

# ROBUST ESTIMATES IN BALANCED NORMS FOR SINGULARLY PERTURBED REACTION DIFFUSION EQUATIONS USING GRADED MESHES

MARÍA GABRIELA ARMENTANO, ARIEL L. LOMBARDI,  
AND CECILIA PENESSI

ABSTRACT. The goal of this paper is to provide almost robust approximations of singularly perturbed reaction-diffusion equations in two dimensions by using finite elements on graded meshes. When the mesh grading parameter is appropriately chosen, we obtain quasioptimal error estimations in a balanced norm for piecewise bilinear elements, by using a weighted variational formulation of the problem introduced by N. Madden and M. Stynes, *Calcolo* 58(2) 2021. We also prove a supercloseness result, namely, that the difference between the finite element solution and the Lagrange interpolation of the exact solution, in the weighted balanced norm, is of higher order than the error itself. We finish the work with numerical examples which show the good performance of our approach.

## 1. INTRODUCTION

The reaction-diffusion equations arise in many applications, indeed, these equations appear naturally in systems consisting of many interacting components and are used to describe pattern-formation phenomena of biological, chemical and physical systems (see, for example, [8, 18, 19]).

It is well known that, when the singular perturbation parameter is very small, the solution of the problem presents boundary layers which downgrade the approximability of the solution when uniform or quasi-uniform meshes are used. The approximation by finite element methods of these singularly perturbed problems have been extensively studied (see, for instance, [21, 13, 10, 14] and its references) where uniform error estimates were analyzed for different norms, including the energy and  $L^\infty$  ones.

It turns out that the natural energy norm associated to the problem is not balanced, i.e, when the singular perturbation parameter tends to zero, the energy norm of the layer contribution vanishes while the energy norm

---

*2020 Mathematics Subject Classification.* 65N30, 65N15.

*Key words and phrases.* reaction diffusion problems, singularly perturbed problems, balanced norms, graded meshes, supercloseness.

of the smooth part of the solution does not. Balanced norms were introduced to reflect the behavior of layers more accurately in the finite element method for singularly perturbed reaction–diffusion problems. This is extensively discussed in [12] where a new bilinear form and a finite element method were designed to facilitate the analysis for a new balanced norm. Subsequently new analysis were performed in several articles, in particular [1, 2, 7, 17].

Therefore, the problem requires especially designed schemes for its effective numerical solution. In a recent work, N. Madden and M. Stynes [16] introduced a weighted balanced norm (whose  $H^1$  component is scaled to the correct size) and obtained an robust almost first-order error bound for piecewise bilinears on the unit square by using Shishkin meshes.

In this paper we consider the bilinear formulation and the weighted balanced norm introduced in [16], and obtain a robust approximation of singularly perturbed reaction-diffusion equation, with homogeneous Dirichlet boundary conditions, in two dimensions by using piecewise bilinear elements on graded meshes. We present quasi-optimal error estimates when appropriate graded meshes are used, in addition we also obtain a supercloseness result for the balanced norm, i.e., we prove that, under suitable hypothesis, the difference between the approximate solution and the Lagrange interpolation of the exact solution is of higher order than the error itself. In particular, to obtain the supercloseness result we need to prove some properties over the weight function which characterize the discrete formulation and also we need to prove some estimations over the derivatives of the solution.

In [6] graded meshes were also used, with bilinear finite elements, to obtain robust and almost optimal error estimates in the energy norm for a reaction diffusion problem similar to the one we consider here. In that work the grading parameter (and therefore the meshes) could be taken independently of the singular perturbation parameter of the equation. Adjusting the grading parameter, but still being independent of the singular perturbation, supercloseness results in the energy norm were obtained in [5]. In the present paper, to obtain almost uniform results in the balanced norm, we use meshes of the same type to those introduced in [6] but with a grading depending on the singular perturbation parameter (see Section 3).

Although the numerical results obtained with Shishkin meshes and graded meshes are similar, graded meshes satisfy some desirable properties. In fact, when one is approximating a singularly perturbed problem with an a priori adapted mesh, it is natural to expect that a mesh designed for some value of the perturbation parameter works well also for larger values of it (we include a numerical test of this performance). This is the case for graded meshes as it is mentioned in [4, 5]. This fact could be an important property

in problems where the diffusion parameter is not constant or, also, to treat systems of equations in which different equations have singular perturbations of different orders.

The paper is organized as follows. In Section 2 we present the reaction diffusion problem, the weighted formulation and the weighted balanced norm under consideration. In Section 3 we introduce the graded meshes and we present some interpolation properties in standard Sobolev norms and in Section 4 we obtain interpolation error estimates on the weighted balanced norm. Section 5 is devoted to the supercloseness results. In Section 6 we present some numerical examples which show the good performance of our method. We finish the paper with an Appendix which includes a technical Lemma used along the paper.

Throughout the paper, the letter  $C$  will denote a generic positive constant, not necessarily the same at each occurrence, which is independent of the singular perturbation parameter  $\varepsilon$  and the mesh size.

## 2. PROBLEM STATEMENT

Let  $\Omega$  be a bounded domain on  $\mathbb{R}^2$  and  $\partial\Omega$  its boundary. We consider the following reaction-diffusion problem

$$(2.1) \quad \begin{aligned} -\varepsilon^2 \Delta u + b(x, y)u &= f(x, y) & (x, y) \in \Omega \\ u &= 0 & \text{on } \partial\Omega \end{aligned}$$

where  $0 < \varepsilon < 1$  and  $b \in L^\infty(\Omega)$ , with  $0 < b_0^2 < b(x, y) < b_1^2$  for almost all  $(x, y) \in \Omega$ .

In a recent paper, Madden and Stynes [16] propose a new variational formulation of this problem as follows. Let

$$\beta(x, y) = 1 + \frac{1}{\varepsilon} e^{-\frac{\gamma d(x, y)}{\varepsilon}}$$

be a weighting function, with  $\gamma$  a fixed positive parameter and  $d(x, y)$  the distance to the boundary  $\partial\Omega$ . It is appropriate to mention that, also this weight function is basically the same used in Adler et al. [2], but there the authors rewrite the reaction-diffusion problem as a system of equations. The property (see [16])

$$|\nabla\beta(x, y)| \leq \frac{C}{\varepsilon} \beta(x, y)$$

almost everywhere in  $(x, y) \in \Omega$  will be used along the manuscript.

We consider the weighted norm

$$\|v\|_\beta = \left( \varepsilon^2 \|\nabla v\|_\beta^2 + \|v\|_\beta^2 \right)^{\frac{1}{2}}$$

where  $\|v\|_\beta = (\beta v, v)^{\frac{1}{2}}$ . We use the notation  $|||\cdot|||_{\beta,D}, \|\cdot\|_{\beta,D}$  to denote the  $\beta$ -weighted norms on the subdomain  $D$ . The domain subscript is dropped for the case  $D = \Omega$ .

Defining the weighted bilinear form  $B_\beta : H_0^1(\Omega)^2 \rightarrow \mathbb{R}$  by

$$B_\beta(v, w) = \varepsilon^2 \int_\Omega \nabla v \cdot \nabla(\beta w) dx dy + \int_\Omega b(x)v(\beta w) dx dy.$$

Then, the variational formulation of problem (2.1) is given by: find  $u \in H_0^1(\Omega)$  such that

$$B_\beta(u, v) = \int_\Omega f(x)(\beta v) dx dy \quad \forall v \in H_0^1(\Omega).$$

**Remark 2.1.** *The  $\beta$ -norm  $|||\cdot|||_\beta$  is balanced, indeed, its components  $\varepsilon^2 \|\nabla u\|_\beta$  and  $\|u\|_\beta^2$  are both  $O(1)$  for a typical solution  $u$  of (2.1) ( see [16] for more details).*

If  $V_h \subseteq H_0^1(\Omega)$  is a finite element space, we define the finite element formulation: find  $u_h \in V_h$  such that

$$(2.2) \quad B_\beta(u_h, v) = \int_\Omega f(x)(\beta v) dx dy \quad \forall v \in V_h.$$

Following [16] we assume  $0 < \gamma \leq b_0$ . In this case, the bilinear form  $B_\beta(\cdot, \cdot)$  is coercive and continuous, and by using Lax-Milgram Theorem and Céa Lemma, the following approximation error estimate holds (see [16, Section 3]):

$$(2.3) \quad |||u - u_h|||_\beta \leq C \inf_{w_h \in V_h} |||u - w_h|||_\beta.$$

It follows that in order to estimate the error in the balanced norm  $|||\cdot|||_\beta$  is enough to compare  $u$  with some interpolant  $\Pi u$  of  $u$ .

### 3. GRADED MESHES AND PRELIMINARY RESULTS

Let  $\Omega = (0, 1)^2$ . Let us introduce a family of meshes in the following way. We consider two parameters,  $h > 0$  which is related with the mesh size (see Remark 3.1), and the grading parameter  $\alpha$  given by

$$\alpha := 1 - \frac{1}{2 \log \frac{1}{\varepsilon}}.$$

Let  $x_0, x_1, \dots, x_{mid}$  the grid points on the interval  $[0, \frac{1}{2}]$  given by

$$(3.4) \quad \begin{cases} x_0 = 0, \\ x_1 = h^s, & \text{with } s := \frac{1}{1-\alpha} \\ x_{i+1} = x_i + hx_i^\alpha, & i = 2, \dots, mid-1, \\ x_{mid} = \frac{1}{2}. \end{cases}$$

This partition is extended to a grid  $\{x_0, x_1, \dots, x_{mid}, \dots, x_M\}$  with  $M = 2mid$  of  $[0, 1]$  by setting  $x_i = 1 - x_{M-i}$  for  $i = mid + 1, \dots, M$ . We consider a 2-dimensional mesh  $\mathcal{T}_h = \{R\}$  of tensor product type of  $\Omega = (0, 1)^2$ , composed by rectangles  $R = R_{ij}$  defined by

$$R_{ij} = (x_{i-1}, x_i) \times (x_{j-1}, x_j).$$

Set  $h_k = x_k - x_{k-1}$ . Then the lengths of the sides of  $R_{ij}$  are  $h_i$  and  $h_j$ . We will use repeatedly along this paper the following property for the meshes  $\mathcal{T}_h$ : For  $R_{ij} \in \mathcal{T}_h$  with  $1 < i < M$  we have

$$h_i \leq h \min\{x, 1-x\}^\alpha \quad \forall (x, y) \in R_{ij}.$$

Similarly, for  $R_{ij} \in \mathcal{T}_h$  with  $1 < j < M$  we have

$$h_j \leq h \min\{y, 1-y\}^\alpha \quad \forall (x, y) \in R_{ij}.$$

**Remark 3.1.** *The number  $M + 1$  of grid points along the  $x$  and  $y$  axis is related with the parameter  $h$  which define the mesh  $\mathcal{T}_h$  by*

$$h \leq C \frac{1}{M} \log \frac{1}{\varepsilon} \log M.$$

(see [6, proof of Corollary 4.5]). Hence, we see that  $h$  is bounded almost uniformly with respect to  $\varepsilon$  and similarly to the case of quasi-uniform meshes except for the logarithmic factor  $\log M$ . In what follows, for simplicity, we write the error estimates in terms of  $h$ , but they can be traduced in terms of the number of degrees of freedom using this relationship.

Given a generic rectangle  $R$  with edges of lengths  $h_x$  and  $h_y$ , let  $\mathcal{Q}_1 : H^2(R) \rightarrow H^1(R)$  be the classical interpolation operator on  $R$ . We know the error estimates (see [3, Th. 2.7])

$$(3.5) \quad \|v - \mathcal{Q}_1 v\|_{0,R} \leq C \{h_x^2 \|\partial_x^2 v\|_{0,R} + h_y^2 \|\partial_y^2 v\|_{0,R}\}$$

and

$$(3.6) \quad \begin{aligned} \|\partial_x(v - \mathcal{Q}_1 v)\|_{0,R} &\leq C \{h_x \|\partial_x^2 v\|_{0,R} + h_y \|\partial_x \partial_y v\|_{0,R}\}, \\ \|\partial_y(v - \mathcal{Q}_1 v)\|_{0,R} &\leq C \{h_x \|\partial_x \partial_y v\|_{0,R} + h_y \|\partial_y^2 v\|_{0,R}\}. \end{aligned}$$

We also have the following results that will be useful later on.

**Lemma 3.1.** *Let  $R = (a, b) \times (c, d)$  be a rectangle with sides of lengths  $h_x = b - a$  and  $h_y = d - c$ . Then we have*

$$\|\nabla(\mathcal{Q}_1 f)\|_{\infty, R} \leq 2\sqrt{2}\|\nabla f\|_{\infty, R}$$

for all  $f \in \mathcal{C}^1(\bar{R})$ .

