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# Iterative solution of elliptic problems by approximate factorization

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#### Abstract

An iterative method for the numerical solution of singularly perturbed second-order linear elliptic problems is presented. It is a defect correction iteration in which the approximate operator is the product of two first-order operators, which is readily inverted numerically. The approximate operator is generated by formal asymptotic factorization of the original operator. Hence this is a QUasi Analytic Defect correction iteration (QUAD). Both its continuous and discrete versions are analyzed in one dimension. The scheme is extended to a variety of two dimensional operators and it is analyzed for a model advection-diffusion equation. Numerical calculations show the effectiveness of the scheme over a wide range of values of the small parameter.

Keywords: Defect correction iteration; Asymptotic factorization; Preconditioners

AMS classification: 65F10; 65N22; 35J25

#### 1. Introduction

We develop an iterative method for the numerical solution of singularly perturbed linear elliptic problems of second order. It was originally designed for the computation of smooth solutions to amplitude equations [6, 7]. Our goal was to devise a stable numerical scheme which incorporates as much analytic information as possible.

The basis of the method is a defect correction iteration [9, 21]. This is a general technique for solving iteratively an equation Lu = f by means of an operator L which "approximates" L but is much easier to invert. In the present work, L is generated by a formal asymptotic factorization

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of the original operator into two first-order operators. The approximate operator is readily inverted numerically by solving a sequence of initial value problems. Hence the scheme can be classified as a QUasi Analytic Defect correction method (QUAD).

The algorithm combines the best of the asymptotic and the numerical approaches. On the one hand the use of asymptotic methods yields an approximate operator which is tailored to the problem. This yields a fast rate of convergence. On the other hand, the numerical implementation of the scheme eliminates the calculational burden associated with the derivation of a large number of terms in an asymptotic series. QUAD is a stable and convergent algorithm, in contrast to a standard asymptotic scheme which is unstable and ill-suited for numerical implementation, in general. Moreover, in the one-dimensional case the approximate operator can be inverted analytically and the scheme yields power series solutions to linear second-order singularly perturbed boundary value problems.

The present results are discussed in detail in [6]. Other papers that combine numerical and asymptotic methods include [3–5, 8, 15–20] and the collections [10, 11]. A third-order approximate factorization of the convection diffusion equation was presented in [14].

In Section 2 we motivate our work by presenting a one-dimensional model problem. Then in Section 3 we introduce the method in one dimension and we demonstrate it numerically on the model problem. The convergence analysis of both the continuous and the discrete iteration, in one dimension, is presented in Section 4. Variations on the themes of QUAD are presented in Section 5. In Section 6 we extend the scheme to a class of two-dimensional singularly perturbed elliptic problems. We analyze it for a model advection-diffusion equation and we present numerical calculations. We conclude with future directions in Section 7.

## 2. A model problem

We first introduce a one-dimensional model problem which arises in the hybrid numerical-asymptotic method for solving singularly perturbed equations in [6, 7]. We consider the propagation of a time harmonic wave of unit amplitude, traveling along the x axis to the right. Its motion is described by the one-dimensional Helmholtz equation

$$\varepsilon^2 u'' + n(x)^2 u = 0, (1)$$

subject to the conditions

$$u(0) = 1, \quad u \text{ outgoing at } x = +\infty.$$
 (2)

Here,  $\varepsilon = 1/k$  and k is the wave number. The index of refraction, n(x), is assumed positive in  $[0, \infty)$ .

The solution to (1) is highly oscillatory as  $\varepsilon \to 0$  and requires a very refined mesh for its numerical solution. Here, we seek solutions for u(x) in (1) of the form

$$u(x) = K(x)e^{iS(x)/\varepsilon},$$
(3)

where S(x) and K(x) are called the phase and the amplitude, respectively. Both functions are slowly varying and they can be resolved numerically on a grid which is substantially coarser than the one required for u.

In order to obtain equations for K and S we introduce expression (3) for u into (1) and we divide the resulting expression by  $e^{iS(x)/\epsilon}/\epsilon^2$  to obtain

$$K(n^2 - S_x^2) + i\varepsilon(2K_x S_x + KS_{xx}) + \varepsilon^2 K_{xx} = 0.$$
(4)

Upon equating the coefficient of the leading order term in (4) to zero we obtain the eiconal equation of geometrical optics

$$S_x(x)^2 = n(x)^2.$$
 (5)

Hence,  $S(x) = \int_0^x n(s) ds$ , in view of the outgoing condition at  $\infty$ . We substitute this expression for S(x) into (4) to obtain

$$\varepsilon K_{xx} + \mathbf{i}(2nK_x + n_xK) = 0. \tag{6}$$

Now we assume that n(x) is constant for  $x > 2 - \delta$  with  $\delta$  small. Then in view of the representation (3) and the outgoing condition we require  $K_x(2) = 0$ .

Our model problem is Eq. (6) for K subject to the conditions

$$K(0) = 1$$
 and  $K_x(2) = 0.$  (7)

In ray theory, for  $\varepsilon \ll 1$ , its solution is approximated by an asymptotic series. Instead we develop a convergent iteration to solve it numerically.

In summary, our approach to solving (1) numerically has two steps. First, we represent the solution as the product of a highly oscillatory factor and a smooth amplitude K. This step reduces dramatically the numerical complexity of solving the original problem because the number of grid points required to resolve K and S accurately is much smaller than the number of points required to resolve the original problem. The savings increase as  $\varepsilon \to 0$ . The second step is to develop a specialized numerical algorithm for solving the equation for K. Here, we focus on this aspect and we develop QUAD. Later [7] we shall discuss the first step, which is usually more complicated than in this one-dimensional example.

## 3. The QUAD algorithm

Let the operator L be defined by

$$L = \varepsilon \frac{d^2}{dx^2} + a_1(x)\frac{d}{dx} + a_0(x),$$
(8)

where  $0 < \varepsilon < 1$ ,  $a_0(x)$ ,  $a_1(x)$ , are continuous functions in [a, b],  $\operatorname{Re}[a_1(x)] \leq 0$  and  $a_1(x) \neq 0$  for  $x \in [a, b]$ . In this section we present the QUAD iteration for the problem

$$Lu = f, (9)$$

$$u(a) = 1, \qquad \left[\alpha \left(\frac{\mathrm{d}}{\mathrm{d}x} - A\right) + (1 - \alpha)\right] u(b) = \beta.$$
(10)

The case  $\alpha = 0$  and  $\alpha = 1$  in (10) corresponds to Dirichlet and robin boundary conditions at x = b, respectively. We assume that this problem has a unique solution.

#### 3.1. The basic iteration

Suppose that there exists a sequence of functions  $\{S_m(x,\varepsilon)\}_{m=1}^{\infty}$  such that the associated operators

$$L_m = \varepsilon \left( \frac{\mathrm{d}}{\mathrm{d}x} + \frac{a_1(x)}{\varepsilon} + S_m(x,\varepsilon) \right) \left( \frac{\mathrm{d}}{\mathrm{d}x} - S_m(x,\varepsilon) \right), \quad m = 1, 2, \dots$$
(11)

satisfy

$$L - L_m = \varepsilon^m r_m(x, \varepsilon), \tag{12}$$

and that

 $\lim_{\varepsilon \to 0} S_m(x,\varepsilon) = S_m(x,0), \tag{13}$ 

$$\lim_{\varepsilon \to 0} r_m(x,\varepsilon) = r_m(x,0). \tag{14}$$

The operators  $L_m$  possess two important properties. First they approximate L for  $\varepsilon \ll 1$ , as indicated by (12) and (14). Second, they are in factored form. Therefore the differential equation

$$L_m d = r, \tag{15}$$

can be solved explicitly, as we shall see in Section 3.2. In view of these properties we develop a defect correction iteration [9] to solve the problem (9), (10). It computes an approximation of u to an arbitrary tolerance  $\eta$ .

## Algorithm 1 (QUAD)

Compute the initial iterate  $u_0$  by solving

 $L_{m} u_{0} = f$   $u_{0}(a) = 1, \quad [\alpha(\frac{d}{dx} - A) + (1 - \alpha)]u_{0}(b) = \beta.$ For i = 1, 2, ...Compute the residual  $r_{i-1} = f - Lu_{i-1}.$ If  $||r_{i-1}||_{\infty}/||r_{0}||_{\infty} < \eta$ STOP
Else solve the following problem for the defect  $d_{i}$   $L_{m} d_{i} = r_{i-1}$   $d_{i}(a) = [\alpha(\frac{d}{dx} - A) + (1 - \alpha)]d_{i}(b) = 0.$ Compute the next iterate

$$u_i=u_{i-1}+d_i.$$

End

In the next section we present Algorithm 3 for the explicit solution of Eq. (15) for  $d_i$ , at each step of Algorithm 1. The existence and uniqueness of the solution  $d_i$  for all  $0 < \varepsilon < \varepsilon_0$  is guaranteed under a mild condition, as is discussed in Appendix A.

