Bounded Error Schemes for the Wave Equation on Complex Domains

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BOUNDED ERROR SCHEMES FOR THE WAVE EQUATION ON COMPLEX DOMAINS[*]

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Abstract. This paper considers the application of the method of boundary penalty terms ("SAT") to the numerical solution of the wave equation on complex shapes with Dirichlet boundary conditions. A theory is developed, in a semi-discrete setting, that allows the use of a Cartesian grid on complex geometries, yet maintains the order of accuracy with only a linear temporal error-bound. A numerical example, involving the solution of Maxwell's equations inside a 2-D circular wave-guide demonstrates the efficacy of this method in comparison to others (e.g. the staggered Yee scheme) - we achieve a decrease of *two orders of magnitude* in the level of the L_2-error.

Key words. Maxwell's equations, wave equation, finite difference–time domain, error bounds, boundary conditions, complex geometries, staircasing

Subject classification. Applied and Numerical Mathematics

1. Introduction. Hyperbolic systems of P.D.E.'s describing physical situations such as electro-magnetism, acoustics, elastic waves, etc, may under many circumstances be cast as wave equations for the various field components.

One class of problems is that of solving numerically the Dirichlet problem on complex shapes, e.g., inside wave guides. For sufficiently non-simple geometries, the option of transforming the problem to body-fitted coordinates is not always a viable option, especially in three space dimensions. There are other options, such as using Cartesian grids and approximating the body shape via "staircasing", "diagonal split cell model", etc (see for example Chapter 10 in reference [4]). It is well known that these devices are not very efficacious, particularly in the high frequency regime. We shall demonstrate that "staircasing" can fail even for low frequencies.

In this paper we consider the application of the method of boundary penalty terms ("SAT", see references [1], [2], [3]) to the numerical solution of the wave equation in a finite domain with Dirichlet boundary conditions.

In Section 2 we develop the theory that allows us to use a Cartesian grid on complex geometries and yet maintain the order accuracy with a linear temporal error-bound.

In Section 3 we construct a second order accurate scheme that fulfills the conditions imposed by the theory presented in Section 2.

Section 4 is devoted to a numerical example – the solution of the transverse magnetic (TM) Maxwell's equations [4] between two concentric circles. (This configuration might be considered as a cross-section of a very long wave-guide.) This problem is solved using four different numerical algorithms. Two of them solve the first order system with "staircasing" – the Yee staggered scheme [6] and a 4th order spatially staggered scheme due to Turkel and Yefet [5]. The other two solve the wave equation directly on a non-staggered

Cartesian grid, one with the SAT formulation and one without. All three "standard" (non-SAT) algorithms have very large errors; the SAT algorithm has errors that are at least two order of magnitude smaller. Summary and conclusions, and ideas for future work are presented in Section 5.

2. Theoretical Framework of the Method. In reference [1], [2] and [3], it was shown how the case of a one-dimensional P.D.E. can be used as a building block for the multidimensional case for constructing error-bounded algorithms over complex geometries with Dirichlet boundary condition. We therefore start with the following one-dimensional problem:

$$(2.1) \qquad \frac{\partial^2 u}{\partial t^2} = \frac{\partial^2 u}{\partial x^2} + f(x,t); \quad \Gamma_L \le x \le \Gamma_R, \quad t > 0$$

$$(2.0a) \qquad u(x,0) = u_0(x)$$

$$(2.0b) \qquad \frac{\partial}{\partial t}u(x,0) = u_{t_0}(x)$$

$$(2.0c) \qquad u(\Gamma_L,t) = g_L(t)$$

$$(2.0d) \qquad u(\Gamma_R,t) = g_R(t)$$

and $f(x,t) \in C^2$.