*Proof.* Let  $A = (a, c)$ ,  $B = (b, c)$ ,  $C = (b, d)$  and  $D = (a, d)$ , and the Lagrange bilinear bases functions

$$\begin{aligned} \lambda_A(x, y) &= \frac{(x-b)(y-d)}{h_x h_y}, & \lambda_B(x, y) &= -\frac{(x-a)(y-d)}{h_x h_y}, \\ \lambda_C(x, y) &= \frac{(x-a)(y-c)}{h_x h_y}, & \lambda_D(x, y) &= -\frac{(x-b)(y-c)}{h_x h_y}. \end{aligned}$$

Then

$$\mathcal{Q}_1 f = f(A)\lambda_A + f(B)\lambda_B + f(C)\lambda_C + f(D)\lambda_D$$

and

$$\partial_x(\mathcal{Q}_1 f)(x, y) = \frac{f(A) - f(B)}{h_x} \frac{y-d}{h_y} + \frac{f(C) - f(D)}{h_x} \frac{y-c}{h_y}.$$

Then, by the Mean Value Theorem we have that there exist  $x_{m_1}, x_{m_2} \in (a, b)$  such that

$$\partial_x(\mathcal{Q}_1 f)(x, y) = \partial_x f(x_{m_1}, c) \frac{y-d}{h_y} + \partial_x f(x_{m_2}, d) \frac{y-c}{h_y}.$$

Since, for  $(x, y) \in R$  it holds  $|y-c|, |y-d| \leq h_y$  then it results

$$|\partial_x(\mathcal{Q}_1 f)(x, y)| \leq |\partial_x f(x_{m_1}, c)| + |\partial_x f(x_{m_2}, d)| \leq 2\|\nabla f\|_{\infty, R}.$$

A similar estimate hold for  $|\partial_y(\mathcal{Q}_1 f)(x, y)|$  and then the proof concludes.  $\square$

**Lemma 3.2.** *Let  $R = (a, b) \times (c, d)$  be a rectangle with sides of lengths  $h_x = b - a$  and  $h_y = d - c$ . If  $0 < \alpha \leq 1$  then, for any  $v \in H^2(R)$ , we have*

$$(3.7) \quad \|\partial_x(v - \mathcal{Q}_1 v)\|_{0, R} \leq C\{h_x^{1-\alpha}\|(x-a)^\alpha \partial_x^2 v\|_{0, R} + h_y\|\partial_x \partial_y v\|_{0, R}\},$$

$$(3.8) \quad \|\partial_y(v - \mathcal{Q}_1 v)\|_{0, R} \leq C\{h_x\|\partial_x \partial_y v\|_{0, R} + h_y^{1-\alpha}\|(y-c)^\alpha \partial_y^2 v\|_{0, R}\}$$

*Proof.* Let  $\hat{R} = (0, 1)^2$  and  $\hat{\mathcal{Q}}_1 : H^2(\hat{R}) \rightarrow H^1(\hat{R})$  be the bilinear interpolation operator. For a function  $v \in H^2(\hat{R})$  define

$$\Pi v(x, y) = v(0, y)(1-x) + xv(1, y), \quad (x, y) \in \hat{R}.$$

Note that  $\Pi v(\cdot, y)$  is the linear interpolation of  $v(\cdot, y)$  for each  $y \in [0, 1]$ . Then we know that for smooth functions  $v$  we have (see [15, Corollary 1.2.3])

$$\|\partial_x[v(\cdot, y) - \Pi v(\cdot, y)]\|_{0, (0,1)} \leq C \|x^\alpha \partial_x^2 v(\cdot, y)\|_{0, (0,1)} \quad \forall y \in (0, 1)$$

and therefore

$$\begin{aligned} \|\partial_x(v - \Pi v)\|_{0, \hat{R}}^2 &= \int_0^1 \|\partial_x[v(\cdot, y) - \Pi v(\cdot, y)]\|_{0, (0,1)}^2 dy \\ &\leq C \int_0^1 \|x^\alpha \partial_x^2 v(\cdot, y)\|_{0, (0,1)}^2 dy = C \|x^\alpha \partial_x^2 v\|_{0, \hat{R}}^2. \end{aligned}$$

Now,

$$\begin{aligned} v - \hat{\mathcal{Q}}_1 v &= (v - \Pi v) + (\Pi v - \hat{\mathcal{Q}}_1 v) \\ &= (v - \Pi v) + [\Pi v - \hat{\mathcal{Q}}_1(\Pi v)] \end{aligned}$$

since  $\hat{\mathcal{Q}}_1 v = \hat{\mathcal{Q}}_1(\Pi v)$ . Then

$$(3.9) \quad \begin{aligned} \|\partial_x(v - \hat{\mathcal{Q}}_1 v)\|_{0, \hat{R}} &\leq C \|x^\alpha \partial_x^2 v\|_{0, \hat{R}} + \|\partial_x[\Pi v - \hat{\mathcal{Q}}_1(\Pi v)]\|_{0, \hat{R}} \\ &\leq C \{ \|x^\alpha \partial_x^2 v\|_{0, \hat{R}} + \|\partial_x^2 \Pi v\|_{0, \hat{R}} + \|\partial_x \partial_y \Pi v\|_{0, \hat{R}} \} \end{aligned}$$

where we used the estimate (3.6) for  $\hat{\mathcal{Q}}_1$ . But

$$\partial_x^2 \Pi v = 0$$

and, since

$$\partial_x \Pi v(x, y) = v(1, y) - v(0, y)$$

it follows, for smooth functions  $v$ , that

$$\partial_y \partial_x \Pi v(x, y) = \partial_y (v(1, y) - v(0, y)) = \int_0^1 \partial_y \partial_x v(t, y) dt.$$

Then

$$\begin{aligned} \|\partial_x \partial_y \Pi v\|_{0, \hat{R}}^2 &= \int_0^1 \int_0^1 \left| \int_0^1 \partial_x \partial_y v(t, y) dt \right|^2 dy dx \\ &\leq \int_0^1 \int_0^1 \int_0^1 |\partial_x \partial_y v(t, y)|^2 dt dy dx \\ &= \|\partial_x \partial_y v\|_{0, \hat{R}}^2. \end{aligned}$$

From (3.9) we obtain

$$\|\partial_x(v - \hat{\mathcal{Q}}_1 v)\|_{0, \hat{R}} \leq C \{ \|x^\alpha \partial_x^2 v\|_{0, \hat{R}} + \|\partial_x \partial_y v\|_{0, \hat{R}} \}.$$

By a density argument, the previous inequality holds for all  $v \in H^2(\hat{R})$ . Then, the inequality (3.7) is obtained by a simple rescaling argument. Inequality (3.8) follows analogously, and then the proof concludes.  $\square$   $\square$

We will denote the global continuous piecewise bilinear interpolation operator  $H^2(\Omega) \rightarrow H^1(\Omega)$  also by  $\mathcal{Q}_1$ .

#### 4. ESTIMATES ON GRADED MESHES

In this Section we obtain interpolation error estimates in the  $\beta$ -norm with graded meshes. We will assume the compatibility conditions (see [10] and the references therein)

$$f(0,0) = f(1,0) = f(1,1) = f(0,1) = 0$$

which ensure that the solution  $u$  belong to  $\mathcal{C}^4(\Omega) \cap \mathcal{C}^2(\overline{\Omega})$ . Such compatibility conditions are necessary for the following pointwise estimates, which for  $k \leq 2$  are proved in [10, Lemmata 3.1, 3.3 and 3.5] and for  $k = 3, 4$  were stated in [11, Lemma 4.1].

**Lemma 4.1.** *We have that, for  $(x, y) \in \Omega$  and  $0 \leq k \leq 4$ ,*

$$\begin{aligned} \left| \partial_x^k u(x, y) \right| &\leq C \left\{ 1 + \frac{1}{\varepsilon^k} e^{-b_0 \frac{x}{\varepsilon}} + \frac{1}{\varepsilon^k} e^{-b_0 \frac{1-x}{\varepsilon}} \right\}, \\ \left| \partial_y^k u(x, y) \right| &\leq C \left\{ 1 + \frac{1}{\varepsilon^k} e^{-b_0 \frac{y}{\varepsilon}} + \frac{1}{\varepsilon^k} e^{-b_0 \frac{1-y}{\varepsilon}} \right\}. \end{aligned}$$

Note that, with  $k = 0$  we obtain that the solution  $u$  is uniformly bounded on the domain  $\Omega$ .

In our analysis we make the following reasonable Assumption.

**Assumption 1.** *Assume that  $h < e^{-\frac{3}{2}}$  and  $\varepsilon < h$ , as otherwise the subsequent analysis can be carried out using standard techniques.*

First, we consider the  $L^2$ -part of the  $\beta$ -norm of the interpolation error.

**Proposition 4.1.** *Let  $u$  be the solution of (2.1) and  $\mathcal{Q}_1 u$  be the piecewise bilinear interpolation of  $u$  on the mesh  $\mathcal{T}_h$ . Then, under Assumption 1, we have that there exists a constant  $C$  such that*

$$\|u - \mathcal{Q}_1 u\|_\beta \leq Ch^2 \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}}.$$

*Proof.* Let us define

$$\begin{aligned} R_1 &= \{[(0, x_1) \cup (1 - x_1, 1)] \times (0, 1)\} \cup \{(0, 1) \times [(0, x_1) \cup (1 - x_1, 1)]\}, \\ R_2 &= \left\{ \left[ (x_1, \gamma_0 \varepsilon \log \frac{1}{\varepsilon}) \cup (1 - \gamma_0 \varepsilon \log \frac{1}{\varepsilon}, 1 - x_1) \right] \times (x_1, 1 - x_1) \right\} \\ &\quad \cup \left\{ (x_1, 1 - x_1) \times \left[ (x_1, \gamma_0 \varepsilon \log \frac{1}{\varepsilon}) \cup (1 - \gamma_0 \varepsilon \log \frac{1}{\varepsilon}, 1 - x_1) \right] \right\}, \\ R_3 &= (\gamma_0 \varepsilon \log \frac{1}{\varepsilon}, 1 - \gamma_0 \varepsilon \log \frac{1}{\varepsilon})^2, \end{aligned}$$



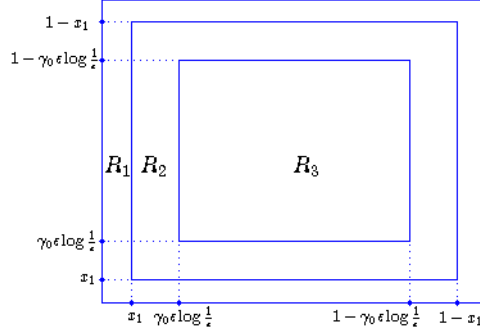


FIGURE 1. Decomposition of  $\Omega$  for the proof of Proposition 4.1

where  $\gamma_0$  is taken greater than or equal to  $\max\left\{\frac{2}{b_0}, \frac{1}{\gamma}\right\}$  and such that  $\gamma_0\epsilon \log \frac{1}{\epsilon}$  and  $1 - \gamma_0\epsilon \log \frac{1}{\epsilon}$  are grid points. Then,  $\Omega = R_1 \cup R_2 \cup R_3$  (see Figure 1).

Let  $S_1 = (0, x_1) \times (0, 1)$ . Note that, since  $1 - \alpha = \frac{1}{-2\log \epsilon}$  and  $\epsilon = h^{\frac{\log \epsilon}{\log h}}$ , we have

$$(4.10) \quad \frac{x_1}{\epsilon} = \frac{h^{-2\log \epsilon}}{\epsilon} = h^{\frac{\log \epsilon}{\log h} [2(-\log h) - 1]}.$$

then, by Assumption 1, since  $h < e^{-\frac{3}{2}}$  and  $\epsilon < h$ , it follows

$$\frac{\log \epsilon}{\log h} [2(-\log h) - 1] > 2(-\log h) - 1 > 2$$

and therefore

$$(4.11) \quad \frac{x_1}{\epsilon} \leq h^2.$$

Then, since  $u$  is uniformly bounded, we get

$$\begin{aligned} \|u - \mathcal{Q}_1 u\|_{\beta, S_1}^2 &= \int_0^{x_1} \int_0^1 \beta (u - \mathcal{Q}_1 u)^2 dy dx \\ &\leq C x_1 \epsilon^{-1} \leq h^2. \end{aligned}$$

Then clearly, by symmetry arguments, we obtain

$$(4.12) \quad \|u - \mathcal{Q}_1 u\|_{\beta, R_1}^2 \leq Ch^2.$$

Let now  $S_2 = (x_1, \gamma_0 \varepsilon \log \frac{1}{\varepsilon}) \times (x_1, 1 - x_1)$ . Using anisotropic interpolation error estimate [3, Th. 2.7] and taking into account that  $\beta \leq C\varepsilon^{-1}$ , we have

$$\begin{aligned} \|u - \mathcal{Q}_1 u\|_{\beta, S_2}^2 &\leq C\varepsilon^{-1} \|u - \mathcal{Q}_1 u\|_{0, S_2}^2 \\ &\leq C\varepsilon^{-1} \sum_{R_{ij} \subset S_2} \left( h_i^4 \|\partial_x^2 u\|_{0, R_{ij}}^2 + h_j^4 \|\partial_y^2 u\|_{0, R_{ij}}^2 \right). \end{aligned}$$

For  $R_{ij} \subset S_2$  we have  $h_i \leq Chx^\alpha$ ,  $h_j \leq Ch \min(y, 1 - y)^\alpha$  for all  $(x, y) \in R_{ij}$ . Using also that  $h_i, h_j \leq h$  and the a priori estimates of Lemma 4.1 we obtain

$$(4.13) \quad \|u - \mathcal{Q}_1 u\|_{\beta, S_2}^2 \leq Ch^4 \varepsilon^{-1} \times \int_{S_2} \left( 1 + x^{4\alpha} \varepsilon^{-4} e^{-2b_0 \frac{x}{\varepsilon}} + y^{4\alpha} \varepsilon^{-4} e^{-2b_0 \frac{y}{\varepsilon}} + (1 - y)^4 \varepsilon^{-4} e^{-2b_0 \frac{1-y}{\varepsilon}} \right) dx dy.$$

Now, taking into account that  $|S_2| \leq \gamma_0 \varepsilon \log \frac{1}{\varepsilon}$  we get

$$Ch^4 \varepsilon^{-1} \int_{S_2} dx dy \leq Ch^4 \log \frac{1}{\varepsilon}.$$

On the other hand, we have

$$\begin{aligned} Ch^4 \varepsilon^{-1} \int_{S_2} \left( x^{4\alpha} \varepsilon^{-4} e^{-2b_0 \frac{x}{\varepsilon}} + y^{4\alpha} \varepsilon^{-4} e^{-2b_0 \frac{y}{\varepsilon}} \right) dx dy &= \\ Ch^4 (1 - 2x_1) \varepsilon^{4(\alpha-1)} \int_{x_1}^{\gamma_0 \varepsilon \log \frac{1}{\varepsilon}} (x \varepsilon^{-1})^{4\alpha} e^{-2b_0 x / \varepsilon} \frac{dx}{\varepsilon} & \\ + Ch^4 (\gamma_0 \varepsilon \log \frac{1}{\varepsilon} - x_1) \varepsilon^{4(\alpha-1)} \int_{x_1}^{1-x_1} (y \varepsilon^{-1})^{4\alpha} e^{-2b_0 y / \varepsilon} \frac{dy}{\varepsilon} & \\ \leq Ch^4. & \end{aligned}$$

where we used that  $\varepsilon \log \frac{1}{\varepsilon} \leq C$ ,  $\varepsilon^{4(\alpha-1)} = e^2$  and that for  $\delta \in [0, 4]$  the integrals  $\int_0^\infty x^\delta e^{-2b_0 x} dx$  are uniformly bounded. Similarly, we have

$$Ch^4 \varepsilon^{-1} \int_{S_2} (1 - y)^4 \varepsilon^{-4} e^{-2b_0 \frac{1-y}{\varepsilon}} dx dy \leq Ch^4.$$

Then from (4.13) we get

$$\|u - \mathcal{Q}_1 u\|_{\beta, S_2} \leq Ch^2 \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}}.$$

Now, with similar arguments we obtain

$$(4.14) \quad \|u - \mathcal{Q}_1 u\|_{\beta, R_2} \leq Ch^2 \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}}.$$

Finally, since  $\gamma_0 \geq \frac{2}{b_0}$ , it follows from Lemma 4.1 that

$$|\partial_x^2 u| + |\partial_y^2 u| \leq C \quad \text{on } R_3.$$

Similarly,  $\beta \leq C$  on  $R_3$  since  $\gamma_0 > \frac{1}{\gamma}$ . Then, using again the anisotropic interpolation error estimates for the operator  $\mathcal{Q}_1$  and that  $h_i, h_j \leq h$  for all  $i, j$ , we easily obtain

$$(4.15) \quad \|u - \mathcal{Q}_1 u\|_{\beta, R_3} \leq Ch^2.$$

Since  $\Omega = R_1 \cup R_2 \cup R_3$ , from (4.12), (4.14) and (4.15) we get the desired result.  $\square$   $\square$

Also, we can prove the following result involving the  $H^1$ -seminorm.