## 3.2. Solution of the approximate equation

Eq. (15) is readily solved with Algorithm 2 below when the boundary conditions are the following special case of (10):

$$\left(\frac{\mathrm{d}}{\mathrm{d}x} - S_m(b,\varepsilon)\right) d(b;\gamma) = \gamma, \quad d(a;\gamma) = 1.$$
(16)

## Algorithm 2

• Solve the initial value problem (17) for w:

$$\left(\frac{\mathrm{d}}{\mathrm{d}x} + \frac{a_1(x)}{\varepsilon} + S_m(x,\varepsilon)\right)w(x;\gamma) = \frac{r(x)}{\varepsilon}, \qquad w(b;\gamma) = \gamma.$$
(17)

• Solve the initial value problem (18) for d:

$$\left(\frac{\mathrm{d}}{\mathrm{d}x} - S_m(x,\varepsilon)\right) d(x;\gamma) = w(x;\gamma), \qquad d(a;\gamma) = 1.$$
(18)

Since the problem (15) and (16) is linear, and  $\gamma$  enters linearly in (16), it follows that

$$d(x;\gamma) = U(x) + \gamma V(x). \tag{19}$$

In order to solve Eq. (15) subject to the general condition (10), we use a technique analogous to the "shooting" method for first-order systems of differential equations (see [12, 13, 22]). We replace (10) by (16), where  $\gamma$  is to be determined so that (10) is satisfied. It follows that (10) will be satisfied if  $\gamma$  is a root of

$$\alpha\gamma - \beta + [\alpha(S_m - A) + (1 - \alpha)]d(b; \gamma) = 0.$$
<sup>(20)</sup>

To determine  $\gamma$  we use (19) in (20) to obtain

$$\{\alpha + [\alpha(S_m - A) + (1 - \alpha)]V(b)\}\gamma = \beta - [\alpha(S_m - A) + (1 - \alpha)]U(b).$$
(21)

If  $\alpha + [\alpha(S_m - A) + (1 - \alpha)]V(b) \neq 0$  there is a unique solution to this equation; otherwise there is no solution, unless the right-hand side is zero in which case there is an infinite number of solutions.

We summarize the scheme for the solution of problem (15), (10) in Algorithm 3 below.

## Algorithm 3

- (1) Evaluate d(x; 0) = U(x) using Algorithm 2.
- (2) Evaluate d(x; 1) = U(x) + V(x) using Algorithm 2.
- (3) Determine  $\gamma(\beta)$  from

$$\gamma(\beta) = \frac{\beta - [\alpha(S_m - A) + (1 - \alpha)]U(b)}{\alpha + [\alpha(S_m - A) + (1 - \alpha)]V(b)}.$$
(22)

(4) Determine  $d[x; \gamma(\beta)]$ 

$$d[x;\gamma(\beta)] = d(x;0) + \gamma(\beta)[d(x;1) - d(x;0)].$$
(23)

## 3.3. Approximate factorization of L

We now present a scheme for obtaining the approximate factorization of L. The approximation of order  $m \ge 0$  requires that  $a_0$  and  $a_1$  have m - 1 derivatives.

In analogy with the exact factorization of second-order operators (see [1, p. 22]), we write

$$\varepsilon \frac{\mathrm{d}^2}{\mathrm{d}x^2} + a_1(x) \frac{\mathrm{d}}{\mathrm{d}x} + a_0(x) = \varepsilon \left(\frac{\mathrm{d}}{\mathrm{d}x} + \frac{a_1(x)}{\varepsilon} + S(x,\varepsilon)\right) \left(\frac{\mathrm{d}}{\mathrm{d}x} - S(x,\varepsilon)\right),\tag{24}$$

where the function  $S(x, \varepsilon)$  is to be determined. We expand the right-hand side of this equation and subtract it from the left-hand side to obtain the following Riccati equation for S:

$$R(S,x) = 0, (25)$$

where

$$R(S,x) = \varepsilon S^2 + \varepsilon S_x + a_1(x)S + a_0(x).$$
<sup>(26)</sup>

Since we do not know how to solve (25) explicitly, we seek an asymptotic approximation to S of the form

$$S(x,\varepsilon) \sim \frac{1}{\varepsilon} \sum_{j=0}^{\infty} s_j \varepsilon^j \quad \text{as } \varepsilon \to 0.$$
 (27)

To determine the functions  $s_j$  we substitute (27) into (25), collect coefficients of equal powers of  $\varepsilon$  and equate the resulting coefficients of each power of  $\varepsilon$  to zero. This yields the following recursive system of algebraic equations for  $s_j$ :

$$a_1 s_0 + s_0^2 = 0, (28)$$

$$a_1s_1 + 2s_0s_1 + s_{0,x} + a_0 = 0, (29)$$

$$a_1s_m + 2s_0s_m + s_{m-1,x} + \sum_{k=1}^{m-1} s_ks_{m-k} = 0, \quad m \ge 2.$$
 (30)

This system is solved readily and yields

$$s_0 = 0$$
 or  $s_0 = -a_1(x)$ , (31)

$$s_1 = -\frac{s_{0,x} + a_0}{2},$$
(32)

$$a_1 + 2s_0$$
, (52)

$$s_m = -\frac{s_{m-1,x} + \sum_{k=1}^{m-1} s_k s_{m-k}}{a_1 + 2s_0}, \quad m \ge 2.$$
(33)

Now we define  $S_m(x,\varepsilon)$  by

$$S_m(x,\varepsilon) = \frac{1}{\varepsilon} \sum_{j=0}^m s_j(x)\varepsilon^j.$$
(34)

Upon substituting  $S_m$  for S in (25), we find that

$$R(S_m, x) = \varepsilon^m r_m(x, \varepsilon), \tag{35}$$

where

$$r_m(x,\varepsilon) = \sum_{j=1}^m \varepsilon^{j-1} \sum_{k=j}^m s_k s_{m+j-k} + s_{m,x}, \quad m \ge 1.$$
 (36)

It follows that the family of operators

$$L_m = \varepsilon \left(\frac{\mathrm{d}}{\mathrm{d}x} + \frac{a_1(x)}{\varepsilon} + S_m(x,\varepsilon)\right) \left(\frac{\mathrm{d}}{\mathrm{d}x} - S_m(x,\varepsilon)\right), \quad m \ge 1,$$
(37)

where  $S_m(x,\varepsilon)$  is defined by (34) with  $a_0 = 0$  satisfy the conditions (12)–(14).

## 3.4. Numerical calculations

We now present a few numerical calculations with QUAD for Eq. (6) of Section 2. Hence, we shall solve (9) in [0,2]. The operator L is defined in (8) with

$$a_1 = i2n(x), \qquad a_0 = in_x,$$
 (38)

and the index of refraction n(x) is chosen to be

$$n(x) = e^{-20(x-1)^2} + 1.$$
(39)

The boundary conditions are (16) with  $\gamma = 0$ .

In our calculations the mesh parameter h = 1/1000 and the order of the approximate operator is m = 1. The approximate operator is inverted with the trapezoidal method and the exact operator is discretized with second-order central differences. The small parameter  $\varepsilon$  is varied from one calculation to the other. Fig. 1 indicates the infinity norm of the error, i.e., the difference between the exact solution of the central difference discretization of problem (6), (7) and the QUAD iterate. We note that even for  $\varepsilon = 1$  the algorithm converges. As  $\varepsilon$  becomes smaller, the rate of convergence increases. The algorithm works well for  $\varepsilon > h$  and  $\varepsilon < h$ . To demonstrate further the stability of the scheme, we performed additional calculations in which 2500 iterations of QUAD where performed. The results are displayed in Fig. 2.

#### 4. Convergence analysis

#### 4.1. The continuous iteration

We shall first show that the continuous version of QUAD converges. We present the analysis of Algorithm 1 for problem (9), (10) with  $\alpha = 0$ . The analysis for other values of  $\alpha$  is analogous.

**Theorem 1.** Suppose that problem (9), (10) with  $\alpha = 0$  has a unique solution. Moreover suppose that  $L_m$  satisfies conditions (12)–(14) and that one of the existence conditions of Lemma 1 in

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Fig. 1. The ordinate is the  $\log_{10}$  of the infinity norm of the difference between the QUAD iterate and the solution to the finite difference discretization of (9). The abscissa is the number of iterations. The value of  $\varepsilon$  is indicated at the top of each graph.



Fig. 2. Long-term behavior of the QUAD algorithm. The ordinate is  $\log_{10}$  of the infinity norm of the difference between the QUAD iterate and the solution to the finite difference discretization of (9). The abscissa is the number of iterations. The value of  $\varepsilon$  is indicated at the top of each graph.