Let us discretize (2.1) spatially on the uniform grid presented in Figure 2.1. Note that the boundary points do not necessarily coincide with x_1 and x_N. Set $x_{j+1} - x_j = h$, $1 \le j \le N-1$; $x_1 - \Gamma_L = \gamma_L h$, $0 \le \gamma_L < 1$; $\Gamma_R - x_N = \gamma_R h$, $0 \le \gamma_R < 1$.
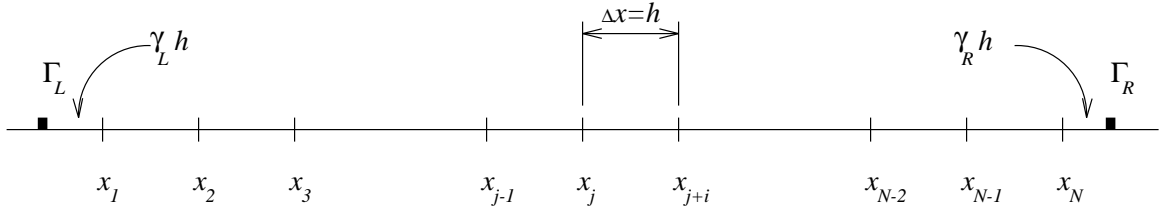

FIG. 2.1. *One dimensional grid.*

Since, unlike the cases discussed in [1], [2], equation (2.1) has a second time derivative, attempts to apply naively the methods presented there fail. The reason is that if we follow the procedure used there and write the following discrete approximation to (2.1),

$$(2.2) \qquad \frac{d^2}{dt^2}\mathbf{u} = D\mathbf{u} + \mathbf{f}(t) + \mathbf{T}_e$$

where $\mathbf{u}$ is the projection of the exact solution $u(x,t)$ onto the grid, i.e. $u(x_j,t) = u_j(t) \stackrel{\triangle}{=} \mathbf{u}(t)$; and write the numerical scheme

$$(2.3) \qquad \frac{d^2\mathbf{v}}{dt^2} = \left[D\mathbf{v} - \tau_L(A_L\mathbf{v} - \mathbf{g}_L) - \tau_R(A_R\mathbf{v} - \mathbf{g}_R)\right] + \mathbf{f}(t) \ ,$$

then the equation for the error vector $\boldsymbol{\epsilon} = \mathbf{u} - \mathbf{v}$ becomes

$$(2.4) \qquad \frac{d^2\,\boldsymbol{\epsilon}}{dt^2} = M\,\boldsymbol{\epsilon} + \mathbf{T}\ .$$

In the above, $\mathbf{v}$ is the numerical approximation to $\mathbf{u}$, and

$$(2.5) \qquad M = D - \tau_L A_L - \tau_R A_R\ .$$

D is a differentiation matrix of the proper order of accuracy that does not use boundary values. The matrices A_L and A_R are defined by the relations

$$(2.6) \qquad A_L \mathbf{u} = \mathbf{g}_L - \mathbf{T}_L, \quad A_R \mathbf{u} = \mathbf{g}_R - \mathbf{T}_R\ ,$$

i.e., each row in $A_L(A_R)$ is composed of the coefficients extrapolating $\mathbf{u}$ to its boundary value $g_L(g_R)$ at $\Gamma_L(\Gamma_R)$ to within the order of accuracy. (The error is then $T_L(T_R)$.) The diagonal matrices τ_L and τ_R are given by

$$\tau_L = \mathrm{diag}(\tau_{L_1}, \tau_{L_2}, \cdots, \tau_{L_N}); \quad \tau_R = \mathrm{diag}(\tau_{R_1}, \tau_{R_2}, \cdots, \tau_{R_N})\ .$$

The constrain on the construction of the A's, τ's and D is that M in (2.4) be negative definite. The negative definiteness of M is a necessary condition for extending the 1-D theory to the multidimensional case (see [1],[3]). Also in (2.4)

$$(2.7) \qquad \mathbf{T} = \mathbf{T}_e - \tau_L \mathbf{T}_L - \tau_R \mathbf{T}_R = (T_1, T_2, \cdots, T_m, \cdots, T_N)^T\ .$$

If the matrix M can be diagonalized*, then

$$(2.8) \qquad M = Q^{-1}\Lambda Q$$

with the diagonal matrix, Λ, having the eigenvalues of M. Defining $\boldsymbol{\mu} = Q\,\boldsymbol{\epsilon}$, equation (2.4) becomes

$$\frac{d^2\,\boldsymbol{\mu}}{dt^2} = \Lambda\,\boldsymbol{\mu} + Q\mathbf{T}$$
$$(2.9) \qquad\qquad = \Lambda\,\boldsymbol{\mu} + \hat{\mathbf{T}}\ .$$