**Proposition 4.2.** *Let  $u$  be the solution of (2.1) and  $\mathcal{Q}_1 u$  be the piecewise bilinear interpolation of  $u$  on the mesh  $\mathcal{T}_h$ . Then, under Assumption 1, we have*

$$\|\nabla(u - \mathcal{Q}_1 u)\|_0 \leq C\varepsilon^{-\frac{1}{2}}h.$$

*Proof.* Let us estimate  $\|\nabla(u - \mathcal{Q}_1 u)\|_{0, \Omega_s}$  where  $\Omega_s = [0, \frac{1}{2}] \times [0, \frac{1}{2}]$ . Then estimate on the rest of the domain follows by symmetry. Let us introduce the notation

$$\Omega_i = \cup_{j=1}^{mid} R_{ij}, \quad \Omega^j = \cup_{i=1}^{mid} R_{ij}.$$

Using inequalities (3.6) and (3.7) on each element  $R_{ij}$  we have

$$\begin{aligned} \|\partial_x(u - \mathcal{Q}_1 u)\|_{0, \Omega_s}^2 &\leq h_1^{2-2\alpha} \|x^\alpha \partial_x^2 u\|_{0, \Omega_1}^2 + \sum_{i=2}^{mid} h_i^2 \|\partial_x^2 u\|_{0, \Omega_i}^2 + \sum_{j=1}^{mid} h_j^2 \|\partial_x \partial_y u\|_{0, \Omega^j}^2 \\ &\leq h_1^{2-2\alpha} \|x^\alpha \partial_x^2 u\|_{0, \Omega_1}^2 + \sum_{i=2}^{mid} h^2 \|x^\alpha \partial_x^2 u\|_{0, \Omega_i}^2 + \sum_{j=1}^{mid} h_j^2 \|\partial_x \partial_y u\|_{0, \Omega^j}^2 \end{aligned}$$

where, for the second line, we used that

$$h_i \leq hx^\alpha \quad \forall (x, y) \in R_{ij}, 2 \leq i.$$

Since  $h_1 = h^s$  with  $s = \frac{1}{1-\alpha}$ , we have that  $h_1^{2-2\alpha} = h^2$ . On the other hand we have

$$h_j \leq hy^\alpha \quad \forall (x, y) \in R_{ij}, 2 \leq j.$$

Then we have

$$(4.16) \quad \|\partial_x(u - \mathcal{Q}_1 u)\|_{0, \Omega_s}^2 \leq h^2 \|x^\alpha \partial_x^2 u\|_{0, \Omega_s}^2 + h^2 \|y^\alpha \partial_x \partial_y u\|_{0, \Omega_s \setminus \Omega_1}^2 + h_1^2 \|\partial_x \partial_y u\|_{0, \Omega_1}^2.$$

We need to bound each term in the last inequality. By integration by parts twice, Lemma 4.1 and using that

$$\partial_y u = 0 \quad \text{on } x = 0 \text{ and } x = 1, \quad \partial_x^2 u = 0 \quad \text{on } y = 0$$

we have

$$\begin{aligned}
\|\partial_x \partial_y u\|_{0,\Omega^1}^2 &\leq \|\partial_x \partial_y u\|_{0,[0,1] \times [0,x_1]}^2 \\
&= \int_0^{x_1} \partial_x \partial_y u \partial_y u|_0^1 dy - \int_0^1 \int_0^{x_1} \partial_y u \partial_y \partial_x^2 u dy dx \\
&= \int_0^1 \int_0^{x_1} \partial_y^2 u \partial_x^2 u dy dx - \int_0^1 \partial_y u \partial_x^2 u|_0^{x_1} dx \\
&\leq \frac{x_1}{\varepsilon^2} \int_0^1 \left(1 + \frac{1}{\varepsilon^2} e^{-b_0 \frac{x}{\varepsilon}}\right) dx + \frac{1}{\varepsilon} \int_0^1 \left(1 + \frac{1}{\varepsilon^2} e^{-b_0 \frac{x}{\varepsilon}}\right) dx \\
&\leq C\varepsilon^{-3}.
\end{aligned}$$

By Assumption 1 we have, in particular, that  $\varepsilon, h < e^{-1}$ , thus

$$h_1^2 = h^{2s} = h^{4 \log \frac{1}{\varepsilon}} = h^{2 \log \frac{1}{\varepsilon}} h^{2 \log \frac{1}{\varepsilon}} \leq h^2 \varepsilon^{2 \log \frac{1}{h}} = h^2 \varepsilon^2,$$

therefore

$$(4.17) \quad h_1^2 \|\partial_x \partial_y u\|_{0,\Omega^1}^2 \leq Ch^2 \varepsilon^{-1}.$$

On the other hand, using again the estimates of Lemma 4.1 we have

$$\begin{aligned}
\|x^\alpha \partial_x^2 u\|_{0,\Omega_s}^2 &\leq C \int_0^{\frac{1}{2}} x^{2\alpha} \left(1 + \frac{1}{\varepsilon^4} e^{-2b_0 \frac{x}{\varepsilon}}\right) dx \\
&\leq C + C\varepsilon^{2\alpha-3} \int_0^{\frac{1}{2}} \left(\frac{x}{\varepsilon}\right)^{2\alpha} e^{-2b_0 \frac{x}{\varepsilon}} \frac{dx}{\varepsilon} \\
&\leq C\varepsilon^{-1},
\end{aligned}$$

and so

$$(4.18) \quad h^2 \|x^\alpha \partial_x^2 u\|_{0,\Omega}^2 \leq Ch^2 \varepsilon^{-1}.$$

Finally, with the same arguments, see [6, ineq. (4.31)] for a similar computation, we have

$$\begin{aligned}
 \|y^\alpha \partial_x \partial_y u\|_{0, \Omega_s \setminus \Omega^1}^2 &\leq \|y^\alpha \partial_x \partial_y u\|_{0, [0,1] \times [0, \frac{1}{2}]}^2 \\
 &= - \left(\frac{1}{2}\right)^{2\alpha} \int_0^1 \partial_y u \left(x, \frac{1}{2}\right) \partial_x^2 u \left(x, \frac{1}{2}\right) dx \\
 &\quad + \int_0^1 \int_0^{\frac{1}{2}} 2\alpha y^{2\alpha-1} \partial_y u \partial_x^2 u dy dx \\
 &\quad + \int_0^1 \int_0^{\frac{1}{2}} y^{2\alpha} \partial_y^2 u \partial_x^2 u dy dx \\
 &\leq C \int_0^1 \left(1 + \frac{1}{\varepsilon^2} e^{-b_0 \frac{x}{\varepsilon}}\right) dx \\
 &\quad + C \int_0^1 y^{2\alpha-1} \left(1 + \frac{1}{\varepsilon} e^{-b_0 \frac{y}{\varepsilon}}\right) dy \int_0^1 \left(1 + \frac{1}{\varepsilon^2} e^{-b_0 \frac{x}{\varepsilon}}\right) dx \\
 &\quad + C \int_0^1 y^{2\alpha} \left(1 + \frac{1}{\varepsilon^2} e^{-b_0 \frac{y}{\varepsilon}}\right) dy \int_0^1 \left(1 + \frac{1}{\varepsilon^2} e^{-b_0 \frac{x}{\varepsilon}}\right) dx \\
 &\leq C\varepsilon^{-1},
 \end{aligned}$$

that is

$$(4.19) \quad h^2 \|y^\alpha \partial_x \partial_y u\|_{0, \Omega_s \setminus \Omega^1}^2 \leq Ch^2 \varepsilon^{-1}.$$

Now inserting (4.17)–(4.19) in (4.16) we obtain

$$\|\partial_x(u - \mathcal{Q}_1 u)\|_{0, \Omega_s} \leq Ch\varepsilon^{-\frac{1}{2}},$$

and by symmetry it follows

$$\|\partial_x(u - \mathcal{Q}_1 u)\|_{0, \Omega} \leq Ch\varepsilon^{-\frac{1}{2}}.$$

Clearly, the estimate

$$\|\partial_y(u - \mathcal{Q}_1 u)\|_{0, \Omega} \leq Ch\varepsilon^{-\frac{1}{2}}$$

can be proved in a similar way concluding the proof.  $\square$   $\square$

We remark that, from the definition of the  $\beta$ -norm  $\|\cdot\|_\beta$  and Proposition 4.2, since  $\beta \leq C\frac{1}{\varepsilon}$ , we get

$$(4.20) \quad \varepsilon \|\nabla(u - \mathcal{Q}_1 u)\|_{\beta, \Omega} \leq C\varepsilon^{\frac{1}{2}} \|\nabla(u - \mathcal{Q}_1 u)\|_{0, \Omega} \leq Ch,$$

which together with Proposition 4.1 allows us to obtain the main result of this Section.

**Theorem 4.1.** *Let  $u$  be the solution of (2.1) and  $\mathcal{Q}_1 u$  be the piecewise bilinear interpolation of  $u$  on the mesh  $\mathcal{T}_h$ . Then, under Assumption 1, we have*

$$\| \|u - \mathcal{Q}_1 u\| \|_{\beta} \leq Ch \left( 1 + h \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} \right).$$

Clearly as a consequence of Céa Lemma, equation (2.3) and this Theorem we have the corresponding error estimate for the finite element approximation  $u_h$ .

## 5. SUPERCLOSENESS

In this section we prove that the  $\beta$ -norm of the difference between the interpolation of the exact solution  $u$  and the finite element approximation  $u_h$  is of higher order than the  $\beta$ -norm of the error  $u - u_h$ .

Let us denote by  $\beta_{min}$  and  $\beta_{max}$  the piecewise constant functions such that on each element  $R \in \mathcal{T}_h$  hold

$$\beta_{min}|_R = \min_{(x,y) \in R} \beta(x,y), \quad \beta_{max}|_R = \max_{(x,y) \in R} \beta(x,y).$$

Clearly  $\beta_{min}$  and  $\beta_{max}$  depend on the mesh  $\mathcal{T}_h$  but this dependence is omitted for the sake of simplicity of the notation. The following Lemma presents an estimation of the relation of  $\beta_{min}$  and  $\beta_{max}$  inside the elements which is fundamental for our estimations.

**Lemma 5.1.** *There exists a positive constant  $\eta$ , independent of  $h$  and  $\varepsilon$ , such that on graded meshes  $\mathcal{T}_h$ , assuming  $h < e^{-1}$ , we get*

$$\frac{\beta_{max}}{\beta_{min}} \leq C\varepsilon^{-\eta h} \quad \text{on } \Omega.$$

*Proof.* Due to the symmetry of the problem it is enough to estimate  $\beta_{max}/\beta_{min}$  for elements contained in  $\Omega_s = [0, \frac{1}{2}] \times [0, \frac{1}{2}]$ . On elements  $R_{1j}$  or  $R_{i1}$  we have

$$\begin{aligned} \beta_{min} &= 1 + \frac{1}{\varepsilon} e^{-\frac{\gamma}{\varepsilon} d_{max}} = 1 + \frac{1}{\varepsilon} e^{-\frac{\gamma}{\varepsilon} h^s}, \\ \beta_{max} &= 1 + \frac{1}{\varepsilon} e^{-\frac{\gamma}{\varepsilon} d_{min}} = 1 + \frac{1}{\varepsilon}, \end{aligned}$$

where naturally  $d_{max}$  and  $d_{min}$  represent the maximum and the minimum of the distance to the boundary. But

$$h^s = h^{2 \log \frac{1}{\varepsilon}} = \varepsilon^{2 \log \frac{1}{h}}$$

and so, since  $h < \frac{1}{e}$ ,

$$-\frac{\gamma}{\varepsilon} h^s = -\gamma \varepsilon^{2 \log \frac{1}{h} - 1} > -\gamma \varepsilon$$

and therefore

$$\beta_{min} > 1 + \frac{1}{\varepsilon} e^{-\gamma\varepsilon}.$$

Then we can conclude that

$$\frac{\beta_{max}}{\beta_{min}} < \frac{1 + \frac{1}{\varepsilon}}{1 + \frac{1}{\varepsilon} e^{-\gamma\varepsilon}} < C.$$

Now, we consider a rectangle  $R_{ij}$  with

$$i > 1 \quad \text{and} \quad x_j \leq \gamma_0 \varepsilon \log \frac{1}{\varepsilon}$$

or

$$j > 1 \quad \text{and} \quad x_i \leq \gamma_0 \varepsilon \log \frac{1}{\varepsilon}.$$