Appendix A is satisfied. Then there is an  $\varepsilon_0 > 0$  such that if  $\varepsilon < \varepsilon_0$ , Algorithm 1 converges and the error is reduced at each iteration by a factor which is  $O(\varepsilon^m)$ .

**Proof.** In order to determine the error  $e_i = u - u_i$  at step *i*, we substitute  $d_i = u_i - u_{i-1}$  and  $r_{i-1} = Lu - Lu_{i-1}$  in the equation for the defect  $d_i$ . Then we add and subtract  $L_m u$  from the left of that equation to obtain after some manipulation

$$L_m e_i = (L_m - L)e_{i-1}, (40)$$

$$e_i(a) = 0, \qquad e_i(b) = 0.$$
 (41)

Upon substituting (12) in (40) we obtain

$$L_m e_i(x) = -\varepsilon^m r_m(x,\varepsilon) e_{i-1}(x).$$
(42)

We compute the solution to (42) and (41) with Algorithm 3 to obtain

$$e_i(x) = \varepsilon^m \left( G_i(x,\varepsilon) - G_i(b,\varepsilon) \frac{H(x,\varepsilon)}{H(b,\varepsilon)} \right), \tag{43}$$

where the homogeneous and particular solutions, H and  $G_i$ , are given by

$$H(x,\varepsilon) = \int_{a}^{x} \exp\left(\int_{z}^{x} S_{m}(t,\varepsilon) dt + \int_{z}^{b} S_{m}(t,\varepsilon) + \int_{z}^{b} \frac{a_{1}(t)}{\varepsilon} dt\right) dz,$$
(44)

$$G_{i}(x,\varepsilon) = -\frac{1}{\varepsilon} \exp\left[\int_{a}^{x} S_{m}(t,\varepsilon) dt\right] \int_{a}^{x} \exp\left[-\int_{a}^{z} \frac{a_{1}(t)}{\varepsilon} + 2S_{m}(t,\varepsilon) dt\right]$$
$$\times \int_{b}^{z} \exp\left[\int_{a}^{y} \frac{a_{1}(t)}{\varepsilon} + S_{m}(t,\varepsilon) dt\right] r_{m}(y,\varepsilon)e_{i-1}(y) dy dz.$$
(45)

We will show that there exist constants  $C_1$ ,  $C_2$  and  $\eta$  such that for all  $\varepsilon > 0$ 

$$|G_i(x,\varepsilon)| \leqslant C_1 ||e_{i-1}||_{\infty},\tag{46}$$

and for all  $0 < \varepsilon < \eta$ 

$$|H(x,\varepsilon)/H(b,\varepsilon)| < C_2. \tag{47}$$

We use these bounds in (43) to obtain

$$\|e_{i}\|_{\infty} \leqslant \varepsilon^{m} C_{1}(1+C_{2})\|e_{i-1}\|_{\infty}.$$
(48)

It follows that the iteration converges if  $\varepsilon < \varepsilon_0$  where

$$\varepsilon_0 = \min\{\eta, (1/C_1(1+C_2))^{1/m}\}.$$
(49)

In order to derive the bound (46) we divide and multiply (45) by

$$a_1(z) + \varepsilon 2S_m(z,\varepsilon),$$

and then we integrate by parts to obtain

$$G_{i}(x,\varepsilon) = \frac{\exp\left[\int_{z}^{x} S_{m}(t,\varepsilon) \,\mathrm{d}t\right]}{a_{1}(z) + 2\varepsilon S_{m}(z,\varepsilon)} \int_{2}^{z} \exp\left[\int_{z}^{y} \frac{a_{1}(t)}{\varepsilon} + S_{m}(t,\varepsilon) \,\mathrm{d}t\right] r_{m}(y,\varepsilon) e_{i-1}(y) \,\mathrm{d}y \Big|_{0}^{x}$$
$$-\int_{0}^{x} \exp\left[\int_{z}^{x} S_{m}(t,\varepsilon) \,\mathrm{d}t\right] v(z) \,\mathrm{d}z, \tag{50}$$

where

$$v(z) = \frac{r_m(z,\varepsilon)e_{i-1}(z)}{a_1(z) + 2\varepsilon S_m(z,\varepsilon)} - \frac{a_{1,z}(z) + 2\varepsilon S_{m,z}(z)}{(a_1(z) + 2\varepsilon S_m(z,\varepsilon))^2} \times \int_2^z \exp\left[\int_z^y \frac{a_1(t)}{\varepsilon} + S_m(t,\varepsilon) dt\right] r_m(y,\varepsilon)e_{i-1}(y) dy.$$
(51)

The expression  $\exp[\int_z^y (a_1(t)/\varepsilon) dt]$  in (50) and (51) is uniformly bounded by 1 for all  $\varepsilon > 0$ , because  $z \le y$  and  $\operatorname{Re}[a_1] \le 0$ . We use this fact and relations (14) and (13) to obtain after some manipulation the bound (46).

In order to derive (47) we use (A.15) of Appendix A:

$$H(x,\varepsilon) \sim \varepsilon \left( \frac{f(a,x)}{a_1(a)} e^{\phi(a)/\varepsilon} - \frac{f(x,x)}{a_1(x)} e^{\phi(x)/\varepsilon} \right), \quad \varepsilon \to 0,$$
(52)

where f is defined in (A.16). Therefore

$$\left|\frac{H(x,\varepsilon)}{H(b,\varepsilon)}\right| \sim \left|\frac{f(a,x)\exp(\int_a^b \frac{a_1(t)}{\varepsilon} dt)/a_1(a) - f(x,x)\exp(\int_x^b \frac{a_1(t)}{\varepsilon} dt)/a_1(x)}{f(a,b)\exp(\int_a^b \frac{a_1(t)}{\varepsilon} dt)/a_1(a) - 1/a_1(b)}\right|, \quad \varepsilon \to 0.$$
(53)

The numerator in this expression is uniformly bounded for all  $\varepsilon$  because Re $[a_1] \leq 0$ . For  $\varepsilon$  sufficiently small the denominator is uniformly bounded away from 0 as indicated in Remark A.3 of Appendix A. Thus the right-hand side of (53) is bounded for  $\varepsilon$  sufficiently small. Then (53) yields the bound (47) and this completes the proof of Theorem 1.  $\Box$ 

For future reference we introduce the error amplification operator

$$M_m = L_m^{-1} (L_m - L), (54)$$

which determines the evolution of the error via the relation

$$e_i = M_m e_{i-1},\tag{55}$$

$$e_i(a) = 0, \quad e_i(b) = 0,$$
 (56)

in view of (40). Eq. (48) yields the bound

$$\|M_m\|_{\infty} < C\varepsilon^m,\tag{57}$$

where  $C = C_1(1 + C_2)$ . An explicit formula for C can be found in [6].

## 4.2. The discrete iteration

#### 4.2.1. Discretization

We next analyze the numerical implementation of Algorithm 1. We consider a mesh of N + 1 points:  $x_j = jh$ , j = 0, ..., N, h = (b - a)/N. We discretize the operator L in (8), with second-order finite difference schemes and we denote the discrete operator by  $L_h$ 

$$L_h = \varepsilon D_+ D_- + a_1(x_j) D_0 + a_0(x_j), \tag{58}$$

$$D_{+}D_{-}K_{j} = \frac{K_{j+1} - 2K_{j} + K_{j-1}}{h^{2}}, \quad D_{0}K_{j} = \frac{K_{j+1} - K_{j-1}}{2h}.$$
 (59)

The boundary conditions are incorporated in a standard fashion. At each step of Algorithm 1, we solve the first-order equations in Algorithm 2 using the trapezoidal or the backward Euler method [22]. Here we analyze the algorithm for an implementation with the trapezoidal method. The results for backward Euler are similar and may be found in [6].

The error in the discrete iteration evolves according to the relation

$$e_{h,j+1} = M_{m,h} e_{h,j}, (60)$$

where  $M_{m,h}$  is the discretization of  $M_m$  in (54). This relation follows from the continuous relation (55). To analyze relation (60) we introduce the modal representation

$$e_{h,j} = \sum_{\omega=0}^{N} \hat{e}_{\omega} \mathrm{e}^{\mathrm{i}2\pi\omega m/(N+1)},\tag{61}$$

into that equation and we study the expression

$$M_{h}e^{i2\pi\omega m/(N+1)},$$
(62)

for each mode of the grid  $\omega = 0, \ldots, n$ , separately.

Standard numerical analysis arguments show that  $M_h e^{i2\pi\omega m/(N+1)}$  is a good approximation to  $M e^{i2\pi\omega m/(N+1)}$  for modes  $\omega$  of the error satisfying the low frequency condition

$$\omega/N < \nu < 1, \tag{63}$$

where v is a small parameter. Hence the low frequency modes are damped by the discrete iteration in view of Theorem 1.