This is an un-coupled system of O.D.E's. The general solution for the mth equation is:

$$\mu_m(t) = c_{m_1} e^{\sqrt{\lambda_m}t} + c_{m_2} e^{-\sqrt{\lambda_m}t} + \frac{1}{\sqrt{\lambda_m}} \int_0^t \sinh\left(\sqrt{\lambda_m}(t-s)\right)\hat{T}_m(s)ds\ .$$

Recalling that at $t = 0$, $\boldsymbol{\epsilon} = \boldsymbol{\epsilon}_t = 0$ (i.e. $\boldsymbol{\mu} = \boldsymbol{\mu}_t = 0$ at $t = 0$), the solution of (2.9) becomes:

$$(2.10) \qquad \mu_m(t) = \frac{1}{\sqrt{\lambda_m}} \int_0^t \hat{T}_m(s)\sinh\left[\sqrt{\lambda_m}(t-s)\right]ds\ .$$

Note that unless all the eigenvalues of M are real and non-positive some of the $\sqrt{\lambda_m}$'s will have a positive real part, in which that case at least one of the μ_m's may grow exponentially in time. In order to prevent this, we have to demand that M, in addition to being negative definite, also possess only real eigenvalues.

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*Extensive numerical evidence has shown that the M in [1],[2] (i.e. representing the second derivative to 4th and 2nd order accuracy, respectively) has distinct eigenvalues and hence is diagonalizable.

Furthermore, in order to use the 1-D scheme as a building block for multidimensional schemes, M should be built in a way that verifies that the property of real negative eigenvalues carries over to the multi-dimensional differentiating matrix. One way to achieve this goal is to construct M as a negative-definite *symmetric* matrix. Then an estimate on the error bound can be derived directly from the solution (2.10),

$$|\mu_m(t)| \leq \frac{1}{\sqrt{|\lambda_m|}} \hat{T}_{m_M} t$$

where $\hat{T}_{m_M} = \max_{0 \leq s \leq t} |\hat{T}_m(s)|$. Then, for a normalized Q,

(2.11) $$\| \boldsymbol{\epsilon} \| = \| \boldsymbol{\mu} \| \leq \frac{1}{c_0} \| \hat{\mathbf{T}}_M \| t ,$$

where $c_0 = \min_{m=1,\cdots,N} \sqrt{|\lambda_m|}$. Therefore $\| \boldsymbol{\epsilon} \|$ grows at most linearly with t.

This result, of a linear temporal bound on the error-norm, can also be derived by resorting to energy method (see [3]), instead of directly from the solution.

Also, as mentioned before, the construction of multi-dimensional case

$$\frac{\partial^2 u}{\partial t^2} = \nabla^2 u + f(\mathbf{x}, t)$$

on complex shapes is completely analogous to the method indicated in [1], [3].

3. Construction of the Scheme. This section is devoted to the task of constructing a symmetric negative definite matrix M for the case of a second order accurate finite difference algorithm.

Let

$$D = \frac{1}{h^2} \left\{ \begin{bmatrix} 1 & -2 & 1 & 0 \\ 1 & -2 & 1 & 0 \\ 0 & 1 & -2 & 1 \\ 0 & 0 & 1 & -2 & 1 \\ & & & \ddots & \ddots & \ddots \\ & & & & 1 & -2 & 1 & 0 & 0 \\ & & & & & 1 & -2 & 1 & 0 \\ & & & & & & 1 & -2 & 1 \\ & & & & & & 1 & -2 & 1 \end{bmatrix} \right.$$

$$+ \begin{bmatrix} 0 \\ & c_2 \\ & & c_3 \\ & & & \ddots \\ & & & & c_{N-2} \\ & & & & & c_{N-1} \\ & & & & & & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 & 0 \\ 1 & -3 & 3 & -1 \\ -1 & 4 & -6 & 4 & -1 \\ & \ddots & \ddots & \ddots & \ddots & \ddots \\ & & -1 & 4 & -6 & 4 & -1 \\ & & & -1 & 3 & -3 & 1 \\ & & & 0 & 0 & 0 & 0 \end{bmatrix}$$