It can be checked that it is enough to consider a case as in the Figure 2, and we will use the notation of that Figure. We have

$$d_{min} = y_1, \quad d_{max} = y_2,$$

and then  $\beta_{min} = 1 + \frac{1}{\varepsilon} e^{-\gamma \frac{y_2}{\varepsilon}}$  and  $\beta_{max} = 1 + \frac{1}{\varepsilon} e^{-\gamma \frac{y_1}{\varepsilon}}$ . Then

$$\begin{aligned} \frac{\beta_{max}}{\beta_{min}} &= \frac{1 + \frac{1}{\varepsilon} e^{-\gamma \frac{d_{min}}{\varepsilon}}}{1 + \frac{1}{\varepsilon} e^{-\gamma \frac{d_{max}}{\varepsilon}}} = 1 + \frac{\frac{1}{\varepsilon} e^{-\gamma \frac{d_{max}}{\varepsilon}}}{1 + \frac{1}{\varepsilon} e^{-\gamma \frac{d_{max}}{\varepsilon}}} \left( e^{-\frac{\gamma}{\varepsilon}(d_{min}-d_{max})} - 1 \right) \\ &= 1 + \frac{\frac{1}{\varepsilon} e^{-\gamma \frac{y_2}{\varepsilon}}}{1 + \frac{1}{\varepsilon} e^{-\gamma \frac{y_2}{\varepsilon}}} \left( e^{-\frac{\gamma}{\varepsilon}(y_1-y_2)} - 1 \right) \leq 1 + e^{\frac{\gamma}{\varepsilon}(y_2-y_1)} - 1 \\ &= e^{\frac{\gamma}{\varepsilon} h y_1^\alpha}. \end{aligned}$$

But, since  $\varepsilon^{\frac{1}{\log \varepsilon}} = e$ , we obtain

$$\begin{aligned} h y_1^\alpha &\leq h \left( \gamma_0 \varepsilon \log \frac{1}{\varepsilon} \right)^{1 - \frac{1}{2 \log \frac{1}{\varepsilon}}} = h \gamma_0 \varepsilon \log \frac{1}{\varepsilon} \left( \gamma_0 \varepsilon \log \frac{1}{\varepsilon} \right)^{\frac{1}{2 \log \varepsilon}} \\ &= h \varepsilon e^{\frac{1}{2}} \gamma_0^{1 + \frac{1}{2 \log \varepsilon}} \log \frac{1}{\varepsilon} \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2 \log \varepsilon}} \leq C h \varepsilon \log \frac{1}{\varepsilon} \end{aligned}$$

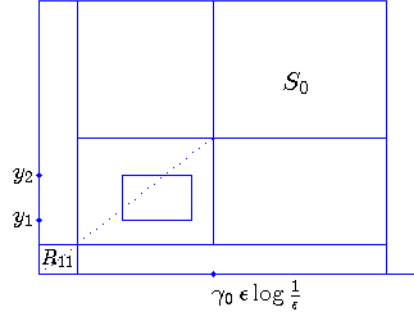
where we used that  $\gamma_0^{1 + \frac{1}{2 \log \varepsilon}} \leq C$  and  $\left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2 \log \frac{1}{\varepsilon}}} \leq C$ . Then we have

$$e^{\frac{\gamma}{\varepsilon} h y_1^\alpha} \leq e^{C h \gamma \log \frac{1}{\varepsilon}} = \varepsilon^{-\gamma C h}.$$

Finally, it is clear that on rectangles  $R_{ij}$  with  $x_i$  or  $x_j$  greater than  $\gamma_0 \varepsilon \log \frac{1}{\varepsilon}$  we have

$$\frac{\beta_{max}}{\beta_{min}} \sim 1$$

and this concludes the proof.  $\square$

FIGURE 2. Notation on  $\Omega_s$  for the proof of Lemma 5.1

□

For the meshes  $\mathcal{T}_h$  we also introduce the piecewise constant function  $h_{min}$  which on each rectangle  $R \in \mathcal{T}_h$  take the minimum of the lengths of the sides of  $R$ . Taking into account that the graph of the distance function  $d$  is a square pyramid with its apex on the point  $(\frac{1}{2}, \frac{1}{2})$ , it can be checked that, given an element  $R \in \mathcal{T}_h$  and  $(x, y) \in R$  there exists  $(x_{int}, y_{int}) \in R$  such that

$$(5.21) \quad |\beta(x, y) - \beta_{min}| \leq Ch_{min} |\nabla \beta(x_{int}, y_{int})|.$$

From the coerciveness and the Galerkin orthogonality of the bilinear form  $B_\beta(\cdot, \cdot)$  we get

$$(5.22) \quad \begin{aligned} C \|u_h - \mathcal{Q}_1 u\|_\beta^2 &\leq B_\beta(u_h - \mathcal{Q}_1 u, u_h - \mathcal{Q}_1 u) \\ &= B_\beta(u - \mathcal{Q}_1 u, u_h - \mathcal{Q}_1 u) \end{aligned}$$



Now, for any  $w \in V_h$  we have

(5.23)

$$\begin{aligned}
 B_\beta(u - \mathcal{Q}_1 u, w) &= \int_{\Omega} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot \nabla(\beta w) dx + \int_{\Omega} b(x)(u - \mathcal{Q}_1 u)(\beta w) dx \\
 &= \int_{\Omega} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \beta \cdot \nabla(w) dx + \int_{\Omega} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot \nabla(\beta) w dx + \\
 &\quad \int_{\Omega} b(x)(u - \mathcal{Q}_1 u)(\beta w) dx \\
 &= \int_{\Omega} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u)(\beta - \beta_{min}) \cdot \nabla w + \int_{\Omega} \varepsilon^2 \beta_{min} \nabla(u - \mathcal{Q}_1 u) \cdot \nabla w + \\
 &\quad \int_{\Omega} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot \nabla(\beta) w + \int_{\Omega} b(x)(u - \mathcal{Q}_1 u) \beta w \\
 &=: I + II + III + IV
 \end{aligned}$$

In the next subsections, we will prove the following estimates for  $I, II, III$  and  $IV$  assuming that  $\varepsilon \leq ch^3$ , for some fixed constant  $c \geq 1$ ,

$$|I|, |III| \leq Ch^2 \varepsilon^{-\eta h} \log \frac{1}{\varepsilon} \|w\|_\beta \quad \text{and} \quad |II|, |IV| \leq Ch^2 \|w\|_\beta.$$

Therefore,

$$|B_\beta(u - \mathcal{Q}_1 u, u_h - \mathcal{Q}_1 u)| \leq Ch^2 \varepsilon^{-\eta h} \log \frac{1}{\varepsilon} \|w\|_\beta,$$

which together with (5.22) will allow us to conclude the following super-closeness result.

**Theorem 5.1.** *There exist positive constants  $C$  and  $\eta$ , independent of  $\varepsilon$  and  $h$ , such that on graded meshes  $\mathcal{T}_h$ , assuming that  $h < e^{-\frac{3}{2}}$  and  $\varepsilon < ch^3$ , we get*

$$\|u_h - \mathcal{Q}_1 u\|_\beta \leq C \varepsilon^{-\eta h} \log(1/\varepsilon)^{\frac{1}{2}} h^2.$$

**5.1. Estimation of term I.** Let us estimate

$$I = \int_{\Omega} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u)(\beta - \beta_{min}) \cdot \nabla w.$$

Using property (5.21) and Lemma 5.1 we have

$$\begin{aligned}
 |\beta(x, y) - \beta_{min}| &\leq Ch_{min} |\nabla \beta(x_{int}, y_{int})| \leq Ch_{min} \varepsilon^{-1} |\beta(x_{int}, y_{int})| \\
 &\leq Ch_{min} \varepsilon^{-1} \varepsilon^{-\eta h} |\beta(x, y)|.
 \end{aligned}$$

Let

$$S_0 = \left\{ (x, y) \in \Omega : \min(x, y, 1-x, 1-y) \leq \gamma_0 \varepsilon \log \frac{1}{\varepsilon} \right\}.$$

For elements  $R \subset S_0$  we have  $h_{\min, R} \leq \gamma_0 h \varepsilon \log \frac{1}{\varepsilon}$ , and therefore

$$|\beta(x, y) - \beta_{min}| \leq Ch \varepsilon^{-\eta h} \beta(x, y) \log \frac{1}{\varepsilon}.$$

Then

$$\begin{aligned} & \left| \int_{S_0} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u)(\beta - \beta_{min}) \cdot \nabla w \right| \\ & \leq Ch \varepsilon^{-\eta h} \log \frac{1}{\varepsilon} \left[ \varepsilon \|\beta^{\frac{1}{2}} \nabla(u - \mathcal{Q}_1 u)\|_{0,S_0} \right] \left[ \varepsilon \|\beta^{\frac{1}{2}} \nabla w\|_{0,S_0} \right] \\ & \leq Ch^2 \varepsilon^{-\eta h} \log \left( \frac{1}{\varepsilon} \right) \|w\|_{\beta,S_0} \end{aligned}$$

where in the last inequality we used estimate (4.20).

Now, let  $S_1 = \Omega \setminus S_0$ . Since  $\gamma_0 \geq \frac{2}{b_0}$  we have

$$\beta(x,y), |\nabla \beta(x,y)| \leq C \quad \forall (x,y) \in S_1$$

and therefore it is easy to check that

$$|\beta(x,y) - \beta_{min}| \leq Ch \quad (x,y) \in R, \quad R \subset S_1.$$

Also, since  $\beta \geq 1$  we have, on  $\|\cdot\|_{\beta,S_1} \sim \|\cdot\|_{0,S_1}$ . Then, using again (4.20) we get

$$\begin{aligned} \left| \int_{S_1} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u)(\beta - \beta_{min}) \cdot \nabla w \right| & \leq Ch \varepsilon \|\nabla(u - \mathcal{Q}_1 u)\|_{0,S_1} \varepsilon \|\nabla w\|_{0,S_1} \\ & \leq Ch \varepsilon \|\nabla(u - \mathcal{Q}_1 u)\|_{\beta,S_1} \varepsilon \|\nabla w\|_{\beta,S_1} \\ & \leq Ch^2 \|w\|_{\beta,S_1}. \end{aligned}$$

Then finally we obtain

$$(5.24) \quad |I| \leq Ch^2 \varepsilon^{-\eta h} \log \left( \frac{1}{\varepsilon} \right) \|w\|_{\beta}.$$

**5.2. Estimate of II.** Now we consider

$$II = \int_{\Omega} \beta_{min} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot \nabla w.$$

Since  $\beta_{min}$  is piecewise constant we can use an argument due to Zlamal [23], as in [5, Lema 4.5], to obtain that for each element  $R_{ij}$  we have

$$\begin{aligned} & \left| \varepsilon^2 \int_{R_{ij}} \beta_{min} \partial_x(u - \mathcal{Q}_1 u) \partial_x w \right| \\ & \leq C \varepsilon \beta_{min}^{\frac{1}{2}} \{ h_i^2 \|\partial_{xxx} u\|_{0,R_{ij}} + h_i h_j \|\partial_{xxy} u\|_{0,R_{ij}} + h_j^2 \|\partial_{xyy} u\|_{0,R_{ij}} \} \\ & \quad \times \varepsilon \beta_{min}^{\frac{1}{2}} \|\partial_x w\|_{0,R_{ij}} \\ & \leq C \varepsilon \beta_{min}^{\frac{1}{2}} \{ h_i^2 \|\partial_{xxx} u\|_{0,R_{ij}} + h_i h_j \|\partial_{xxy} u\|_{0,R_{ij}} + h_j^2 \|\partial_{xyy} u\|_{0,R_{ij}} \} \\ & \quad \times \|w\|_{\beta,R_{ij}} \end{aligned}$$

In Lemma 6.2, in the Appendix, we prove that

$$\begin{aligned} \varepsilon \left[ \sum_{i,j} \beta_{min} (h_i^2 \|\partial_x^3 u\|_{0,R_{ij}} + h_i h_j \|\partial_x^2 \partial_y u\|_{0,R_{ij}} + h_j^2 \|\partial_x \partial_y^2 u\|_{0,R_{ij}})^2 \right]^{\frac{1}{2}} \\ \leq C \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} h^2 \end{aligned}$$

from which we can conclude that

$$(5.25) \quad H \leq C \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} h^2 \|w\|_{\beta}.$$

**5.3. Estimate of III.** Now we deal with the estimate for *III*. Let  $\Omega_s = [0, \frac{1}{2}] \times [0, \frac{1}{2}]$ . Then it is clear that due to symmetry arguments it is enough to estimate  $III_s$  which is defined as *III* but with the integral over  $\Omega_s$ . We have

$$\begin{aligned} III_s &= \int_{\Omega_s} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla \beta) w \\ &= \int_{D_1} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla \beta) w + \int_{D_2} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla \beta) w \\ &= III_1 + III_2 \end{aligned}$$

with

$$D_1 = \bigcup \{R : \nabla d \text{ is discontinuous on } R\}, \quad D_2 = \Omega \setminus D_1.$$

Let  $S_A = [0, \gamma_0 \varepsilon \log \frac{1}{\varepsilon}]^2$  and  $S_B = [\gamma_0 \varepsilon \log \frac{1}{\varepsilon}, \frac{1}{2}]^2$ . Then

$$D_1 = (D_1 \cap S_A) \cup (D_1 \cap S_B)$$

and we can put

$$\begin{aligned} III_{1,A} &:= \int_{D_1 \cap S_A} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla \beta) w, \\ III_{1,B} &:= \int_{D_1 \cap S_B} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla \beta) w. \end{aligned}$$

We will use that

$$|\nabla(u - \mathcal{Q}_1 u)| \leq C \|\nabla u\|_{\infty} \leq C \varepsilon^{-1}$$

which follows from Lemma 3.1 and from the a priori estimates of Lemma 4.1. Then using that  $|\nabla\beta| \leq C\beta/\varepsilon$  we have

$$\begin{aligned} |III_{1,A}| &= \left| \int_{D_1 \cap S_A} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla\beta) w \right| \leq C \int_{D_1 \cap S_A} \beta |w| \\ &\leq C \left( \int_{D_1 \cap S_A} \beta \right)^{\frac{1}{2}} \|\beta^{\frac{1}{2}} w\|_{0, D_1 \cap S_A} \leq C \left( \int_{D_1 \cap S_A} \beta \right)^{\frac{1}{2}} \|w\|_{\beta}. \end{aligned}$$

Since rectangles in  $D_1 \cap S_A$  have sides of lengths  $O\left(h\left(\varepsilon \log \frac{1}{\varepsilon}\right)^\alpha\right)$  it is not difficult to see that

$$\begin{aligned} \int_{D_1 \cap S_A} \beta &\leq Ch \left( \varepsilon \log \frac{1}{\varepsilon} \right)^\alpha \int_0^{\gamma_0 \varepsilon \log \frac{1}{\varepsilon}} \frac{1}{\varepsilon} e^{-\gamma x/\varepsilon} dx + |D_1 \cap S_A| \\ &\leq Ch \left( \varepsilon \log \frac{1}{\varepsilon} \right)^\alpha \\ &\leq Ch \varepsilon \log \frac{1}{\varepsilon} \end{aligned}$$

where for the last inequality we recall that  $\alpha = 1 - \frac{1}{2 \log \frac{1}{\varepsilon}}$ . It follows that

$$|III_{1,A}| \leq Ch^{\frac{1}{2}} \varepsilon^{\frac{1}{2}} \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} \|w\|_{\beta} \leq Ch^2 \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} \|w\|_{\beta},$$

since  $\varepsilon \leq ch^3$ .