In the remainder of this section we evaluate (62) for the high-frequency modes. To simplify the analysis and assume that the coefficients of  $L_h$  and of  $L_{h,m}$  are constant, and that the boundary conditions are the homogeneous version of (16). Hence in the remainder of this section  $S_m$ ,  $a_1$  and

$$p = a_1 + \varepsilon S_m, \tag{64}$$

are constants.

4.2.2. Numerical stability

To evaluate (62) we first apply  $L_h$  to mode  $\omega$  and obtain that for all interior points of the grid

$$L_{h}e^{i2\pi\omega m/(N+1)} = \mu(\omega)e^{i2\pi\omega m/(N+1)},$$
(65)

(02)

where

$$\mu(\omega) = \varepsilon \lambda^2(\omega) + a_1 \lambda(\omega) \cos(\omega \pi/(N+1)) + a_0, \tag{66}$$

and

$$\lambda(\omega) = i\pi\omega \operatorname{sinc}\left(\frac{\omega}{N+1}\right)\frac{N}{N+1}.$$
(67)

Then, we apply  $L_h^{-1}$  to the right-hand side of (65). We perform this calculation explicitly using the discrete version of Algorithm 2. Indeed, the discretization of a differential equation by the trapezoidal method yields a first-order recurrence equation, which can be solved explicitly. We find that

$$M_{h} e^{i2\pi m\omega/(N+1)} = (1 - \mu c_{1}) e^{i2\pi \omega m/(N+1)} - \mu c_{2} \theta_{1}^{-m} + \mu (c_{1} + c_{2}) \theta_{2}^{m},$$
(68)

where

$$c_1(\omega) = \frac{\cos^2(\pi\omega/(N+1))}{\varepsilon\lambda^2(\omega) + a_1\lambda(\omega)\cos(\pi\omega/(N+1)) - pS\cos^2(\pi\omega/(N+1))},$$
(69)

$$c_2(\omega) = \left(\frac{1 + \frac{1}{2}\frac{hp}{\epsilon}}{1 - \frac{1}{2}\frac{hp}{\epsilon}}\right)^N \frac{e^{-i2\pi\omega/(N+1)}\cos(\pi\omega/(N+1))}{(a_1 + 2\epsilon S)\lambda(\omega) + (\frac{p^2}{\epsilon} + Sp)\cos(\pi\omega/(N+1))}.$$
(70)

The values of  $\theta_1$  and  $\theta_2$  are

$$\theta_1 = \frac{1 + \frac{1}{2}\frac{hp}{\varepsilon}}{1 - \frac{1}{2}\frac{hp}{\varepsilon}}, \quad \theta_2 = \frac{1 + \frac{1}{2}hS}{1 - \frac{1}{2}hS},$$
(71)

and p is defined in (64).

Expression (68) is a linear combination of three functions: mode  $\omega$ ,  $\theta_1^{-m}$  and  $\theta_2^m$ , the solutions to the discretized homogeneous form of (17) and (18), respectively. We analyze the magnitude of this linear combination for the high-frequency modes of the grid, for which the discrete iteration is not a good approximation to the continuous one. Specifically we consider this expression under two limits

- Limit I:  $h \rightarrow 0$  with  $\varepsilon$  fixed. This is the "standard" numerical analysis limit.
- Limit II: h→0 and h/ε→∞. Here we assume that the solution to the equation is smooth and that it is well resolved with a mesh parameter h≫ε.

In both limits we keep  $\omega/(N+1)$  fixed and therefore

$$|\lambda(\omega)| \to \infty. \tag{72}$$

Here  $\lambda(\omega)$  is defined in (67).

We use (69), (70) and (66) for  $c_1(\omega)$ ,  $c_2(\omega)$  and  $\mu(\omega)$ , respectively, in (68) to obtain for the trapezoidal method

$$1 - \mu c_1 = \frac{(\varepsilon \lambda^2 + a_1 \lambda \cos(\alpha))(1 - \cos^2(\alpha)) - (a_0 + pS)\cos^2(\alpha)}{\varepsilon \lambda^2 + a_1 \lambda \cos(\alpha) - pS\cos^2(\alpha)},$$
(73)

where

$$\alpha = \pi \omega / (N+1). \tag{74}$$

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In the limit  $h \rightarrow 0$ , we find that

$$\lim_{h \to 0} 1 - \mu c_1 = 1 - \cos^2(\pi \omega / (N+1)), \tag{75}$$

in view of relation (72). Hence, the coefficient of  $e^{i2\pi m\omega/(N+1)}$  in (68) satisfies

$$|1 - \mu c_1| \leqslant 1. \tag{76}$$

We now analyze the remaining terms in (68):  $\theta_1^m$  and  $\theta_2^m$  where  $\theta_i$  are defined in (71). Limit 1: With  $\varepsilon$  fixed,

$$\lim_{h \to 0} \theta_1^{-m} = e^{-px/\epsilon},$$
(77)
$$\lim_{h \to 0} \theta_2^m = e^{Sx}.$$
(78)

Both of these terms are *smooth* in the sense that they are resolved by the numerical scheme and will be damped by additional iterations, in view of Theorem 1.

*Limit* 2: The term  $\theta_2^m$  converges to the limit (78) which is smooth and therefore is of no concern. We evaluate the coefficient of  $\theta_1^{-m}$  using (66) and (70) to obtain after some manipulation

$$\mu c_2 = \frac{\theta_1^N \mathrm{e}^{-i2\pi\omega/(N+1)} \cos(\pi\omega/(N+1))\varepsilon\lambda(\varepsilon\lambda + a_1\cos(\pi\omega/(N+1)) + a_0/\lambda)}{(a_1 + 2\varepsilon S)\varepsilon\lambda + (p^2 + \varepsilon Sp)\cos(\pi\omega/(N+1))}.$$
(79)

If  $\cos(\pi\omega/(N+1)) = 0$ , then  $\mu c_2 = 0$ . Otherwise, we use the relations

$$a_1 \cos(\pi \omega / (N+1)) \gg a_0 / \lambda, \quad \lambda \to \infty,$$
(80)

$$a_1 \gg 2\varepsilon S$$
 and  $p^2 + \varepsilon Sp \sim a_1^2$ ,  $\varepsilon \to 0$  (81)

in the numerator and denominator of (79) to obtain that for large  $\lambda$  and small  $\varepsilon$ 

$$|c_2\mu| \sim \left|\frac{\theta_1^N \cos(\pi\omega/(N+1))\epsilon\lambda}{a_1}\right|.$$
(82)

This expression is arbitrarily small because  $\epsilon \lambda \to 0$  as  $h \to 0$ . We also note that for the highest modes of the grid,  $\omega \approx (N+1)/2$ ,  $\cos(\pi \omega/(N+1)) \approx 0$ .

In summary, under the first limit, expression (73) consists of mode  $\omega$  with a coefficient smaller than one and an additional term which is smooth and therefore is damped by additional iterations of the scheme, in view of Theorem 1. Under the second limit, (73) consists of mode  $\omega$  with a coefficient smaller than one, a smooth term which is damped by additional iterations of the scheme, and a high-frequency term. The high-frequency term is of small magnitude and will be damped by additional iterations, in view of the analysis above. Hence the numerical implementation of QUAD is a stable numerical algorithm.

#### 5. Exact factorization of an approximate operator

In Section 3.3 we derived  $L_m$  by approximately factoring L. Here we choose  $L_m$  such that it can be factored exactly. We do this to solve problems locally, in the neighborhood of singular points or

singular lines, such as turning points or caustics. This approach also works when there is no small parameter in the equation.

In order to illustrate the scheme we return to the model equation of Section 2:

$$\varepsilon u'' + n(x)u = 0, \tag{83}$$

in  $[-\delta, \delta]$  subject to

$$u(-\delta) = \alpha, \quad u(\delta) = \beta. \tag{84}$$

We denote the operator in this equation by L and we assume that n(0) = 0 is the only zero of n in  $[-\delta, \delta]$ . Moreover we assume that the solution

Ai 
$$\left( \left[ \frac{n'(0)}{\varepsilon} \right]^{1/3} x \right)$$

to the Airy equation

$$\varepsilon u'' + n'(0)xu = 0, \tag{85}$$

is nonzero in  $[-\delta, \delta]$ .