4

(3.1)

$$-\tilde{c} \left. \begin{bmatrix} 0 & 0 & 0 & & & & & \\ 0 & 1 & -2 & 1 & & & & \\ -1 & 2 & 0 & -2 & 1 & & & \\ & \ddots & \ddots & \ddots & \ddots & \ddots & & \\ & & -1 & 2 & 0 & -2 & 1 \\ & & & -1 & 2 & -1 & 0 \\ & & & 0 & 0 & 0 & 0 \end{bmatrix} \right\}$$

where

(3.2)
$$c_k = c_2 + \frac{c_{N-1} - c_2}{N-3}(k-2)\,,$$

and

(3.3)
$$\tilde{c} = \frac{c_{N-1} - c_2}{N-3}\,.$$

Note, that as in [2] and [3], we had to resort to using *connectivity terms*, the last two matrices in (3.1).

(3.4)
$$A_L = \begin{bmatrix} \frac{1}{2}(2+\gamma_L)(1+\gamma_L) & -\gamma_L(2+\gamma_L) & \frac{1}{2}(\gamma_L + \gamma_L^2) & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & & \vdots \\ \frac{1}{2}(2+\gamma_L)(1+\gamma_L) & -\gamma_L(2+\gamma_L) & \frac{1}{2}(\gamma_L + \gamma_L^2) & 0 & \dots & 0 \end{bmatrix};$$

(3.5)
$$A_R = \begin{bmatrix} 0 & \dots & 0 & \frac{1}{2}(\gamma_R + \gamma_R^2) & -\gamma_R(2+\gamma_R) & \frac{1}{2}(2+\gamma_R)(1+\gamma_R) \\ \vdots & \vdots & \vdots & \vdots & & \vdots \\ 0 & \dots & 0 & \frac{1}{2}(\gamma_R + \gamma_R^2) & -\gamma_R(2+\gamma_R) & \frac{1}{2}(2+\gamma_R)(1+\gamma_R) \end{bmatrix}.$$

(3.6)
$$\tau_L = \frac{1}{h^2}\mathrm{diag}\left[\tau_{L_1}, \tau_{L_2}, \tau_{L_3}, 0, \dots, 0, 0\right];$$

(3.7)
$$\tau_R = \frac{1}{h^2}\mathrm{diag}\left[0, 0, \dots, 0, \tau_{R_{N-2}}, \tau_{R_{N-1}}, \tau_{R_N}\right];$$

In order to make the matrix $M = D - \tau_L A_L - \tau_R A_R$ symmetric we choose:

$$c_2 = \frac{(1-\gamma_L)\,\gamma_L}{2}$$

$$c_{N-1} = \frac{(1-\gamma_R)\,\gamma_R}{2}$$

$$\tau_{L_2} = \frac{3 - \gamma_L - 2\,\gamma_L\,\tau_{L_1}}{1 + \gamma_L}$$

(3.8)

$$\tau_{L_3} = \frac{-2 + \gamma_L + \gamma_L\,\tau_{L_1}}{2 + \gamma_L}$$

$$\tau_{R_{N-1}} = \frac{3 - \gamma_R - 2\,\gamma_R\,\tau_{R_N}}{1 + \gamma_R}$$

$$\tau_{R_{N-2}} = \frac{-2 + \gamma_R + \gamma_R\,\tau_{R_N}}{2 + \gamma_R}$$

$$\tau_{L_1}, \tau_{R_N} \geq 4\ .$$

(3.9)

The proof that the symmetric matrix M is indeed negative-definite is given in the Appendix to this paper.

Note also that instead of solving (2.3) directly as a 2$^{\text{nd}}$ order O.D.E. system in time, one can solve

$$\frac{d\mathbf{w}}{dt} = [D\mathbf{v} - \tau_L(A_L\mathbf{v} - \mathbf{g}_L) - \tau_R(A_R\mathbf{v} - \mathbf{g}_R)] + \mathbf{f}$$

$$\frac{d\mathbf{v}}{dt} = \mathbf{w}\ .$$

(3.10)

The number of 'variables' has increased from N to $2N$ but one gains in the simplicity of the time integration.