On the other hand on  $S_B$  we have  $|\nabla\beta|, |\nabla u|, |\nabla \mathcal{Q}_1 u|$  bounded independently of  $\varepsilon$ . We also have  $|D_1 \cap S_B| \leq Ch$ . Then

$$\begin{aligned} |III_{1,B}| &= \left| \int_{D_1 \cap S_B} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla\beta) w \right| \\ &\leq C\varepsilon^2 |D_1 \cap S_B|^{\frac{1}{2}} \|w\|_{0, D_1 \cap S_B} \\ &\leq C\varepsilon^2 h^{\frac{1}{2}} \|w\|_{\beta} \leq Ch^2 \|w\|_{\beta}, \end{aligned}$$

by using that  $\varepsilon < h$ .

Thus, we obtain

$$(5.26) \quad III_1 \leq Ch^2 \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} \|w\|_{\beta}.$$

Now we consider

$$III_2 = \int_{D_2} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla\beta) w.$$

On each element  $R$  let  $(\nabla\beta)_{min,R}$  be the componentwise minimum of  $\nabla\beta$  on  $R$ . We write

$$\begin{aligned} III_2 &= \int_{D_2} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla\beta - (\nabla\beta)_{min}) w + \\ &\quad \int_{D_2} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla\beta)_{min} w \\ &= III_{21} + III_{22}. \end{aligned}$$

Let  $D_{2A}^1 = D_2 \cap \{(x, y) : x \leq y \leq 1 - x, x \leq \gamma_0 \varepsilon \log \frac{1}{\varepsilon}\}$ . Notice that  $d(x, y) = x$  on  $D_{2A}^1$ . It follows that

$$\nabla\beta(x, y) = \left( -\frac{\gamma}{\varepsilon^2} e^{-\gamma \frac{x}{\varepsilon}}, 0 \right) \quad \text{on } D_{2A}^1.$$

By the Mean Value Theorem, since  $\nabla\beta$  depends only on  $x$ , and that for rectangular elements on  $D_{2A}^1$  the horizontal sides have the minimum length, we have

$$\nabla\beta(x, y) - (\nabla\beta)_{min,R} = \left( \frac{\gamma}{\varepsilon^3} e^{-\gamma \frac{x_{int}}{\varepsilon}} (x - x_{min,R}), 0 \right)$$

with  $x_{int} \in R$  and  $x_{min,R}$  being the minimum value of  $x$  on  $R$ . Now, since  $\frac{1}{\varepsilon} e^{-\gamma \frac{x_{int}}{\varepsilon}} \leq \beta(x_{int}) \leq \beta_{max}$  taking into account Lemma 5.1 we have

$$|\nabla\beta(x, y) - (\nabla\beta)_{min,R}| \leq C \varepsilon^{-\eta h} \varepsilon^{-2} \beta_{min,R} h_{min,R}.$$

But  $h_{min,R} \leq Ch(\varepsilon \log \frac{1}{\varepsilon})^\alpha \leq Ch\varepsilon \log \frac{1}{\varepsilon}$  (on elements touching  $x = 0$  we also have  $h_{min} \leq Ch\varepsilon$ ). So

$$|\nabla\beta(x, y) - (\nabla\beta)_{min,R}| \leq C \varepsilon^{-\eta h} \varepsilon^{-1} \beta_{min,R} h \log \frac{1}{\varepsilon}.$$

Then

$$\begin{aligned} &\left| \int_{D_{2A}^1} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla\beta - (\nabla\beta)_{min}) w \right| \\ &\leq C \varepsilon^{-\eta h} h \log \left( \frac{1}{\varepsilon} \right) \int_{D_{2A}^1} \left[ \varepsilon \beta^{\frac{1}{2}} |\nabla(u - \mathcal{Q}_1 u)| \right] \left[ \beta^{\frac{1}{2}} |w| \right] \\ &\leq C \varepsilon^{-\eta h} h^2 \log \frac{1}{\varepsilon} \|w\|_\beta \end{aligned}$$

where we have used (4.20).

On  $D_{2A}^2 = D_2 \cap \{(x, y) : x \leq y \leq 1 - x, x > \gamma_0 \varepsilon \log \frac{1}{\varepsilon}\}$  we also have that  $\beta$ ,  $|D^\delta(\beta)|$ ,  $0 \leq |\delta| \leq 2$ , are uniformly bounded respect of  $\varepsilon$  and we also have

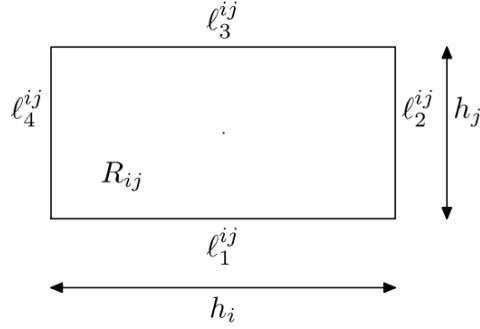


FIGURE 3. Notation

$\beta \geq 1$ , so a simple computation leaves

$$\begin{aligned} & \left| \int_{D_{2A}^2} \varepsilon^2 \nabla(u - \mathcal{Q}_1 u) \cdot (\nabla \beta - (\nabla \beta)_{min}) w \right| \\ & \leq Ch\varepsilon \|\varepsilon \beta^{\frac{1}{2}} \nabla(u - \mathcal{Q}_1 u)\|_{0,\Omega} \|\beta^{\frac{1}{2}} w\|_{0,\Omega} \\ & \leq Ch^2 \varepsilon \|w\|_{\beta}. \end{aligned}$$

Clearly, similar arguments can be used on  $D_2 \setminus (D_{2A}^1 \cup D_{2A}^2)$  to obtain

$$(5.27) \quad |III_{21}| \leq C\varepsilon^{-\eta h} h^2 \log\left(\frac{1}{\varepsilon}\right) \|w\|_{\beta}.$$

Now we have to estimate  $III_{22}$ . Let call  $(\nabla \beta)_{min,R_{ij}} = q_{ij} = (q_{ij}^1, q_{ij}^2)$ . Then we will estimate

$$\sum_{R_{ij} \subset D_2} \int_{R_{ij}} \varepsilon^2 q_{ij}^1 \partial_x(u - \mathcal{Q}_1 u) w.$$

We will follow a technique used in [22, 4]. Take into account Figure 3 for the notation of the sides of an element and its lengths. Let

$$K_{ij}(u, w) = \int_{R_{ij}} \partial_x(u - \mathcal{Q}_1 u) w - \frac{h_i^2}{12} \left( \int_{\ell_2^{ij}} (\partial_{xx} u) w dy - \int_{\ell_4^{ij}} (\partial_{xx} u) w dy \right).$$

Then we can write

$$\begin{aligned} & \sum_{R_{ij} \subset D_2} \int_{R_{ij}} \varepsilon^2 q_{ij}^1 \partial_x(u - \mathcal{Q}_1 u) w = \\ & \sum_{R_{ij} \subset D_2} \varepsilon^2 q_{ij}^1 K_{ij}(u, w) + \sum_{R_{ij} \subset D_2} \varepsilon^2 q_{ij}^1 \frac{h_i^2}{12} \left( \int_{\ell_2^{ij}} (\partial_{xx} u) w dy - \int_{\ell_4^{ij}} (\partial_{xx} u) w dy \right). \end{aligned}$$

From [4, eq. (3.14)] we know that

$$|K_{ij}(u, w)| \leq C (h_i^2 \|\partial_x^3 u\|_{0,R_{ij}} + h_i h_j \|\partial_x^2 \partial_y u\|_{0,R_{ij}} + h_j^2 \|\partial_x \partial_y^2\|_{0,R_{ij}}) \|w\|_{0,R_{ij}}$$

then, since

$$|q_{ij}| \leq |\nabla \beta| \leq C \frac{\beta}{\varepsilon} \leq C \varepsilon^{-\eta h} \frac{\beta_{\min}}{\varepsilon}$$

using Lemma 6.2 it follows that

$$(5.28) \quad \sum_{R_{ij} \subset D_2} \varepsilon^2 q_{ij}^1 K_{ij}(u, w) \leq C \varepsilon^{-\eta h} \times \\ \sum_{ij} \varepsilon \beta_{\min}^{\frac{1}{2}} (h_i^2 \|\partial_x^3 u\|_{0, R_{ij}} + h_i h_j \|\partial_x^2 \partial_y u\|_{0, R_{ij}} + h_j^2 \|\partial_x \partial_y^2\|_{0, R_{ij}}) \|\beta^{\frac{1}{2}} w\|_{0, R_{ij}} \\ \leq C \varepsilon^{-\eta h} h^2 \log\left(\frac{1}{\varepsilon}\right)^{\frac{1}{2}} \|\beta^{\frac{1}{2}} w\|_{0, \Omega}.$$

It remains to deal with

$$\mathfrak{E} := \sum_{R_{ij} \subset D_2} \varepsilon^2 q_{ij}^1 \frac{h_i^2}{12} \left( \int_{\ell_2^{ij}} (\partial_{xx} u) w dy - \int_{\ell_4^{ij}} (\partial_{xx} u) w dy \right)$$

which can be written as

$$\begin{aligned} \mathfrak{E} &= - \sum_{R_{ij} \subset D_2} \varepsilon^2 q_{ij}^1 \frac{h_i^2}{12} \int_{R_{ij}} \partial_x [(\partial_x^2 u) w] \\ &= - \sum_{R_{ij} \subset D_2} \varepsilon^2 q_{ij}^1 \frac{h_i^2}{12} \int_{R_{ij}} (\partial_x^3 u) w - \sum_{R_{ij} \subset D_2} \varepsilon^2 q_{ij}^1 \frac{h_i^2}{12} \int_{R_{ij}} \partial_x^2 u \partial_x w \\ &=: \mathfrak{E}_1 + \mathfrak{E}_2. \end{aligned}$$

Now we take into account that

$$|q_{ij}^1| \leq |\nabla \beta| \leq C \frac{\beta}{\varepsilon} \leq C \varepsilon^{-\eta h} \frac{\beta_{\min}}{\varepsilon}.$$

Then

$$|\mathfrak{E}_1| \leq C \varepsilon^{-\eta h} \sum_{R_{ij} \subset D_2} \left( \varepsilon \beta_{\min}^{\frac{1}{2}} h_i^2 \|\partial^3 u\|_{0, R_{ij}} \right) \left( \beta_{\min}^{\frac{1}{2}} \|w\|_{0, R_{ij}} \right).$$

Therefore, after applying Cauchy–Schwarz inequality and by using Lemma 6.2 it follows that

$$|\mathfrak{E}_1| \leq C \varepsilon^{-\eta h} \log\left(\frac{1}{\varepsilon}\right)^{\frac{1}{2}} h^2 \|w\| \|\beta\|.$$

Analogously we have

$$|\mathfrak{E}_2| \leq C \varepsilon^{-\eta h} \sum_{R_{ij} \subset D_2} \left( \beta_{\min}^{\frac{1}{2}} h_i^2 \|\partial_x^2 u\|_{0, R_{ij}} \right) \left( \varepsilon \beta_{\min}^{\frac{1}{2}} \|\partial_x w\|_{0, R_{ij}} \right).$$

With similar arguments we easily obtain

$$|\mathfrak{E}_2| \leq C\varepsilon^{-\eta h} h^2 \|w\|_\beta.$$

Then we arrived at

$$|\mathfrak{E}| \leq C\varepsilon^{-\eta h} \log\left(\frac{1}{\varepsilon}\right)^{\frac{1}{2}} h^2 \|w\|_\beta.$$

This inequality together with (5.28) give

$$\left| \sum_{R_{ij} \subset D_2} \int_{R_{ij}} \varepsilon^2 q_{ij}^1 \partial_x(u - \mathcal{Q}_1 u) w \right| \leq C\varepsilon^{-\eta h} \log\left(\frac{1}{\varepsilon}\right)^{\frac{1}{2}} h^2 \|w\|_\beta.$$

Clearly a similar argument allow us to conclude that

$$(5.29) \quad |III_{22}| \leq C\varepsilon^{-\eta h} \log\left(\frac{1}{\varepsilon}\right)^{\frac{1}{2}} h^2 \|w\|_\beta.$$

With inequalities (5.26), (5.27) and (5.29) we arrive at

$$(5.30) \quad |III| \leq C\varepsilon^{-\eta h} \log\left(\frac{1}{\varepsilon}\right)^{\frac{1}{2}} h^2 \|w\|_\beta.$$

**5.4. Acotación de IV.** From Proposition 4.1 we have

$$(5.31) \quad \begin{aligned} |IV| &\leq C \|\beta^{\frac{1}{2}}(u - \mathcal{Q}_1 u)\|_{0,\Omega} \|\beta^{\frac{1}{2}} w\|_{0,\frac{1}{2}} \\ &\leq Ch^2 \left(\log \frac{1}{\varepsilon}\right)^{\frac{1}{2}} \|w\|_\beta. \end{aligned}$$

**5.5. Proof of Theorem 5.1.** From (5.22), the splitting (5.23) with  $w = u_h - \mathcal{Q}_1 u \in V_h$  and the estimates (5.24), (5.25), (5.30) and (5.31) we obtain

$$\|u_h - \mathcal{Q}_1 u\|_\beta^2 \leq C\varepsilon^{-\eta h} \log\left(\frac{1}{\varepsilon}\right)^{\frac{1}{2}} h^2 \|u_h - \mathcal{Q}_1 u\|_\beta$$

from where the poof concludes.