The representation (3) for u is not valid in a neighborhood of 0 because the nature of u changes from oscillatory to exponentially decaying there (see [1]). Hence in a small neighborhood of 0 we solve (83) directly by iteration. Indeed, the operator in (85) can be factored exactly as follows (see discussion in [1, p. 22])

$$\varepsilon \frac{\mathrm{d}^2}{\mathrm{d}x^2} + n'(0)x = \varepsilon \left(\frac{\mathrm{d}}{\mathrm{d}x} + \left[\frac{n'(0)}{\varepsilon}\right]^{1/3}\frac{\mathrm{Ai}'}{\mathrm{Ai}}\right) \left(\frac{\mathrm{d}}{\mathrm{d}x} - \left[\frac{n'(0)}{\varepsilon}\right]^{1/3}\frac{\mathrm{Ai}'}{\mathrm{Ai}}\right). \tag{86}$$

We denote this operator by  $\hat{L}$  and we note that Algorithm 1 with  $\hat{L}$  substituted for  $L_m$  converges to the solution of (83), provided  $\delta$  is sufficiently small. This can be shown with an analysis analogous to the one used in the proof of Theorem 1, in view of the fact that

$$L - \hat{L} = n(x) - n'(0)x = n''(\xi)x^2/2, \quad \xi, x \in [-\delta, \delta].$$

The convergence factor is  $O(\delta^2)$  and it does not depend on  $\varepsilon$ .

## 6. Generalization to higher dimensions

#### 6.1. A formal factorization scheme

We now describe a formal scheme for the approximate factorization of a class of singularly perturbed partial differential operators. This scheme is a natural generalization of the one-dimensional scheme of Section 3.3. We consider operators of the form

$$L = \varepsilon \left( \frac{\partial^2}{\partial \sigma^2} + b_1 \frac{\partial^2}{\partial \beta \partial \sigma} + b_0 \frac{\partial^2}{\partial \beta^2} + c_1 \frac{\partial}{\partial \sigma} + c_0 \frac{\partial}{\partial \beta} \right) + a_1 \frac{\partial}{\partial \sigma} + a_0, \tag{87}$$

in a domain  $\Omega \in \mathbb{R}^2$ . The coefficients in (87) are functions of  $\sigma$  and  $\beta$  and may be complex valued, and  $a_1 \neq 0$ . To formally factor (87) we rewrite this operator emphasizing the  $\sigma$  direction

$$L = \varepsilon \left( \frac{\partial^2}{\partial \sigma^2} + \left( \frac{a_1}{\varepsilon} + c_1 + b_1 \frac{\partial}{\partial \beta} \right) \frac{\partial}{\partial \sigma} + \left( \frac{a_0}{\varepsilon} + b_0 \frac{\partial^2}{\partial \beta^2} + c_0 \frac{\partial}{\partial \beta} \right) \right).$$
(88)

Now in analogy with the one-dimensional case we seek an approximate factorization of L of the form

$$L = \varepsilon \left( \frac{\partial}{\partial \sigma} + \left( \frac{a_1}{\varepsilon} + c_1 + b_1 \frac{\partial}{\partial \beta} \right) + S \right) \left( \frac{\partial}{\partial \sigma} - S \right).$$
(89)

In (89) we have introduced the unknown operator  $S(\sigma, \beta, \varepsilon, \partial/\partial \beta)$ .

We expand the right-hand side of (89) and find that it is equal to

$$L-\varepsilon R,$$
 (90)

were the residual R is given by

$$R = b_0 \frac{\partial^2}{\partial \beta^2} + c_0 \frac{\partial}{\partial \beta} + \frac{a_0}{\varepsilon} + S_\sigma + b_1 S_\beta + \left(c_1 + \frac{a_1}{\varepsilon}\right) S + b_1 S \frac{\partial}{\partial \beta} + S^2.$$
(91)

The expressions  $S_{\sigma}$  and  $S_{\beta}$  in (91) represent the operators obtained by differentiating the coefficients of S by  $\sigma$  and  $\beta$ , respectively. We now equate R to 0 and obtain an equation for S. This equation is a quadratic in S with first-order derivatives and it is similar to the Riccati equation for the function S of Section 3.3.

We construct a formal asymptotic expansion in powers of  $\varepsilon$  for S of the form

$$S \sim \frac{1}{\varepsilon} \sum_{j=0}^{\infty} s_j \left( \sigma, \beta, \frac{\partial}{\partial \beta} \right).$$
(92)

The operators  $s_j(\sigma, \beta, \partial/\partial \beta)$  are linear operators in  $\partial^n/\partial \beta^n$ . Upon substituting (92) for S into (91), collecting terms with equal powers of  $\varepsilon$  and multiplying by  $\varepsilon^2$  we obtain

$$R = \varepsilon^{2} \left( b_{0} \frac{\partial^{2}}{\partial \beta^{2}} + c_{0} \frac{\partial}{\partial \beta} \right) + \varepsilon a_{0}$$
$$+ \sum_{j=0}^{\infty} \left( s_{j-1,\sigma} + b_{1} s_{j-1,\beta} + b_{1} s_{j-1} \frac{\partial}{\partial \beta} + c_{1} s_{j-1} + a_{1} s_{j} + \sum_{k=0}^{j} s_{k} s_{j-k} \right) \varepsilon^{j}.$$
(93)

Then we equate the coefficient of each power of  $\varepsilon$  to 0 and obtain the following recursive system of equations for  $s_i$ :

$$a_1 s_0 + s_0^2 = 0, (94)$$

$$a_{1}s_{1} + s_{0}s_{1} + s_{1}s_{0} = -\left(a_{0} + s_{0,\sigma} + b_{1}s_{0,\beta} + b_{1}s_{0}\frac{\partial}{\partial\beta} + c_{1}s_{0}\right),$$
(95)

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$$a_{1}s_{2} + s_{0}s_{2} + s_{2}s_{0} = -\left(s_{1}^{2} + b_{0}\frac{\partial^{2}}{\partial\beta^{2}} + c_{0}\frac{\partial}{\partial\beta} + s_{1,\sigma} + b_{1}s_{1,\beta} + b_{1}s_{1}\frac{\partial}{\partial\beta} + c_{1}s_{1}\right),$$
(96)

$$a_{1}s_{j} + s_{0}s_{j} + s_{j}s_{0} = -\left(s_{j-1,\sigma} + b_{1}s_{j-1,\beta} + b_{1}s_{j-1}\frac{\partial}{\partial\beta} + c_{1}s_{j-1} + \sum_{k=1}^{j-1}s_{k}s_{j-k}\right).$$
(97)

Eq. (94) is readily solved and yields

$$s_0 = 0$$
 or  $s_0 = -a_1$ . (98)

Therefore, as in the one-dimensional case, there are two possible solutions and we denote them by Case 1 and 2, respectively. We shall base our factorization on the solution with  $s_0 = 0$ . However, for the sake of completeness we shall also obtain the second solution in Appendix B. A solution of (94)-(97) is

Case 1

$$s_0 = 0,$$

$$s_1 = -a_0/a_1,$$
 (99)

$$s_2 = -\frac{b_0}{a_1}\frac{\partial^2}{\partial\beta^2} + \left(\frac{a_0b_1}{a_1^2} - \frac{c_0}{a_1}\right)\frac{\partial}{\partial\beta} + d(\sigma,\beta),\tag{100}$$

where  $d(\sigma, \beta)$  is defined by

$$d(\sigma,\beta) = (a_1 a_{0,\sigma} - a_0 a_{1,\sigma} + b_1 (a_1 a_{0,\beta} - a_0 a_{1,\beta}) + c_1 a_0 a_1 - a_0^2) / a_1^3,$$
(101)

$$s_{j} = -\frac{1}{a_{1}} \left( s_{j-1,\sigma} + b_{1} s_{j-1,\beta} + b_{1} s_{j-1} \frac{\partial}{\partial \beta} + c_{1} s_{j-1} + \sum_{k=1}^{j-1} s_{k} s_{j-k} \right).$$
(102)

## 6.2. Application to the steady-state convection diffusion equation

We now apply (99)-(102) to the steady-state convection diffusion equation with sub-characteristics parallel to the x axis

$$\varepsilon \left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2}\right) u - a(x, y) \frac{\partial u}{\partial x} = f.$$
(103)

To make (103) agree with (87), we identify x with  $\sigma$  and y with  $\beta$ . We find that

$$a_0 = 0,$$
  $b_1 = 0,$   $c_1 = 0,$   $c_0 = 0,$   $a_1 = -a(x, y),$   $b_0 = 1.$  (104)

We introduce these coefficients in (99)-(102) to obtain

$$s_0 = 0, \qquad s_1 = 0, \qquad s_2 = \frac{1}{a} \frac{\partial^2}{\partial y^2}, \qquad s_3 = -\frac{a_x}{a^3} \frac{\partial^2}{\partial y^2}.$$
 (105)