4. Numerical Example. We consider the dimensionless Maxwell's equation for transverse magnetic field (TM, see [4]) in two space dimensions:

(4.1)
$$\frac{\partial E}{\partial t} = \frac{\partial H_y}{\partial x} - \frac{\partial H_x}{\partial y}$$

(4.2)
$$\frac{\partial H_x}{\partial t} = -\frac{\partial E}{\partial y}$$

(4.3)
$$\frac{\partial H_y}{\partial t} = \frac{\partial E}{\partial x}$$

where H_x and H_y are the x and y components of the magnetic vector, $\mathbf{H}$, and E is the electric field in the z-direction. The set (4.1)–(4.3) is to be solved in the space between two concentric circles, $\frac{1}{6} < r < \frac{1}{2}$. We consider the case of perfectly conducting boundaries. Thus the boundary conditions are given by

(4.4)
$$E(\tfrac{1}{2}, \theta, t) = 0$$

(4.5)
$$E(\tfrac{1}{6}, \theta, t) = 0\ .$$

We choose the following initial conditions (note the polar coordinates r, θ):

(4.6)
$$E(r, \theta, 0) = \cos\theta\,[J_1(\omega r) + a\,Y_1(\omega r)]$$

$$H_y(r, \theta, 0) = -\sin 2\theta\Big\{\frac{1}{2\omega r}\,[J_1(\omega r) + a\,Y_1(\omega r)]$$

(4.7)
$$-\tfrac{1}{4}\,[J_0(\omega r) - J_2(\omega r) + a\,Y_0(\omega r) - a\,Y_2(\omega r)]\,\Big\}$$

$$H_x(r, \theta, 0) = \frac{\cos^2\theta}{\omega r}\,[J_1(\omega r) + a\,Y_1(\omega r)]$$

(4.8)
$$-\frac{\sin^2\theta}{2}\,[J_0(\omega r) - J_2(\omega r) + a\,Y_0(\omega r) - a\,Y_2(\omega r)]$$

where the J_n's and the Y_n's are Bessel functions of the first and second kind of order n, respectively. Also,

(4.9)
$$a \cong 1.76368380110927; \qquad \omega \cong 9.813695999428405\ .$$

6

The exact solution of the IBV problem (4.1)–(4.8) is given by:

(4.10)
$$E(r, \theta, t) = \cos(\omega t + \theta) \left[J_1(\omega r) + a\, Y_1(\omega r) \right]$$

$$H_y(r, \theta, t) = -\frac{1}{\omega r} \cos\theta \cos(\omega t + \theta) \left[J_1(\omega r) + a\, Y_1(\omega r) \right]$$

(4.11)
$$+ \tfrac{1}{2} \cos\theta \sin(\omega t + \theta) \left[J_0(\omega r) - J_2(\omega r) + a\, Y_0(\omega r) - a\, Y_2(\omega r) \right]$$

$$H_x(r, \theta, t) = \frac{1}{\omega r} \cos\theta \cos(\omega t + \theta) \left[J_1(\omega r) + a\, Y_1(\omega r) \right]$$

(4.12)
$$- \tfrac{1}{2} \sin\theta \sin(\omega t + \theta) \left[J_0(\omega r) - J_2(\omega r) + a\, Y_0(\omega r) - a\, Y_2(\omega r) \right]$$

We note that we can extract from (4.1)–(4.3) a wave equation for the electric field E,

(4.13)
$$\frac{\partial^2 E}{\partial t^2} = \frac{\partial^2 E}{\partial x^2} + \frac{\partial^2 E}{\partial y^2} \ .$$

The boundary conditions on E in (4.13) are given by (4.4)–(4.5). The initial condition $E(r, \theta, 0)$ is given by (4.6). We need an additional initial condition on E_t, which we obtain by differentiating (4.10), namely

(4.14)
$$E_t(r, \theta, 0) = -\omega \sin\theta \left[J_1(\omega r) + a\, Y_1(\omega r) \right] \ .$$

Four numerical schemes were used to solve the problem:

(i) The Yee scheme [6]. This second order accurate scheme is staggered both in space and time. This entails putting initial conditions of H_x and H_y at $\Delta t/2$ rather than at $t = 0$. These initial conditions are derived from the exact solution. The numerical solution is carried out on the "staircased" domain shown in Figure 4.1.

(ii) A modification of the Yee scheme (designated Ty(2,4)), see [5]. This one has 4th order spatial accuracy and 2nd order in time. The stagger and the "staircased" domain are maintained as before.