## 6. NUMERICAL EXPERIMENTS

We consider the problem

$$(6.32) \quad -\varepsilon^2 \Delta u + u = f \quad \text{in } \Omega, \quad u = 0 \quad \text{on } \partial\Omega$$

on  $\Omega = [0, 1]^2$  with two different choices for the function  $f$ . The first one is taken from [5] and the second one was introduced by Kopteva [9] and is widely used in the literature (see, for example, [1, 16]). In both cases we take  $\varepsilon = 1e - 6$  and  $\varepsilon = 1e - 8$ . All the numerical results were computed



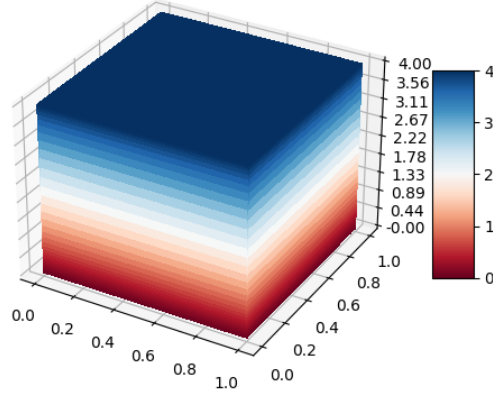


FIGURE 4. Solution of Example 6.1 with  $\varepsilon = 10^{-6}$ .

using Firedrake [20]. In Tables 1-5 we report the estimated order of convergence (eoc) of distinct quantities with respect to  $M$ , the number of grid points along  $x$  and  $y$  axis. We recall that the number of degrees of freedom is  $\sim M^2$ .

**Example 6.1.** Take  $f$  given by

$$f(x, y) = -2 \frac{1 - e^{-\frac{1}{\sqrt{2\varepsilon}}}}{1 - e^{-\frac{\sqrt{2}}{\varepsilon}}} \left( e^{-\frac{x}{\sqrt{2\varepsilon}}} + e^{-\frac{1-x}{\sqrt{2\varepsilon}}} + e^{-\frac{y}{\sqrt{2\varepsilon}}} + e^{-\frac{1-y}{\sqrt{2\varepsilon}}} \right) + 4.$$

By setting

$$u_0(t) = -2 \frac{1 - e^{-\frac{1}{\sqrt{2\varepsilon}}}}{1 - e^{-\frac{\sqrt{2}}{\varepsilon}}} \left( e^{-\frac{t}{\sqrt{2\varepsilon}}} + e^{-\frac{1-t}{\sqrt{2\varepsilon}}} \right) + 2$$

it follows that the exact solution  $u$  is

$$u(x, y) = u_0(x)u_0(y).$$

We report in Table 1 (resp. Table 2) the errors and convergence orders obtained using the discretization (2.2) with  $V_h$  being the space of piecewise bilinear functions on the graded meshes introduced in Section 3 with  $\varepsilon = 1e-6$  (resp.  $\varepsilon = 1e-8$ ).

**Example 6.2.** Now,  $f$  is chosen such that

$$u(x, y) = \left[ \cos\left(\frac{\pi}{2}x\right) - \frac{e^{-\frac{x}{\varepsilon}} - e^{-\frac{1}{\varepsilon}}}{1 - e^{-\frac{1}{\varepsilon}}} \right] \left( 1 - y - \frac{e^{-\frac{y}{\varepsilon}} - e^{-\frac{1}{\varepsilon}}}{1 - e^{-\frac{1}{\varepsilon}}} \right)$$

$h$	$M$	$\ u - u_h\ _0$	$eoc$	$\ u - u_h\ _\beta$	$eoc$	$\ u_I - u_h\ _\beta$	$eoc$
0.2	245	$5.5510e-5$	-	$1.2455e-1$	-	$2.2199e-2$	-
0.1	521	$1.6126e-5$	1.6384	$6.6549e-2$	0.8307	$6.3379e-3$	1.6614
0.05	1055	$4.3530e-6$	1.8561	$3.4499e-2$	0.9312	$1.7021e-3$	1.8633
0.03	1758	$1.6160e-6$	1.9406	$2.1014e-2$	0.9708	$6.3126e-4$	1.9425

TABLE 1. Report of errors for the numerical experiment of Example 6.1 with  $\varepsilon = 10^{-6}$ .

$h$	$M$	$\ u - u_h\ _0$	$eoc$	$\ u - u_h\ _\beta$	$eoc$	$\ u_I - u_h\ _\beta$	$eoc$
0.2	333	$5.6529e-6$	-	$1.2443e-01$	-	$2.2354e-2$	-
0.1	707	$1.6423e-6$	1.6418	$6.6517e-02$	0.8318	$6.3922e-3$	1.6628
0.05	1431	$4.4411e-7$	1.8547	$3.4493e-02$	0.9314	$1.7183e-3$	1.8632
0.03	2384	$1.6498e-7$	1.9401	$2.1013e-02$	0.9710	$6.3753e-4$	1.9425

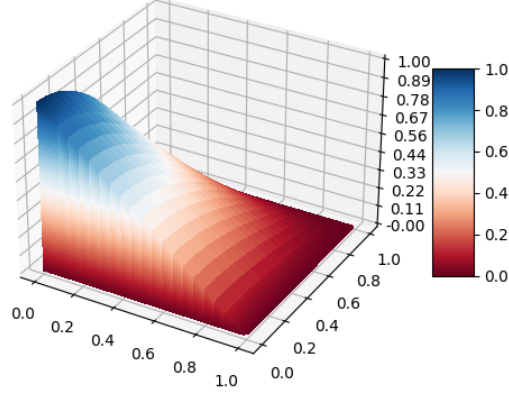
TABLE 2. Report of errors for the numerical experiment of Example 6.1 with  $\varepsilon = 10^{-8}$ .

$h$	$M$	$\ u - u_h\ _0$	$eoc$	$\ u - u_h\ _\beta$	$eoc$	$\ u_I - u_h\ _\beta$	$eoc$
0.2	245	$1.3125e-3$	-	$2.0775e-2$	-	$6.6800e-3$	-
0.1	521	$3.3423e-4$	1.8129	$1.1058e-2$	0.8358	$1.8082e-3$	1.7320
0.05	1055	$8.4654e-5$	1.9464	$5.7287e-3$	0.9322	$4.7234e-4$	1.9026
0.03	1758	$2.9457e-5$	2.0673	$3.4892e-3$	0.9710	$1.6900e-4$	2.0128

TABLE 3. Report of errors for Example 6.2 using graded meshes towards the entire boundary of  $\Omega$  with  $\varepsilon = 10^{-6}$ .

is the solution of (6.32). This solution exhibits boundary layers only along the sides  $x = 0$  and  $y = 0$ . In Table 3 (resp. Table 4) we report the convergence results obtained by using meshes graded towards the entire boundary  $\partial\Omega$  for  $\varepsilon = 1e - 6$  (resp.  $\varepsilon = 1e - 8$ ), and we note that the expected orders of convergence are observed. On the other hand, in Table 5 (resp. Table 6) we report the results obtained by grading the mesh only close to the boundary layers of the solution. In this case, we observe the correct order of convergence in  $\| \cdot \|_\beta$ , but the ones for the  $L^2$ -norm and the supercloseness are suboptimal. This curious behavior will be in the future subject of further investigation.

**Remark 6.1.** As we mentioned in the Introduction, it is desirable that graded meshes designed for a small value of  $\varepsilon$  work well for reaction–diffusion problems with larger values of the diffusion parameter. Although this fact is not included in our analysis, we show computationally that behaviour. As


 FIGURE 5. Solution of Example 6.2 for  $\varepsilon = 10^{-6}$ .

$h$	$M$	$\ u - u_h\ _0$	$eoc$	$\ u - u_h\ _\beta$	$eoc$	$\ u_I - u_h\ _\beta$	$eoc$
0.2	333	$1.2517e-3$	-	$2.0709e-2$	-	$6.4672e-3$	-
0.1	707	$3.2785e-4$	1.7794	$1.1032e-2$	0.8365	$1.7876e-3$	1.7079
0.05	1431	$8.3918e-5$	1.9327	$5.7167e-3$	0.9324	$4.7048e-4$	1.8932
0.03	2384	$2.9208e-5$	2.0678	$3.4823e-3$	0.9712	$1.6842e-4$	2.0126

 TABLE 4. Report of errors for Example 6.2 using graded meshes towards the entire boundary of  $\Omega$  with  $\varepsilon = 10^{-8}$ .

$h$	$M$	$\ u - u_h\ _0$	$eoc$	$\ u - u_h\ _\beta$	$eoc$	$\ u_I - u_h\ _\beta$	$eoc$
0.2	125	$4.6034e-2$	-	$7.9234e-2$	-	$7.6137e-2$	-
0.1	265	$1.9384e-2$	1.1511	$3.3988e-2$	1.1264	$3.2117e-2$	1.1487
0.05	537	$6.9052e-3$	1.4614	$1.2799e-2$	1.3828	$1.1448e-2$	1.4606
0.03	895	$3.2052e-3$	1.5024	$6.3562e-3$	1.3703	$5.3146e-3$	1.5024

 TABLE 5. Report of errors for Example 6.2 using graded meshes towards  $x = 0$  and  $y = 0$  with  $\varepsilon = 10^{-6}$ .

an example, Figure 6 exhibits the solution of Example 6.2 with  $\varepsilon = 1e - 3$  obtained using the graded mesh designed for  $h = 0.1$  and  $\varepsilon = 1e - 8$ . Using the same fixed graded mesh, Table 7 shows the errors obtained for  $\varepsilon$  varying between  $1e - 8$  and  $1e - 3$ . We observe that for all the values of diffusion parameter the errors are almost the same.

$h$	$M$	$\ u - u_h\ _0$	$eoc$	$\ u - u_h\ _\beta$	$eoc$	$\ u_I - u_h\ _\beta$	$eoc$
0.2	169	$4.5459e-2$	-	$7.8290e-2$	-	$7.5187e-2$	-
0.1	358	$1.9286e-2$	1.1423	$3.3825e-2$	1.1180	$3.1953e-2$	1.1400
0.05	725	$6.8893e-3$	1.4588	$1.2771e-2$	1.3804	$1.1423e-2$	1.4578
0.03	1208	$3.1980e-3$	1.5032	$6.3424e-3$	1.3708	$5.3026e-3$	1.5031

TABLE 6. Report of errors for Example 6.2 using graded meshes towards  $x = 0$  and  $y = 0$  with  $\varepsilon = 10^{-8}$ .

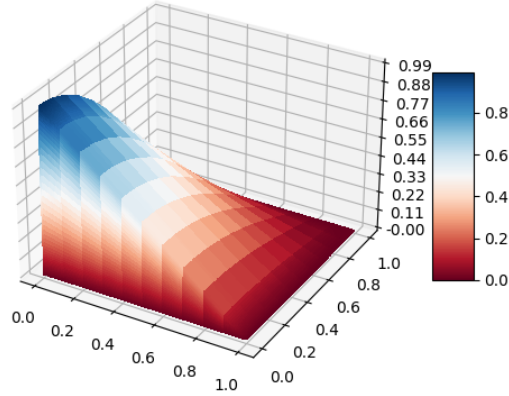


FIGURE 6. Solution of Example 6.2, for  $\varepsilon = 10^{-3}$ , obtained using the mesh designed with  $h = 0.1$  and  $\varepsilon = 10^{-8}$ .

$\varepsilon$	$\ u - u_h\ _0$	$\ u - u_h\ _\beta$	$\ u_I - u_h\ _\beta$
$1.0e-3$	$3.2820e-04$	$8.4978e-03$	$1.5048e-03$
$1.0e-4$	$3.2793e-04$	$8.7397e-03$	$1.5486e-03$
$1.0e-5$	$3.2786e-04$	$9.2645e-03$	$1.5964e-03$
$1.0e-6$	$3.2785e-04$	$9.8217e-03$	$1.6513e-03$
$1.0e-7$	$3.2785e-04$	$1.0410e-02$	$1.7146e-03$
$1.0e-8$	$3.2785e-04$	$1.1032e-02$	$1.7876e-03$

TABLE 7. Report of errors for the numerical experiment of Example 6.2 for distinct values of  $\varepsilon$  with a fixed mesh designed with  $h = 0.1$  and  $\varepsilon = 10^{-8}$ .

#### DECLARATIONS

**Funding.** This work was partially supported by Agencia Nacional de Promoción de la Investigación, el Desarrollo Tecnológico y la Innovación (Argentina) under grant PICT 2018-3017. Additionally, the first author (M.G.

Armentano) was supported by Universidad de Buenos Aires under grant 20020170100056BA, and the second and third authors (A.L. Lombardi and C. Penessi) were supported by Universidad Nacional de Rosario under grant 80020190100020UR.

**Competing interests.** The authors have no competing interests to declare that are relevant to the content of this article.

**Data availability.** Data sharing not applicable to this article as no datasets were generated or analysed during the current study.