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Then

$$L_2 = \varepsilon \left(\frac{\partial}{\partial x} - \frac{a}{\varepsilon} + \frac{\varepsilon}{a}\frac{\partial^2}{\partial y^2}\right) \left(\frac{\partial}{\partial x} - \frac{\varepsilon}{a}\frac{\partial^2}{\partial y^2}\right),\tag{106}$$

$$L_{3} = \varepsilon \left(\frac{\partial}{\partial x} - \frac{a}{\varepsilon} + \left(\frac{\varepsilon}{a} - \frac{\varepsilon^{2} a_{x}}{a^{3}}\right) \frac{\partial^{2}}{\partial y^{2}}\right) \left(\frac{\partial}{\partial x} - \left(\frac{\varepsilon}{a} - \frac{\varepsilon^{2} a_{x}}{a^{3}}\right) \frac{\partial^{2}}{\partial y^{2}}\right).$$
(107)

For  $m \ge 2$  we find that

$$L_m - L = -R_m, \tag{108}$$

where

$$R_m = \varepsilon^m \left( s_{m,x} + \sum_{j=1}^m \left( \sum_{k=j}^m s_k s_{m+j-k} \right) \varepsilon^{j-1} \right).$$
(109)

Therefore

$$R_2 = -\varepsilon^2 \frac{a_x}{a^2} \frac{\partial^2}{\partial y^2} + \frac{\varepsilon^3}{a} \frac{\partial^2}{\partial y^2} \frac{1}{a} \frac{\partial^2}{\partial y^2},\tag{110}$$

$$R_{3} = \varepsilon^{3} \left( \frac{1}{a} \frac{\partial^{2}}{\partial y^{2}} \frac{1}{a} \frac{\partial^{2}}{\partial y^{2}} - \left( \frac{a_{x}}{a^{3}} \right)_{x} \frac{\partial^{2}}{\partial y^{2}} \right) - \varepsilon^{4} \left( \frac{1}{a} \frac{\partial^{2}}{\partial y^{2}} \frac{a_{x}}{a^{3}} \frac{\partial^{2}}{\partial y^{2}} + \frac{a_{x}}{a^{3}} \frac{\partial^{2}}{\partial y^{2}} \frac{1}{a} \frac{\partial^{2}}{\partial y^{2}} \right) + \varepsilon^{5} \frac{a_{x}}{a^{3}} \frac{\partial^{2}}{\partial y^{2}} \frac{a_{x}}{a^{3}} \frac{\partial^{2}}{\partial y^{2}}.$$
(111)

The results (107), (111) agree with a third-order approximate factorization of the convection diffusion operator, described in [14]. That factorization is obtained by modifying one for the constant case by a method designed for the convection diffusion operator. The present method gives factorizations of all orders.

#### 6.3. Convergence analysis

We now analyze the convergence of QUAD applied to (103) in  $[0,1] \times [0,1]$  with  $L_2$  given by the second-order approximation (106). Here we consider the special case in which the coefficient a(x, y) in (103) depends on x alone and is continuously differentiable. We also assume Dirichlet boundary conditions at x = 0, y = 0 and y = 1 and the special condition

$$\left(\frac{\partial}{\partial x} - \frac{\varepsilon}{a}\frac{\partial^2}{\partial y^2}\right)u = 0, \tag{112}$$

at x = 1.

The evolution of the error is determined by the relation

$$L_2 e_{j+1} = (L_2 - L) e_j, (113)$$

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$$e_{j+1}(x,0) = 0, \qquad e_{j+1}(x,1) = 0,$$
(114)

$$e_{j+1}(0, y) = 0, \qquad \left(\frac{\partial}{\partial x} - \frac{\varepsilon}{a}\frac{\partial^2}{\partial y^2}\right)e_{j+1}(1, y) = 0.$$
(115)

Relation (113) is derived in the same way as relation (40).

We now use (108) in (113) and we solve the resulting problem (113)-(115) in two steps. First we solve

$$\varepsilon \left(\frac{\partial}{\partial x} - \frac{a}{\varepsilon} + \frac{\varepsilon}{a}\frac{\partial^2}{\partial y^2}\right) w(x, y) = -R_2 e_j(x, y), \qquad (116)$$

$$w(1, y) = w(x, 0) = w(x, 1) = 0,$$
(117)

for w and then we solve

$$\left(\frac{\partial}{\partial x} - \frac{\varepsilon}{a}\frac{\partial^2}{\partial y^2}\right)e_{j+1}(x,y) = w(x,y),\tag{118}$$

$$e_{j+1}(0, y) = e_{j+1}(x, 0) = e_{j+1}(x, 1) = 0,$$
(119)

for  $e_{j+1}$ .

We use (110) for  $R_2$  in problem (116), (117) which we then solve to obtain

$$w(x, y) = \sum_{\omega=1}^{\infty} \hat{w}(x, \omega) \sqrt{2} \sin(\omega \pi y), \qquad (120)$$

where

$$\hat{w}(x,\omega) = \int_{1}^{x} \exp\left(-\int_{x}^{t} p(s,\omega) \,\mathrm{d}s\right) \frac{\hat{r}(t,\omega)}{\varepsilon} \,\mathrm{d}t,\tag{121}$$

$$p(s,\omega) = \frac{a(s)}{\varepsilon} + \frac{\varepsilon}{a(s)} [\omega\pi]^2, \qquad (122)$$

$$\hat{r}(t,\omega) = -\left(\frac{\varepsilon\omega\pi}{a(t)}\right)^2 [a_x(t) + \varepsilon(\omega\pi)^2]\hat{e}_j(t,\omega), \qquad (123)$$

and  $\hat{e}_j(x,\omega)$  is the  $\omega$  Fourier coefficient of  $e_j(x,y)$ 

$$e_j(x,y) = \sum_{\omega=1}^{\infty} \hat{e}_j(x,\omega) \sqrt{2} \sin(\omega \pi y).$$
(124)

We now define  $r^*$ ,  $p^*$ ,  $x_p$ , and  $x_r$  as

$$r^* = \max_{x \in [0,1]} |\hat{r}(x,\omega)| = \left| \left( \frac{\varepsilon \omega \pi}{a(x_r)} \right)^2 [a_x(x_r) + \varepsilon(\omega \pi)^2] \hat{e}(x_r,\omega) \right|,$$
(125)

$$p^* = \min_{x \in [0,1]} p(x,\omega) = \frac{a(x_p)}{\varepsilon} + \frac{\varepsilon}{a(x_p)} [\omega\pi]^2.$$
(126)

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Then we manipulate (121) to obtain

$$|\hat{w}(x,\omega)| \leq \frac{r^*}{\epsilon p^*} [1 - \exp(p^*(x-1))].$$
 (127)

We now solve the problem (118), (119) to obtain

$$e_{j+1}(x, y) = \sum_{\omega=1}^{\infty} \hat{e}_{j+1}(x, \omega) \sqrt{2} \sin(\omega \pi y),$$
(128)

$$\hat{e}_{j+1}(x,\omega) = \int_0^x \exp\left(-\int_t^x q(s,\omega)\,\mathrm{d}s\right) \hat{w}(t,\omega)\,\mathrm{d}t,\tag{129}$$

$$q(s,\omega) = \frac{\varepsilon}{a(s)} [\omega\pi]^2.$$
(130)

We note that the series (120), (124) and (128) are uniformly convergent in view of the homogeneous conditions at y = 0 and y = 1 that w,  $e_j$  and  $e_{j+1}$  satisfy.

We define  $q^*$  and  $x_q$  as

$$q^* = \min_{x \in [0,1]} q(x,\omega) = \frac{\varepsilon}{a(x_q)} [\omega\pi]^2,$$
(131)

and we manipulate (129) to obtain

$$|\hat{e}_{j+1}(x,\omega)| \leq \int_0^x \exp(q^*(t-x))|\hat{w}(t,\omega)| \,\mathrm{d}t.$$
 (132)

Then we use (127) in (132) and manipulate the resulting expression to obtain

$$|\hat{e}_{j+1}(x,\omega)| \leq \frac{r^*}{\varepsilon p^* q^*} [1 - \exp(-q^* x)].$$
 (133)

We use the definitions of  $r^*$ ,  $p^*$ , and  $q^*$  in (125)–(131) to obtain

$$\frac{\|\hat{e}_{j+1}(x,\omega)\|_{\infty}}{\|\hat{e}_{j}(x,\omega)\|_{\infty}} \leq \frac{a(x_q)a(x_p)}{a(x_p)^2} \frac{|\varepsilon a_x(x_p) + (\varepsilon \omega \pi)^2|}{a(x_p)^2 + (\varepsilon \omega \pi)^2}.$$
(134)

We now define the convergence factor of mode  $\omega$ ,  $\rho_{\omega}$ , by

$$\rho_{\omega} = \frac{\|\hat{e}_{j+1}(x,\omega)\|_{\infty}}{\|\hat{e}_{j}(x,\omega)\|_{\infty}},\tag{135}$$

and we note from (134) that

$$\rho_{\omega} \leq A^2 \frac{|\varepsilon a_x(x_r) + (\varepsilon \omega \pi)^2|}{a(x_p)^2 + (\varepsilon \omega \pi)^2},\tag{136}$$

where

$$A = \frac{\max_{x \in [0,1]} a(x)}{\min_{x \in [0,1]} a(x)}.$$
(137)



Fig. 3. The bound for  $\rho_{\omega}$  as a function of  $\omega$  for  $\varepsilon = \frac{1}{25}$ ,  $\frac{1}{50}$  and  $\frac{1}{100}$ , from left to right.