(iii) The SAT algorithm for the wave equation described in Sections 2 and 3. The grid used for the numerical integration is shown in the right side of Figure 4.1. The time evolution is done by a 4th order Runge-Kutta method.

(iv) An algorithm which formally looks like the SAT in (iii), but is applied to the "staircased" domain of Figure 4.1 (rather than SAT one). To order $O(h^2)$, this is equivalent to using a standard spatial central differencing scheme with the nodal points at edges of the domain given the boundary value zero. The time integration is done as in the case (iii).

We first present the L_2 error in E for all four schemes at $t = 1$ and $t = 10$ for the cases $\Delta x = \Delta y = h = 1/40$, $h = 1/80$ and $h = 1/160$, see Table 1. Δt was $2/3\, h$ for the Yee scheme, $h/18$ for the Ty(2,4) scheme and $h/5$ for the SAT schemes.

It is immediately apparent from the table that the SAT-error (scheme iii) *is at least 2 orders of magnitude smaller* than that of the other three algorithms at all the various times and grid spacings.

Since the non-SAT schemes have errors which are unacceptably large we do not show details of their temporal behavior. The SAT algorithm (scheme iii) has an L_2 error which grows in time as shown in Figure 4.2. We see that this temporal growth is bound by a linear curve, whose slope depends on h. We note that for all reasonable dimensionless time the error is quite small, especially for $h \leq 1/80$.

5. Conclusions and Discussion.

(i) It seems quite clear from the evidence that the failure of the non-SAT schemes is due to the fact that "staircasing" misrepresents the shape of the body. In the SAT scheme, on the other hand, the penalty terms take account of the true shape.
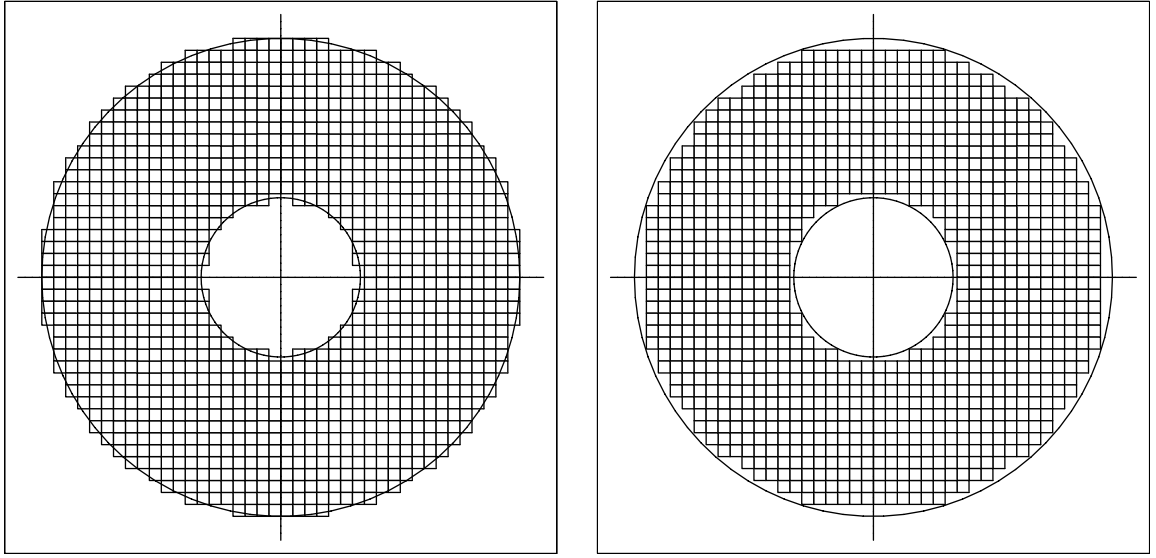
FIG. 4.1. *The "staircased" domain (left) and the SAT grid (right), $h = 1/40$.*

		$h = 1/40$	$h = 1/80$	$h = 1/160$
	$t = 1$			
i	Yee	0.4322	0.3635	0.1742
ii	Ty(2,4)	0.4038	0.3347	0.1579
iii	SAT	0.001203	0.0001705	1.5019e-05
iv	Staircased	0.1022	0.05041	0.01936
		$h = 1/40$	$h = 1/80$	$h = 1/160$
	$t = 10$			
i	Yee	0.5101	0.4364	0.6683
ii	Ty(2,4)	0.2642	0.7079	0.7243
iii	SAT	0.008435	0.0008354	8.2707e-05
iv	Staircased	0.7929	0.4735	0.7829

TABLE 4.1

The L_2 error.