## REFERENCES

- [1] J. Adler, S. MacLachlan, and N. Madden. A first-order system Petrov-Galerkin discretization for a reaction-diffusion problem on a fitted mesh. *IMA J. Numer. Anal.*, 36(3):1281–1309, 2016.
- [2] J Adler, S MacLachlan, and N Madden. First-Order System Least Squares Finite-Elements for Singularly Perturbed Reaction-Diffusion Equations. *Large-Scale Scientific Computing*, 11958:3–14, 2020.
- [3] Thomas Apel. *Anisotropic finite elements: Local estimates and applications*. Leipzig: Teubner; Chemnitz: Technische Univ., 1999.
- [4] R. G. Durán, A. L. Lombardi, and M. I. Prieto. Superconvergence for finite element approximation of a convection-diffusion equation using graded meshes. *IMA J. Numer. Anal.*, 32(2):511–533, 2012.
- [5] R. G. Durán, A. L. Lombardi, and M. I. Prieto. Supercloseness on graded meshes for  $\mathcal{Q}_1$  finite element approximation of a reaction-diffusion equation. *J. Comput. Appl. Math.*, 242:232–247, 2013.
- [6] Ricardo G. Durán and Ariel L. Lombardi. Error estimates on anisotropic  $\mathcal{Q}_1$  elements for functions in weighted Sobolev spaces. *Math. Comput.*, 74(252):1679–1706, 2005.
- [7] H. G. Roos and M. Schopf. Convergence and stability in balanced norms of finite element methods on Shishkin meshes for reaction-diffusion problems. *ZAMM Z. Angew. Math. Mech.*, 95(6):551–565, 2015.
- [8] S Gaucel and M Langlais. Some remarks on a singular reaction-diffusion system arising in predator-prey modeling. *Discrete Contin. Dyn. Syst. Ser. B*, 8(1):71–72, 2007.
- [9] N. Kopteva. Maximum norm a posteriori error estimate for a 2D singularly perturbed semilinear reaction-diffusion problem. *SIAM J Numer Anal*, 46(3):1602–1618, 2008.
- [10] J. Li and I. M. Navon. Uniformly convergent finite element methods for singularly perturbed elliptic boundary value problems. I: Reaction-diffusion type. *Comput. Math. Appl.*, 35(3):57–70, 1998.
- [11] Jichun Li and Mary F. Wheeler. Uniform convergence and superconvergence of mixed finite element methods on anisotropically refined grids. *SIAM J. Numer. Anal.*, 38(3):770–798, 2000.
- [12] R. Lin and M. Stynes. A balanced finite element method for singularly perturbed reaction-diffusion problems. *SIAM J. Numer. Anal.*, 50(5):2729–2743, 2012.
- [13] Torsten Linß. *Layer-adapted meshes for reaction-convection-diffusion problems*, volume 1985 of *Lect. Notes Math.* Berlin: Springer, 2010.

- [14] Fang Liu, Niall Madden, Martin Stynes, and Aihui Zhou. A two-scale sparse grid method for a singularly perturbed reaction-diffusion problem in two dimensions. *IMA J. Numer. Anal.*, 29(4):986–1007, 2009.
- [15] Ariel L. Lombardi. *Analysis of Finite Element Methods for Singularly Perturbed Problems*. PhD thesis, Universidad de Buenos Aires, 2004.
- [16] Madden, N. and Stynes, M. A weighted and balanced FEM for singularly perturbed reaction-diffusion problems. *Calcolo*, 58(2):1–16, 2021.
- [17] J. M. Melenk and C. Xenophontos. Robust exponential convergence of *hp*-FEM in balanced norms for singularly perturbed reaction-diffusion equations. *Calcolo*, 53(1):105–132, 2016.
- [18] J Mo and K Zhou. Singular perturbation for nonlinear species group reaction diffusion systems . *J. Biomath*, 21(4):481–488, 2006.
- [19] C Pao. Singular reaction diffusion equations of porous medium type. *Nonlinear Anal.*, 71(5-6):2033–2052, 2009.
- [20] Florian Rathgeber, David A. Ham, Lawrence Mitchell, Michael Lange, Fabio Luporini, Andrew T. T. McRae, Gheorghe-Teodor Bercea, Graham R. Markall, and Paul H. J. Kelly. Firedrake: automating the finite element method by composing abstractions. *ACM Trans. Math. Softw.*, 43(3):24:1–24:27, 2016.
- [21] Hans-Görg Roos, Martin Stynes, and Lutz Tobiska. *Robust numerical methods for singularly perturbed differential equations. Convection-diffusion-reaction and flow problems*, volume 24 of *Springer Ser. Comput. Math*. Berlin: Springer, 2nd ed. edition, 2008.
- [22] Zhimin Zhang. Finite element superconvergence on Shishkin mesh for 2-D convection-diffusion problems. *Math. Comput.*, 72(243):1147–1177, 2003.
- [23] M. Zlamal. Superconvergence and reduced integration in the finite element method. *Math.Comp.*, 32(143):663–685, 1978.

## APPENDIX

In this section we present some technical results which have been used along the paper.

The following Lemma is a consequence of [14, Lemmata 1.1 and 1.2]. In addition to the compatibility conditions of Section 4 we assume here that the fourth order derivatives of  $f$  and  $b$  are Hölder continuous up to the boundary. It is also assumed that  $b(x, y) \geq 2b_0^2$ .

**Lemma 6.1.** *Let  $u$  be the solution of (2.1). Then for all  $x \in (0, \frac{3}{4}) \times (0, \frac{3}{4})$  and  $k \leq 2$ , it holds*

$$\begin{aligned} \left| \partial_x \partial_y^k u(x, y) \right| &\leq C \left( 1 + \varepsilon^{1-k} \right) + \varepsilon^{-1} e^{-b_0 \frac{x}{\varepsilon}} + \varepsilon^{-k} e^{-b_0 \frac{y}{\varepsilon}} + \varepsilon^{-1-k} e^{-b_0 \frac{x+y}{\varepsilon}}, \\ \left| \partial_y \partial_x^k u(x, y) \right| &\leq C \left( 1 + \varepsilon^{1-k} \right) + \varepsilon^{-k} e^{-b_0 \frac{x}{\varepsilon}} + \varepsilon^{-1} e^{-b_0 \frac{y}{\varepsilon}} + \varepsilon^{-1-k} e^{-b_0 \frac{x+y}{\varepsilon}}. \end{aligned}$$

*Similar estimates are valid on the subdomains  $(0, \frac{3}{4}) \times (\frac{1}{4}, 1)$  (replace  $y$  by  $1 - y$ ),  $(\frac{1}{4}, 1) \times (0, \frac{3}{4})$  (replace  $x$  by  $(1 - x)$ ) and  $(\frac{1}{4}, 1) \times (\frac{1}{4}, 1)$  replace  $(x$  by  $1 - x$  and  $y$  by  $1 - y$ ).*

This Lemma allows us to obtain the next result.

**Lemma 6.2.** *Let  $u$  be the solution of (2.1). Then, under Assumption 1, we have that there exists a constant  $C$  such that*

$$(6.33) \quad \varepsilon \left[ \sum_{i,j} \beta_{min} (h_i^2 \|\partial_x^3 u\|_{0,R_{ij}} + h_i h_j \|\partial_x^2 \partial_y u\|_{0,R_{ij}} + h_j^2 \|\partial_x \partial_y^2 u\|_{0,R_{ij}})^2 \right]^{\frac{1}{2}} \leq C \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} h^2.$$

*Proof.* It is clear that by symmetry arguments it is enough to obtain (6.33) when the sum on the right hand side is restricted to the indices  $i, j$  with  $R_{ij} \subseteq \Omega_s := [0, \frac{1}{2}] \times [0, \frac{1}{2}]$ . Let us split  $\Omega_s$  as indicated in Figure 7. More precisely we set

$$\begin{aligned} \Lambda_0 &= (x_{\bar{m}}, \frac{1}{2}) \times (x_{\bar{m}}, \frac{1}{2}), & \Lambda_1 &= (x_{\bar{m}}, \frac{1}{2}) \times (x_1, x_{\bar{m}}), \\ \Lambda_2 &= (x_1, x_{\bar{m}}) \times (x_1, \frac{1}{2}), & \Lambda_3 &= (x_1, \frac{1}{2}) \times (0, x_1), \\ \Lambda_4 &= (0, x_1) \times (0, \frac{1}{2}), \end{aligned}$$

where  $x_{\bar{m}}$  is a grid point with  $x_{\bar{m}} = \gamma_0 \varepsilon \log \frac{1}{\varepsilon}$ . We use the notation

$$A(\Lambda_k) := \varepsilon \left[ \sum_{i,j:R_{ij} \subseteq \Lambda_k} \beta_{min} (h_i^2 \|\partial_x^3 u\|_{0,R_{ij}} + h_i h_j \|\partial_x^2 \partial_y u\|_{0,R_{ij}} + h_j^2 \|\partial_x \partial_y^2 u\|_{0,R_{ij}})^2 \right]^{\frac{1}{2}}.$$

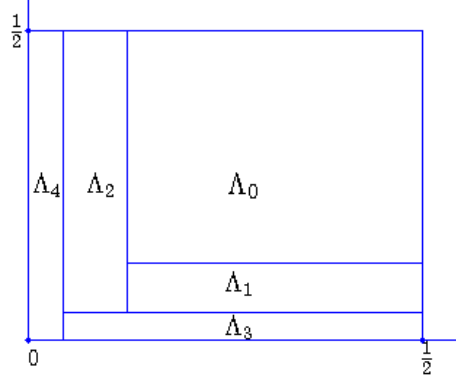
We will estimate separately  $A(\Lambda_k)$  for  $k = 0, \dots, 4$ .

- (0) Since  $\gamma_0 \geq \frac{2}{b_0}$  we have from Lemmata 4.1 and 6.1 that  $|D^3 u(x, y)| \leq C\varepsilon^{-1}$  and being  $\gamma_0 \geq \frac{1}{\gamma}$  we also have  $\beta_{min} \leq |\beta(x, y)| \leq C$  for all  $(x, y) \in \Lambda_0$ . Since  $h_i \leq h$  for all  $i$  easily arrive at

$$A(\Lambda_0) \leq Ch^2.$$

- (1) On  $\Lambda_1$  we also have  $\beta \leq C/\varepsilon$ . Taking into account that the length of  $\Lambda_1$  in the  $y$ -direction is  $\leq C\varepsilon \log \frac{1}{\varepsilon}$ ,  $h_i \leq h x^\alpha$  for  $(x, y) \in R_{ij} \subseteq \Lambda_1$ , and using Lemma 4.1 we have

$$(6.34) \quad \sum_{R_{ij} \subseteq \Lambda_1} \beta_{min} (h_i^2 \|\partial_x^3 u\|_{0,R_{ij}})^2 \leq C\varepsilon^{-1} h^4 \log \frac{1}{\varepsilon}.$$

FIGURE 7. Split of  $\Omega_s = [0, \frac{1}{2}]^2$  used in the proof of Lemma 6.2

Now we again have into account the estimate

$$(6.35) \quad |\partial_x^2 \partial_y u(x, y)| \leq C(1 + \varepsilon^{-1}) + \varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}} + \varepsilon^{-1} e^{-b_0 \frac{y}{\varepsilon}} + \varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}.$$

With the previous arguments, and in addition using that  $\gamma_0 \geq \frac{2}{b_0}$ , we have  $\varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}} \leq C$  on  $\Lambda_1$ ,  $h_i, h_j \leq h$ ,  $h_j \leq Chy^\alpha$  for  $(x, y) \in R_{ij} \subseteq \Lambda_1$ . Thus we obtain

$$\begin{aligned} \sum_{R_{ij} \subset \Lambda_1} \beta_{\min} (h_i h_j \| (1 + \varepsilon^{-1}) \|_{0, R_{ij}})^2 &\leq Ch^4 \varepsilon^{-2} \log \frac{1}{\varepsilon}, \\ \sum_{R_{ij} \subset \Lambda_1} \beta_{\min} (h_i h_j \| \varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}} \|_{0, R_{ij}})^2 &\leq Ch^4 \log \frac{1}{\varepsilon}, \\ \sum_{R_{ij} \subset \Lambda_1} \beta_{\min} (h_i h_j \| \varepsilon^{-1} e^{-b_0 \frac{y}{\varepsilon}} \|_{0, R_{ij}})^2 &\leq Ch^4, \\ \sum_{R_{ij} \subset \Lambda_1} \beta_{\min} (h_i h_j \| \varepsilon^{-3} e^{-b_0 \frac{(x+y)}{\varepsilon}} \|_{0, R_{ij}})^2 &\leq Ch^4. \end{aligned}$$

Then, together with (6.35) we arrive at

$$(6.36) \quad \sum_{R_{ij} \subset \Lambda_1} \beta_{\min} (h_i h_j \| \partial_x^2 \partial_y u \|_{0, R_{ij}})^2 \leq Ch^4 \varepsilon^{-2} \log \frac{1}{\varepsilon}.$$

Now, from Lemma 6.1 we further have

$$(6.37) \quad |\partial_x \partial_y^2 u(x, y)| \leq C(1 + \varepsilon^{-1}) + \varepsilon^{-1} e^{-b_0 \frac{x}{\varepsilon}} + \varepsilon^{-2} e^{-b_0 \frac{y}{\varepsilon}} + \varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}.$$



Now we use that  $h_j \leq h$ ,  $|S_4| \leq C\varepsilon \log \frac{1}{\varepsilon}$ ,  $\varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}} \leq C$  on  $\Lambda_1$  and  $h_j \leq hy^\alpha$  for  $(x, y) \in R_{ij} \subseteq \Lambda_1$  to obtain

$$\begin{aligned} \sum_{R_{ij} \subseteq \Lambda_1} \beta_{\min} \left( h_j^2 \|(1 + \varepsilon^{-1}) + \varepsilon^{-1} e^{-b_0 \frac{x}{\varepsilon}}\|_{0, R_{ij}} \right)^2 &\leq Ch^4 \varepsilon^{-2} \log \frac{1}{\varepsilon}, \\ \sum_{R_{ij} \subseteq \Lambda_1} \beta_{\min} \left( h_j^2 \|\varepsilon^{-2} e^{-b_0 \frac{y}{\varepsilon}}\|_{0, R_{ij}} \right)^2 &\leq Ch^4, \\ \sum_{R_{ij} \subseteq \Lambda_1} \beta_{\min} \left( h_j^2 \|\varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}\|_{0, R_{ij}} \right)^2 &\leq Ch^4 \varepsilon^2, \end{aligned}$$

which joint with (6.37) give

$$(6.38) \quad \sum_{R_{ij} \subseteq \Lambda_1} \beta_{\min} \left( h_j^2 \|\partial_x \partial_y^2 u\|_{0, R_{ij}} \right)^2 \leq Ch^4 \varepsilon^{-2} \log \frac{1}{\varepsilon}.$$

Inequalities (6.34), (6.36) and (6.38) leave

$$A(\Lambda_1) \leq C \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} h^2.$$