Fig. 3 shows the bound (136) normalized by  $A^2$ , as a function of  $\omega$  for  $\varepsilon = \frac{1}{25}$ ,  $\frac{1}{50}$  and  $\frac{1}{100}$ . The bound is evaluated with  $a(x_p) = a_x(x_r) = 1$ . We see that modes  $\omega \ll \varepsilon^{-1}$  are rapidly damped and that  $\rho_{\omega}$  is uniformly bounded for all modes. It follows that the numerical implementation of QUAD is particularly effective when the mesh parameter required to accurately resolve the problem satisfies  $h > \varepsilon$ . Then all the modes of the grid are in the rapid convergence range. This occurs for the amplitude equations generated by the hybrid numerical asymptotic method described in Section 2.

In order to enhance the rate of decay of the high frequency modes of the grid, an additional smoothing step [2] can be used at each iteration. Moreover, we are currently seeking an approximate factorization scheme which generates operators that effectively damp modes in intervals  $[\omega_1, \omega_2]$ , with  $\omega_1 > 1/\varepsilon$ .

The formal factorization scheme of Section 6.1 does not always yield a useful approximate operator for QUAD. Indeed, if we substitute  $a(x_p) = ib$  with b real in (136), we find that the resulting expression has a pole at  $\omega = b/(\epsilon \pi)$ . Hence we expect that  $\rho_{\omega} \gg 1$  for modes  $\omega \approx b/(\epsilon \pi)$  and QUAD is highly unstable in this case. We have observed this instability in numerical calculations for Eq. (103) in which ib(x, y) was substituted for a(x, y), with b real.

#### 6.4. Numerical calculations

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We now apply QUAD to solve (103) in  $[0,1] \times [0,1]$ . The forcing function f is chosen such that the solution to the problem is

$$u(x, y) = (x - 1)^2 \sin(10\pi y) \cos(10\pi x).$$
(138)



Fig. 4. The magnitude of the residual (141) for the QUAD iterates. In each graph the ordinate is  $\log_{10} ||residual||_{\infty}$  for the value of  $\varepsilon$  shown at the top. The abscissa is the number of iterations.

We impose Dirichlet conditions at y = 0, y = 1 and x = 0 which are compatible with the solution (138). At x = 1, we impose the boundary condition

$$\left(\frac{\partial}{\partial x} - \frac{\varepsilon}{a}\frac{\partial^2}{\partial y^2}\right)u = 0.$$
(139)

The coefficient a(x, y) in (103) is

$$a(x, y) = 0.5 \times (3 - \cos(20\pi\sqrt{x^2 + y^2})). \tag{140}$$

We use the second-order factorization (106) and the factors are inverted with the Crank Nicholson methods. In our calculation the mesh parameter is  $h = \frac{1}{100}$ . The magnitude of the residual

$$f_h - L_h u_k, \tag{141}$$

is shown in Fig. 4 for several values of  $\varepsilon$ . The rate of convergence increases as  $\varepsilon$  decreases and the iteration is stable in all cases.

#### 6.5. Other applications

We now consider operators of the form

$$L = \varepsilon \left( \frac{\partial^2}{\partial x^2} + A \frac{\partial^2}{\partial x \partial y} + B \frac{\partial^2}{\partial y^2} + C \frac{\partial}{\partial x} + D \frac{\partial}{\partial y} \right) + E \frac{\partial}{\partial y} + F \frac{\partial}{\partial x} + G,$$
(142)

in  $\Omega \subset \mathbb{R}^2$  where the functions A-G may be complex valued, and  $[E, F] \neq 0$ . The operators (142) can be formally factored using the scheme of Section 6.1, once they are transformed into the form (87). This is possible by first representing (142) in a characteristic coordinate system determined by curves  $(X(\sigma, \tau), Y(\sigma, \tau))$  [23] satisfying

$$\frac{\partial Y}{\partial \sigma} = E(X, Y), \qquad \frac{\partial X}{\partial \sigma} = F(X, Y),$$
(143)

and then normalizing the resulting operator by the coefficient of  $\partial^2/\partial\sigma^2$ .

This transformation is possible only when the equations

$$x = X(\sigma, \tau), \qquad y = Y(\sigma, \tau),$$
 (144)

can be uniquely inverted for  $\sigma$  and  $\tau$  in  $\Omega$ . Moreover, the transformation of variables requires derivatives of  $\sigma(x, y)$  and  $\tau(x, y)$  which in some cases can only be approximated numerically.

## 7. Future direction

The treatment of general boundary conditions in more than one dimension is possible with a "shooting method" for second order systems in factored form, analogous to the method of Section 3.2. We shall present this method in a future paper. We are also interested in the use of the factored operators as preconditioners for Krylov accelerators and the combination of QUAD with a smoothing step. Approximate operators that would complement the operators of Section 6.1 and enable QUAD to effectively damp modes of the error in other frequency ranges are desirable.

## Appendix A. Existence and uniqueness of solution

We now study the existence and uniqueness of the solution to the boundary value problem

$$\varepsilon \left(\frac{\mathrm{d}}{\mathrm{d}x} + \frac{a_1(x)}{\varepsilon} + S_m(x,\varepsilon)\right) \left(\frac{\mathrm{d}}{\mathrm{d}x} - S_m(x,\varepsilon)\right) = f,\tag{A.1}$$

subject to either of the boundary conditions

$$u(b) = \beta, \qquad u(a) = 1,$$
 (A.2)

$$\left(\frac{\mathrm{d}}{\mathrm{d}x} - A\right)u(b) = \beta, \qquad u(a) = 1.$$
 (A.3)

Here we assume that

$$S_m(b,0) \neq A,\tag{A.4}$$

 $a_1 \neq 0$  and  $\operatorname{Re}[a_1] \leq 0$ . We will now prove

**Lemma A.1.** (1) The problem (A.1)–(A.3) has a unique solution for  $\varepsilon$  sufficiently small.

(2) The problem (A.1), (A.2) has a unique solution for  $\varepsilon$  sufficiently small provided Re[a<sub>1</sub>] is not identically 0 in [a,b].

(3) The problem (A.1), (A.2) has a unique solution for  $\varepsilon$  sufficiently small provided Re[a<sub>1</sub>] is identically 0 in [a,b] and

$$\exp\left(2\int_{a}^{b}S_{m}(t,0)\,\mathrm{d}t\right)\bigg|\neq |a_{1}(a)/a_{2}(b)|. \tag{A.5}$$

**Proof.** We first rewrite (A.1) as the first-order system

$$\begin{pmatrix} u \\ w \end{pmatrix}' = \begin{pmatrix} S_m & 1 \\ 0 & -\left(\frac{a_1}{\varepsilon} + S_m\right) \end{pmatrix} \begin{pmatrix} u \\ w \end{pmatrix} + \begin{pmatrix} 0 \\ f \end{pmatrix},$$
 (A.6)

subject to

$$B_{a} \begin{pmatrix} u(a) \\ w(a) \end{pmatrix} + B_{b} \begin{pmatrix} u(b) \\ w(b) \end{pmatrix} = \begin{pmatrix} 1 \\ \beta \end{pmatrix},$$
(A.7)

where

$$B_a = \begin{pmatrix} 1 & 0\\ 0 & 0 \end{pmatrix},\tag{A.8}$$

and  $B_b$  is given for conditions (A.2) and (A.3) respectively by

$$B_b = \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix}, \qquad B_b = \begin{pmatrix} 0 & 0 \\ S_m - A & 1 \end{pmatrix}, \tag{A.9}$$

respectively. The existence theory for linear boundary value problems (see [13, Theorem 3.26]) guarantees a unique solution to these equations provided the matrix

$$Q(\varepsilon) = B_a Y(a, \varepsilon) + B_b Y(b, \varepsilon), \tag{A.10}$$

is nonsingular. Here Y is a fundamental solution matrix for the system (A.6) and we choose

$$Y(x,\varepsilon) = \begin{pmatrix} \exp(\int_{b}^{x} S_{m} dt) \int_{a}^{x} \exp(\int_{z}^{b} [2S_{m}(t,\varepsilon) + \frac{a_{1}(t)}{\varepsilon}] dt) dz & \exp(\int_{a}^{x} S_{m} dt) \\ \exp(\int_{x}^{b} [S_{m} + \frac{a_{1}(t)}{\varepsilon}] dt) & 0 \end{pmatrix}.$$
 (A.11)

The determinant of  $Q(\varepsilon)$  is nonzero provided that

$$H(b,\varepsilon) \neq \mu, \tag{A.12}$$

where

$$H(x,\varepsilon) = \int_{a}^{x} \exp\left(\int_{z}^{x} S_{m}(t,\varepsilon) dt + \int_{z}^{b} S_{m}(t,\varepsilon) + \int_{z}^{b} \frac{a_{1}(t)}{\varepsilon} dt\right) dz.$$
(A.13)

and the constant  $\mu$  is given by

$$\mu = 0, \qquad \mu = \frac{1}{A - S_m(b,\varepsilon)},\tag{A.14}$$

for conditions (A.2) and (A.3), respectively. In order to verify when condition (A.12) holds, we first determine the asymptotic expansion of H as  $\varepsilon \rightarrow 0$ .