(ii) The numerical results validate the theoretical predictions of the temporal behavior of the L_2 norm of the error.

(iii) Grosso-modo the CPU time per node is of the same order for all schemes.

(iv) The results from Table 1 and Figure 4.2 seem to indicate that the scheme (iii) converges as h^3, although the algorithm has a truncation error of order h^2. We do not understand this pleasant anomaly, although it is possible that even with $h = 1/160$ we are not yet in the asymptotic convergence regime.

(v) In the future, we would like to apply the SAT methodology directly to hyperbolic systems such as (4.1)–(4.3). The theory is not complete yet.
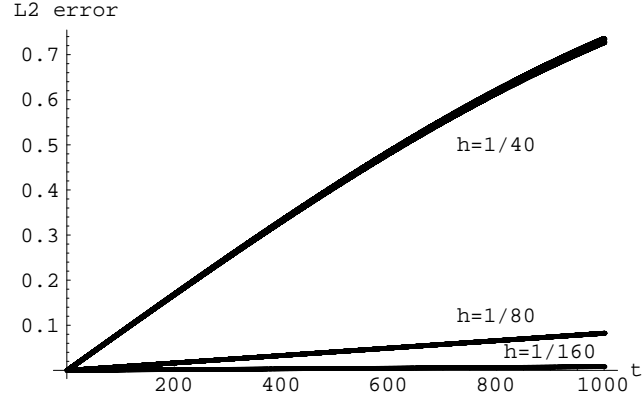
L2 error

0.7
0.6
0.5   h=1/40
0.4
0.3
0.2
0.1   h=1/80
       h=1/160
                                    t
  200   400   600   800   1000

FIG. 4.2. *SAT, L_2 error vs. time.*

Appendix. We decompose the matrix M, defined in (2.5) and (3.1) to (3.8) as follows:

$$(5.1) \qquad M = \frac{1}{h^2}\left[\alpha M_1 + (1-\alpha)M_2 + M_3 + M_4 + M_5\right]$$

where:

$$(5.2) \qquad M_1 = \begin{bmatrix}
-2 & 1 & & & & & \\
1 & -2 & 1 & & & & \\
& 1 & -2 & 1 & & & \\
& & \ddots & \ddots & \ddots & & \\
& & & 1 & -2 & 1 & \\
& & & & 1 & -2 & 1 \\
& & & & & 1 & -2
\end{bmatrix},$$

$$(5.3) \qquad M_2 = \begin{bmatrix}
0 & 0 & 0 & & & & \\
0 & 0 & 0 & & & & \\
0 & 0 & -1 & 1 & & & \\
& & 1 & -2 & 1 & & \\
& & & \ddots & \ddots & \ddots & \\
& & & & 1 & -2 & 1 \\
& & & & 0 & 1 & -1 & 0 & 0 \\
& & & & & & 0 & 0 & 0 \\
& & & & & & 0 & 0 & 0
\end{bmatrix},$$

9

$$M_3 = - \left\{ \begin{bmatrix} 0 & 0 & 0 \\ 0 & -1 & 1 \\ 0 & 1 & -2 & 1 \\ & & \ddots & \ddots & \ddots \\ & & & 1 & -2 & 1 & 0 \\ & & & & 1 & -1 & 0 \\ & & & & 0 & 0 & 0 \end{bmatrix} \right.$$

(5.4)
$$\left. \begin{bmatrix} 0 \\ & c_2 \\ & & c_3 \\ & & & \ddots \\ & & & & c_{N-2} \\ & & & & & c_{N-1} \\ & & & & & & 0 \end{bmatrix} \begin{bmatrix} 0 & 0 & 0 \\ 0 & -1 & 1 \\ 0 & 1 & -2 & 1 \\ & & \ddots & \ddots & \ddots \\ & & & 1 & -2 & 1 & 0 \\ & & & & 1 & -1 & 0 \\ & & & & 0 & 0 & 0 \end{bmatrix} \right\},$$