(2) On  $\Lambda_2$  we use that  $\beta \leq C/\varepsilon$ . In order to estimate  $A(\Lambda_2)$  we first note that since  $h_i \leq Chx^\alpha$  for  $(x, y) \in R_{ij} \subseteq \Lambda_2$  we have from Lemma 4.1 with  $k = 3$  that

$$(6.39) \quad \sum_{R_{ij} \subseteq \Lambda_2} \beta_{\min} h_i^4 \|\partial_x^3 u\|_{0, R_{ij}}^2 \leq Ch^4 \varepsilon^{-2}.$$

We use again (6.35) stated in Lemma 6.1. Using that for  $R_{ij} \subseteq \Lambda_2$  the inequalities  $h_i, h_j \leq Ch$ ,  $h_i \leq hx^\alpha$ ,  $h_j \leq hy^\alpha$  for  $(x, y) \in R_{ij}$ ,  $h_i \leq Ch\varepsilon \log \frac{1}{\varepsilon}$  and  $|\Lambda_2| \leq C\varepsilon \log \frac{1}{\varepsilon}$  hold true, it can be checked that

$$\begin{aligned} \sum_{R_{ij} \subseteq \Lambda_2} \beta_{\min} h_i^2 h_j^2 \|(1 + \varepsilon^{-1})\|_{0, R_{ij}}^2 &\leq C \left( \log \frac{1}{\varepsilon} \right)^3 h^4, \\ \sum_{R_{ij} \subseteq \Lambda_2} \beta_{\min} h_i^2 h_j^2 \|\varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}}\|_{0, R_{ij}}^2 &\leq C\varepsilon^{-2} h^4, \\ \sum_{R_{ij} \subseteq \Lambda_2} \beta_{\min} h_i^2 h_j^2 \|\varepsilon^{-1} e^{-b_0 \frac{y}{\varepsilon}}\|_{0, R_{ij}}^2 &\leq Ch^4, \\ \sum_{R_{ij} \subseteq \Lambda_2} \beta_{\min} h_i^2 h_j^2 \|\varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}\|_{0, R_{ij}}^2 &\leq C\varepsilon^{-1} h^4. \end{aligned}$$

Therefore we obtain

$$(6.40) \quad \sum_{R_{ij} \subseteq \Lambda_2} \beta_{\min} \left( h_i h_j \|\partial_x^2 \partial_y u\|_{0, R_{ij}} \right)^2 \leq Ch^4 \varepsilon^{-2}.$$

We use now the estimate (6.37). Then, using that for  $R_{ij} \subseteq \Lambda_2$  we have  $h_j \leq h$  and  $h_j \leq hy^\alpha$  for  $(x, y) \in R_{ij}$  and since  $|\Lambda_2| \leq C\varepsilon \log \frac{1}{\varepsilon}$  it follows

$$\begin{aligned} \sum_{R_{ij} \subset \Lambda_2} \beta_{min} \left( h_j^2 \| (1 + \varepsilon^{-1} + \varepsilon^{-1} e^{-b_0 \frac{x}{\varepsilon}}) \|_{0, R_{ij}} \right)^2 &\leq Ch^4 \varepsilon^{-2} \log \frac{1}{\varepsilon}, \\ \sum_{R_{ij} \subset \Lambda_2} \beta_{min} \left( h_j^2 \| \varepsilon^{-2} e^{-b_0 \frac{y}{\varepsilon}} \|_{0, R_{ij}} \right)^2 &\leq Ch^4, \\ \sum_{R_{ij} \subset \Lambda_2} \beta_{min} \left( h_j^2 \| \varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}} \|_{0, R_{ij}} \right)^2 &\leq C\varepsilon^{-2} h^4. \end{aligned}$$

It follows that

$$(6.41) \quad \sum_{R_{ij} \subset \mathcal{S}_1} \beta_{min} \left( h_j^2 \| \partial_x \partial_y^2 u \|_{0, R_{ij}} \right)^2 \leq Ch^4 \varepsilon^{-2} \log \frac{1}{\varepsilon}.$$

Collecting (6.39)–(6.41) we find

$$A(\Lambda_2) \leq C \left( \log \frac{1}{\varepsilon} \right)^{\frac{1}{2}} h^2.$$

- (3) We consider the estimate on  $\Lambda_3$ . We note that  $R_{11}$  is exterior to  $\Lambda_3$  and then we have  $h_i \leq hx^\alpha$  for all  $(x, y) \in R_{i1} \subseteq \Lambda_3$ . Since  $h \leq e^{-1}$  we have

$$h_1 = h^{2 \log \frac{1}{\varepsilon}} = \varepsilon^{2 \log \frac{1}{h}} < \varepsilon^2$$

and then we also have  $|\Lambda_3| \leq C\varepsilon^2$ . We will also use that  $\beta \leq \frac{C}{\varepsilon}$  on  $\Lambda_3$ . Then, from the estimate for  $\partial_x^3 u$  from Lemma 4.1 we have

$$(6.42) \quad \begin{aligned} \sum_{R_{i1} \subset \Lambda_3} \beta_{min} \left( h_i^2 \|_{0, R_{i1}} \partial_x^3 u \right)^2 \\ \leq Ch^4 \int_0^{h_1} \int_0^1 \left( 1 + \varepsilon^{-3} x^{2\alpha} e^{-b_0 \frac{x}{\varepsilon}} \right)^2 dx dy \leq Ch^4. \end{aligned}$$

Now we again take into account the estimate (6.35). Following the previous argument and since  $h_i \leq h$ ,  $h_1 \leq h\varepsilon$  we have

$$\begin{aligned} \sum_{R_{i1} \subset \Lambda_3} \beta_{min} \left( h_i h_1 \| 1 + \varepsilon^{-1} + \varepsilon^{-1} e^{-b_0 \frac{y}{\varepsilon}} \|_{0, R_{i1}} \right)^2 &\leq Ch^4 \varepsilon, \\ \sum_{R_{i1} \subset \Lambda_3} \beta_{min} \left( h_i h_1 \| \varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}} \|_{0, R_{i1}} \right)^2 &\leq Ch^4 \varepsilon^2, \end{aligned}$$

and since  $h_i \leq hx^\alpha$  for all  $(x, y) \in R_{i1} \subseteq \Lambda_3$  we also have

$$\sum_{R_{i1} \subset \Lambda_3} \beta_{min} \left( h_i h_1 \| \varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}} \| \right)^2 \leq Ch^4.$$

Thus we arrive at

$$(6.43) \quad \sum_{R_{i1} \subset \Lambda_3} \beta_{\min} (h_i h_1 \|\partial_x^2 \partial_y u\|_{0, R_{i1}})^2 \leq h^4.$$

On the other hand, we now use the estimate (6.37). Since again  $h_1 \leq h\varepsilon$  we obtain

$$\begin{aligned} \sum_{R_{i1} \subset \Lambda_3} \beta_{\min} \left( h_1^2 \|1 + \varepsilon^{-1} + \varepsilon^{-1} e^{-b_0 \frac{x}{\varepsilon}}\|_{0, R_{i1}} \right)^2 &\leq Ch^4 \varepsilon^3, \\ \sum_{R_{i1} \subset \mathcal{S}_8} \beta_{\min} \left( h_1^2 \|\varepsilon^{-2} e^{-b_0 \frac{y}{\varepsilon}}\|_{0, R_{i1}} \right)^2 &\leq Ch^4 \varepsilon. \end{aligned}$$

With all the previous arguments we also can check that

$$\sum_{R_{i1} \subset \Lambda_3} \beta_{\min} \left( h_1^2 \|\varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}\| \right)^2 \leq Ch^4 \varepsilon^4.$$

The last three inequalities give us

$$(6.44) \quad \sum_{R_{i1} \subset \Lambda_3} \beta_{\min} (h_1^2 \|\partial_x \partial_y^2 u\|_{0, R_{i1}})^2 \leq h^4 \varepsilon.$$

Finally, from (6.42)–(6.44) leave

$$A(\Lambda_3) \leq Ch^2 \varepsilon.$$

(4) Now, we consider the estimate on  $\Lambda_4$ . We note that

$$h_1 = h^{2 \log \frac{1}{\varepsilon}} = h^{\log \frac{1}{\varepsilon}} h^{\log \frac{1}{\varepsilon}} = h^{\log \frac{1}{\varepsilon}} \varepsilon^{\log \frac{1}{h}} \leq h\varepsilon.$$

Furthermore, as we proved in the previous item, we also have  $h_1 < \varepsilon^2$ , and as a consequence  $|\Lambda_4| \leq \varepsilon^2$ . Then, we can simply use that  $\partial_x^3 u \leq C\varepsilon^{-3}$ , which follows from Lemma 4.1 to obtain

$$(6.45) \quad \sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} (h_1^2 \|\partial_x^3 u\|_{0, R_{1j}})^2 \leq Ch^4 \varepsilon^{-1}.$$

Now, take into account again (6.35)

$$|\partial_x^2 \partial_y u(x, y)| \leq C(1 + \varepsilon^{-1}) + \varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}} + \varepsilon^{-1} e^{-b_0 \frac{y}{\varepsilon}} + \varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}.$$

We firstly note that, since  $h_j \leq h$ , we have

$$\begin{aligned} \sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_1 h_j \|(1 + \varepsilon^{-1} + \varepsilon^{-1} e^{-b_0 \frac{y}{\varepsilon}})\|_{0, R_{1j}} \right)^2 &\leq Ch^4 \varepsilon, \\ \sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_1 h_j \|\varepsilon^{-2} e^{-b_0 \frac{x}{\varepsilon}}\|_{0, R_{1j}} \right)^2 &\leq Ch^4 \varepsilon^{-1}. \end{aligned}$$

and secondly, since  $h_j \leq hy^\alpha$  for all  $(x, y) \in R_{1j} \subseteq \Lambda_4$ ,  $j \neq 1$  we have

$$\sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_1 h_j \|\varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}\|_{0, R_{1j}} \right)^2 \leq Ch^4.$$

From the last three inequalities we obtain

$$(6.46) \quad \sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_1 h_j \|\partial_x^2 \partial_y u\|_{0, R_{1j}} \right) \leq Ch^4 \varepsilon^{-1}.$$

Now we use the estimate (6.37)

$$|\partial_x \partial_y^2 u(x, y)| \leq C \left( 1 + \varepsilon^{-1} \right) + \varepsilon^{-1} e^{-b_0 \frac{x}{\varepsilon}} + \varepsilon^{-2} e^{-b_0 \frac{y}{\varepsilon}} + \varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}.$$

Since  $|\Lambda_4| \leq \varepsilon^2$  and  $h_j \leq h$  we have

$$\sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_j^2 \|(1 + \varepsilon^{-1} + \varepsilon^{-1} e^{-b_0 \frac{x}{\varepsilon}})\|_{0, R_{1j}} \right)^2 \leq Ch^4 \varepsilon^{-1}.$$

We also have

$$\begin{aligned} \sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_j^2 \|\varepsilon^{-2} e^{-b_0 \frac{y}{\varepsilon}}\|_{0, R_{1j}} \right)^2 &\leq Ch^4 \varepsilon^2, \\ \sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_j^2 \|\varepsilon^{-3} e^{-b_0 \frac{x+y}{\varepsilon}}\|_{0, R_{1j}} \right)^2 &\leq Ch^4, \end{aligned}$$

where we used again  $h_1 \leq \varepsilon^2$  and  $h_j \leq hy^\alpha$  for  $(x, y) \in R_{1j} \subseteq \Lambda_4$ ,  $j \neq 1$ . Then we obtain

$$(6.47) \quad \sum_{\substack{R_{1j} \subset \Lambda_4 \\ j \neq 1}} \beta_{\min} \left( h_j^2 \|\partial_x \partial_y^2 u\|_{0, R_{1j}} \right) \leq Ch^4 \varepsilon^{-1}.$$

Finally, since

$$|\partial_x^3 u|, |\partial_x^2 \partial_y u|, |\partial_x \partial_y^2 u| \leq C \varepsilon^{-3}$$

and using  $h_1 \leq h\varepsilon$  and  $h_1 \leq \varepsilon^2$ , and so  $|R_{11}| \leq \varepsilon^4$ , we obtain

$$(6.48) \quad \beta_{\min} h_1^4 \left( \|\partial_x^3 u\|_{0, R_{11}} + \|\partial_x^2 \partial_y u\|_{0, R_{11}} + \|\partial_x \partial_y^2 u\|_{0, R_{11}} \right)^2 \leq Ch^4 \varepsilon.$$

Therefore, inequalities (6.45)–(6.48) leave

$$A(\Lambda_4) \leq Ch^2 \varepsilon^{\frac{1}{2}}.$$

In this way we obtain (6.33) when the indices  $i, j$  are restricted to the ones for which  $R_{ij} \subset \Omega_s$ . The proof concludes by symmetry arguments.  $\square$   $\square$

ROBUST ESTIMATES FOR SINGULARLY PERTURBED REACTION DIFFUSION EQUATIONS

UNIVERSIDAD DE BUENOS AIRES. FACULTAD DE CIENCIAS EXACTAS Y NATURALES. DEPARTAMENTO DE MATEMÁTICA. PABELLÓN I – CIUDAD UNIVERSITARIA, 1428 BUENOS AIRES, ARGENTINA., CONICET – UNIVERSIDAD DE BUENOS AIRES. INSTITUTO DE INVESTIGACIONES MATEMÁTICAS “LUIS A. SANTALÓ” (IMAS). PABELLÓN I – CIUDAD UNIVERSITARIA, 1428 BUENOS AIRES, ARGENTINA.

*Email address:* garmenta@dm.uba.ar

UNIVERSIDAD NACIONAL DE ROSARIO. FACULTAD DE CIENCIAS EXACTAS, INGENIERÍA Y AGRIMENSURA. DEPARTAMENTO DE MATEMÁTICA. AV. PELLEGRINI 250, 2000 ROSARIO, ARGENTINA., CONICET, CCT ROSARIO, ARGENTINA.

*Email address:* ariel@fceia.unr.edu.ar

UNIVERSIDAD NACIONAL DE ROSARIO. FACULTAD DE CIENCIAS EXACTAS, INGENIERÍA Y AGRIMENSURA. DEPARTAMENTO DE MATEMÁTICA. AV. PELLEGRINI 250, 2000 ROSARIO, ARGENTINA., CONICET, CCT ROSARIO, ARGENTINA.

*Email address:* cecilia@fceia.unr.edu.ar