Remark A.2. H defined in (A.13) satisfies

$$H(x,\varepsilon) \sim \varepsilon \left( \frac{f(a,x)}{a_1(a)} e^{\phi(a)/\varepsilon} - \frac{f(x,x)}{a_1(x)} e^{\phi(x)/\varepsilon} \right), \quad \varepsilon \to 0.$$
(A.15)

Here we have introduced the functions

$$f(z,x) = \exp\left(\int_{z}^{x} S_{m}(t,\varepsilon) dt + \int_{z}^{b} S_{m}(t,\varepsilon) dt\right), \quad \phi(z) = \int_{z}^{b} a_{1}(t) dt.$$
(A.16)

To prove (A.15) we rewrite H as the Laplace integral

$$H(x,\varepsilon) = \int_{a}^{x} f(z,x) e^{\phi(z)/\varepsilon} dz.$$
 (A.17)

In view of the relation  $\phi'(z) = -a_1(z) \neq 0$ , and the differentiability of f and  $a_1$ , we integrate (A.17) by parts to obtain

$$H(x,\varepsilon) = \varepsilon \left( \frac{f(a,x)}{a_1(a)} e^{\phi(a)/\varepsilon} - \frac{f(x,x)}{a_1(x)} e^{\phi(x)/\varepsilon} \right) + \varepsilon \int_a^x \frac{\mathrm{d}}{\mathrm{d}z} \left[ \frac{f(z,x)}{a_1(z)} \right] e^{\phi(z)/\varepsilon} \,\mathrm{d}z. \tag{A.18}$$

The integral in (A.18) is o(1) as  $\varepsilon \rightarrow 0$ . Indeed, let

$$x_0 = \max \{ x \mid x \in [a, b], \ \operatorname{Re}[a_1] < 0 \}.$$
(A.19)

Then for all  $x \in [a, x_0)$ ,  $\text{Re}[\phi(z)] < 0$ , and the integrand

$$\frac{\mathrm{d}}{\mathrm{d}z}\left[\frac{f(z,x)}{a_1(z)}\right]\mathrm{e}^{\phi(z)/\varepsilon},$$

tends to zero as  $\varepsilon \to 0$ . Hence, the integral in (A.18) is o(1) as  $\varepsilon \to 0$ , for all  $x \in [a, x_0]$ . If  $x_0 < b$  then for all  $x \in [x_0, b]$ ,  $\operatorname{Re}[a_1] = 0$  and  $\operatorname{Im}[a_1] \neq 0$ . Hence  $\phi(z)$  is purely imaginary, continuously differentiable and  $-i\phi$  is monotone, because  $\phi'(z) = -a_1(z) \neq 0$ . The change of variables  $u = \phi(z)/i$  yields a Fourier integral which is o(1) as  $\varepsilon \to 0$ , in view of the Riemann Lebesgue Lemma. Thus (A.15) is proved.

We now complete the proof of the lemma by verifying when condition (A.12) holds in each of the two cases given in (A.14).

Case I:  $\mu \neq 0$ 

- In view of the expansion (A.15),  $H \rightarrow 0$  as  $\varepsilon \rightarrow 0$  and therefore condition (A.12) holds for all sufficiently small  $\varepsilon$ .
  - Case II:  $\mu = 0$
- If both  $a_1$  and  $S_m$  are real then H in (A.13) is positive for b > a and (A.12) holds for all  $\varepsilon > 0$ .

- If  $\operatorname{Re}[a_1]$  is not identically zero in [a, b], then  $\phi(a)$  has a negative real part and  $H \sim -\varepsilon/a_1(b)$  as  $\varepsilon \to 0$ , in view of (A.18). Hence, (A.12) holds for all sufficiently small  $\varepsilon$ .
- If  $a_1$  is purely imaginary so is  $\phi(a)$  and (A.12) holds for all sufficiently small  $\varepsilon$  provided (A.5) holds, in view of (A.18).

It follows from the above discussion that

**Remark A.3.** When one of the conditions of Lemma A.1 is satisfied, the function  $H/\varepsilon$  is uniformly bounded away from 0 for all sufficiently small  $\varepsilon$ .

#### Appendix B. Case 2 of the asymptotic factorization

We now solve (95)-(97) when  $s_0 = -a_1$ . Specifically, we solve (95) in detail and describe how to extend the solution procedure to (96) and (97).

In order to solve (95) we substitute  $-a_1$  for  $s_0$  into that equation to obtain

$$s_1 a_1 = a_0 - a_{1,\sigma} - b_1 a_{1,\beta} - b_1 a_1 \frac{\partial}{\partial \beta} - c_1 a_1.$$
(B.1)

The right-hand side of (B.1) is a first-order differential operator and we conclude that  $s_1$  has the form

$$s_1 = \alpha_0 + \alpha_1 \frac{\partial}{\partial \beta}.$$
 (B.2)

The coefficients  $\alpha_0$  and  $\alpha_1$  in (B.2) are to be determined so that  $s_1$  satisfies (B.1). Upon substituting (B.2) into (B.1), we obtain

$$a_1\alpha_0 + a_1\alpha_1\frac{\partial}{\partial\beta} + \alpha_1a_{1,\beta} = a_0 - a_{1,\sigma} - b_1a_{1,\beta} - b_1a_1\frac{\partial}{\partial\beta} - c_1a_1.$$
(B.3)

We equate the coefficient of  $\partial/\partial\beta$  on each side of (B.3) to obtain

$$\alpha_1 = -b_1. \tag{B.4}$$

Then, we introduce the value of  $\alpha_1$  back into (B.3) and find that

$$\alpha_0 = \frac{a_0 - a_{1,\sigma} - c_1 a_1}{a_1}.$$
(B.5)

Combining (B.2), (B.4) and (B.5) we obtain

$$s_1 = \frac{a_0 - a_{1,\sigma} - c_1 a_1}{a_1} - b_1 \frac{\partial}{\partial \beta}.$$
 (B.6)

The procedure we used to solve (95) can be generalized to solve (96) and (97). We now describe this extension to (97). In order to do so, we first show by induction that the right-hand side of that equation is a linear differential operator of order j. We denote this operator by

$$\beta_0 + \beta_1 \frac{\partial}{\partial \beta} + \dots + \beta_j \frac{\partial^j}{\partial \beta^j}.$$
 (B.7)

The coefficients  $\beta_j$  in (B.7) depend on the coefficients of the operators  $s_0, \ldots s_{j-1}$ . The left-hand side of (97) is  $-s_j a_1$ . Hence in order for  $s_j$  to satisfy (97), it has to be a linear differential operator of order j

$$s_j = \alpha_0 + \alpha_1 \frac{\partial}{\partial \beta} + \dots + \alpha_j \frac{\partial^j}{\partial \beta^j}.$$
 (B.8)

The unknown coefficients  $\alpha_l$  in (B.8) are to be determined so that  $s_j$  satisfies (97). Upon introducing the right-hand side of (B.8) into the left-hand side of (97) and then replacing the right-hand side of (97) with (B.7) we find that

$$a_1\left(\alpha_1\frac{\partial}{\partial\beta}+\cdots+\alpha_j\frac{\partial^j}{\partial\beta^j}\right)+\alpha_0a_1+\alpha_1a_1^{(1)}+\cdots\alpha_ja_1^{(j)}=\beta_0+\beta_1\frac{\partial}{\partial\beta}+\cdots+\beta_j\frac{\partial^j}{\partial\beta^j},$$
(B.9)

where  $a_1^{(j)}$  denotes  $\partial^j/\partial\beta^j a_1$ . Upon equating coefficients of derivatives of the same order we find that

$$\alpha_j = \beta_j / a_1 \quad \text{for } j \ge 1, \tag{B.10}$$

and

$$\alpha_0 = \frac{1}{a_1} \left( \beta_0 - \frac{\beta_1 a_1^{(1)}}{a_1} - \dots - \frac{\beta_1 a_1^{(j)}}{a_1} \right).$$
(B.11)

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