(5.5)
$$M_4 = \begin{bmatrix} m_4^{1,1} & m_4^{1,2} & m_4^{1,3} \\ m_4^{1,2} & m_4^{2,2} & m_4^{2,3} & 0 \\ m_4^{1,3} & m_4^{2,3} & m_4^{3,3} \\ \\ & 0 & & & 0 \end{bmatrix}$$

where:

$$m_4^{1,1} = 1 + 2\alpha - \frac{(1 + \gamma_L)(2 + \gamma_L)\tau_{L_1}}{2}$$

$$m_4^{1,2} = -2 - \alpha + \gamma_L (2 + \gamma_L) \tau_{L_1}$$

$$m_4^{1,3} = 1 - \frac{\gamma_L (1 + \gamma_L) \tau_{L_1}}{2}$$

$$m_4^{2,2} = 2\alpha + \frac{7\gamma_L - 4\left(1 + \gamma_L^2\right) - \gamma_L{}^2 (2 + \gamma_L)(1 + 4\tau_{L_1})}{2(1 + \gamma_L)}$$

$$m_4^{2,3} = 1 - \alpha - \frac{3\gamma_L}{2} + \frac{\gamma_L{}^2}{2} + \gamma_L{}^2 \tau_{L_1}$$

$$m_4^{3,3} = 2\alpha + \frac{-4 + \gamma_L{}^2 - \gamma_L{}^3 - \gamma_L{}^2 (1 + \gamma_L)\tau_{L_1}}{2(2 + \gamma_L)}$$

and

$$(5.6) \qquad M_5 = \begin{bmatrix} 0 & & 0 & \\ & m_5^{N-2,N-2} & m_5^{N-1,N-2} & m_5^{N,N-2} \\ 0 & m_5^{N-1,N-2} & m_5^{N-1,N-1} & m_5^{N,N-1} \\ & m_5^{N,N-2} & m_5^{N,N-1} & m_5^{N,N} \end{bmatrix}$$

where:

$$m_5^{N,N} = 1 + 2\alpha - \frac{(1+\gamma_R)(2+\gamma_R)\tau_{R_N}}{2}$$

$$m_5^{N,N-1} = -2 - \alpha + \gamma_R(2+\gamma_R)\tau_{R_N}$$

$$m_5^{N,N-2} = 1 - \frac{\gamma_R(1+\gamma_R)\tau_{R_N}}{2}$$

$$m_5^{N-1,N-1} = 2\alpha + \frac{7\gamma_R - 4(1+\gamma_R^2) - \gamma_R^2(2+\gamma_R)(1+4\tau_{R_N})}{2(1+\gamma_R)}$$

$$m_5^{N-1,N-2} = 1 - \alpha - \frac{3\gamma_R}{2} + \frac{\gamma_R^2}{2} + \gamma_R^2\tau_{R_N}$$

$$m_5^{N-2,N-2} = 2\alpha + \frac{-4 + \gamma_R^2 - \gamma_R^3 - \gamma_R^2(1+\gamma_R)\tau_{R_N}}{2(2+\gamma_R)}.$$

The matrix M_1 is negative-definite and bounded away from 0 by $h^2\pi^2$ by the argument leading to eq. (2.4.31), see appendix to chapter 2 in [3]. M_2 is non-positive definite, see eq. (2.4.34) and (2.4.35) in that appendix. From (3.2), (3.3) and (3.8) follows that $c_k \geq 0$, $k = 1, \ldots, N$, therefore, the matrix M_3 is non-positive. For a given value of $0 \leq \alpha \leq 1$, τ_{L_1} and τ_{R_N} can be found such that the matrices M_4 and M_5 will be non-positive, for all γ_L and γ_R. For example: for $\alpha = 1/10$, $\tau_{L_1} = \tau_{R_N} = 4$; for $\alpha = 1/2$, $\tau_{L_1} = \tau_{R_N} = 9$ and for $\alpha = 8/10$, $\tau_{L_1} = \tau_{R_N} = 24$. This completes the proof that M is indeed a negative-definite matrix, bounded away from 0 by $\alpha\pi^2$. Therefore the norm of the error vector $\|\epsilon\|$ can grow at most linearly in time, see equation (2.11).

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