\gamma \lambda(t)/\varepsilon) = \exp(-\gamma \frac{\alpha \varepsilon t}{(q-t)\varepsilon}) = \exp(\gamma \alpha - \frac{\gamma \alpha q}{q-t}) \leq M \exp(-\frac{M}{q-t})$ ;

thus,  $|u^{(6)}(\xi_i^j)| \leq M[1 + \varepsilon^{-6} M \exp(-\frac{M}{q-t_{i-1}})]$ ,  $j = 1, 2, 3, 4$ .

Note that

$$h_i = \lambda(t_i) - \lambda(t_{i-1}) = \lambda'(t_{i-1} + \theta_1 h)h = \frac{\alpha \varepsilon q h}{(q - t_{i-1} - \theta_1 h)^2} \leq \frac{\alpha \varepsilon q}{(q - t_{i+1})^2} h,$$

$$h_{i+1} = \lambda(t_{i+1}) - \lambda(t_i) = \lambda'(t_i + \theta_2 h)h = \frac{\alpha \varepsilon q h}{(q - t_i - \theta_2 h)^2} \leq \frac{\alpha \varepsilon q}{(q - t_{i+1})^2} h,$$

$$0 < \theta_1, \theta_2 < 1, \quad t_{i-1} < t_k \leq p = t_k + s\sqrt{\varepsilon} - \frac{2h}{s} < q - 3h, \quad (s = 0.395),$$

$$\frac{q - t_{i+1}}{q - t_{i-1}} = \frac{q - t_{i-1} - 2h}{q - t_{i-1}} = 1 - \frac{2h}{q - t_{i-1}} \geq 1 - \frac{2h}{3h} = \frac{1}{3}, \quad \frac{1}{q - t_{i+1}} \leq \frac{3}{q - t_{i-1}}.$$

Thus,  $|r_i| = |N_h u(x_i) - N_h u_i| \leq M \varepsilon^2 \left( \frac{\alpha \varepsilon q}{(q - t_{i+1})^2} \right)^6 h^6 [1 + \varepsilon^{-6} M \exp(-M/(q - t_{i-1}))] \leq M \varepsilon^2 \left( \frac{9 \alpha q \varepsilon}{(q - t_{i-1})^2} \right)^6 h^6 [1 + \varepsilon^{-6} M \exp(-M/(q - t_{i-1}))] \leq M \varepsilon^2 h^6 (q - t_{i-1})^{-12} \exp(-M/(q - t_{i-1})) \leq M h^6$  ( $M$ , a constant independent of  $\varepsilon$  and  $h$ , may take different values).

In the case of  $t \in [t_k, \frac{1}{2}]$ , the mesh is equidistant and is far from the boundary layers. Note that  $x \geq \lambda(t_k) = \omega(t_k) = \alpha \sqrt{\varepsilon} t_k$ ,  $\varepsilon^{-i} \exp(-\gamma \frac{x}{\varepsilon}) \leq \varepsilon^{-i} \exp(-\gamma \frac{\alpha t_k}{\sqrt{\varepsilon}}) \leq M$ . Analogously, it is easy to prove  $|r_i| \leq M h^6$ . Thus we get

**Theorem 2.** Let condition (2) hold. On the mesh (8), we have, for  $i = 1, 2, \dots, n$ ,  $|r_i| = |N_h u(x_i) - N_h u_i| \leq M h^6$ , where  $u(x_i)$  are the same as in Theorem 1, and  $M$  is independent of  $h$  and  $\varepsilon$ .

Summarizing the results above, we have the main result of this paper.

**Theorem 3.** Let condition (2) be satisfied, then  $\|u_h - V_h\|_\infty \leq M h^4$ , where  $u_h = [u(x_0), u(x_1), \dots, u(x_n)]^T \in R^{n+1}$  is the restriction of solution  $u(x)$  of problem (1) on mesh (8),  $V_h = [u_0, u_1, \dots, u_n]^T$  is the solution of problem (7), and  $M$  is independent of  $h$  and  $\varepsilon$ .

*Proof.* By using Theorem 2 we have  $|N_h u(x_i) - N_h u_i| \leq M h^6$ , where  $M$  is independent of  $h$  and  $\varepsilon$ . By using Theorem 1 we have  $\|u_h - V_h\|_\infty \leq \sigma^{-1} \|N_h u_h - N_h V_h\|_\infty$ , where  $\sigma$  is a positive constant independent of  $\varepsilon$  and  $\sigma^{-1} = O(h^{-2})$ . So,  $\|u_h - V_h\|_\infty \leq M h^4$ , where  $M$  is independent of  $h$  and  $\varepsilon$ . Thus, we have proven the theorem.

## 5. A Numerical Example

Consider the following example:

$$\begin{cases} -\varepsilon^2 u'' + u - x^2 + x + 1 + 2\varepsilon^2 = 0, \\ u(0) = u(1) = 0, \end{cases}$$

whose solution is  $u(x) = \frac{\exp(-\frac{x}{\varepsilon}) + \exp(\frac{x-1}{\varepsilon})}{1 + \exp(-\frac{1}{\varepsilon})} + x^2 - x - 1$ .

In the table we give the error  $E = \|u_h - V_h\|_\infty$ . Different values of  $\varepsilon$  and  $h$  are taken.

$n \backslash \varepsilon$	$2^{-2}$	$2^{-4}$	$2^{-8}$	$2^{-16}$
8	5.27690	5.46145(-2)	5.48783(-2)	5.48565(-2)
16	4.11203(-2)	9.29340(-3)	9.27858(-3)	9.29014(-3)
32	3.42830(-4)	2.71855(-3)	2.71941(-3)	2.71940(-3)
64	2.24534(-5)	1.38178(-4)	1.38218(-4)	1.38202(-4)
128	1.29380(-6)	9.00482(-6)	9.01035(-6)	9.01040(-6)
256	9.21414(-8)	5.48899(-7)	5.48889(-7)	5.48890(-7)

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## References

- [1] L. M., Delevie and J. L. MoHamed, Computational Methods for Integral Equations, Cambridge University Press, 1985.
- [2] R. Vulanovic, D. Herceg and N. Petrovic, On the extrapolation for a singularly perturbed boundary value problem, *Computing*, **36** (1986), 69–79.

- [3] J. Lorenz, Zur Theorie und Numerik Von Differenzenverfahren fur Singulare Storungen, Habilitationsschrift, Konstanz, 1980.
- [4] G. I. Shishkin, Raznostnaya skhema na neravnomoernoisetke dlya differential nogo uravneniya s malym parametrom pri starshei proizvodnot. Zh, Vychisl. Mat. Mat. Fiz., **23** (1983), 609–619.
- [5] J. M. Ortega and W. C. Rheinboldt, Iterative Solution of Nonlinear Equations in Several Variables, Academic Press, New York, London, 1st Ed. 1970.