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A STABLE AND CONSERVATIVE INTERFACE TREATMENT OF ARBITRARY SPATIAL ACCURACY*

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Abstract. Stable and accurate interface conditions are derived for the linear advection-diffusion equation. The conditions are functionally independent of the spatial order of accuracy and rely only on the form of the discrete operator. We focus on high-order finite-difference operators that satisfy the summation-by-parts (SBP) property. We prove that stability is a natural consequence of the SBP operators used in conjunction with the new boundary conditions. In addition, we show that the interface treatments are conservative.

New finite-difference operators of spatial accuracy up to sixth order are constructed: these operators satisfy the SBP property. Finite-difference operators are shown to admit design accuracy (p^{th} -order global accuracy) when $(p - 1)^{th}$ -order stencil closures are used near the boundaries if the physical boundary conditions are implemented to at least p^{th} -order accuracy. Stability and accuracy are demonstrated on the nonlinear Burgers' equation for an twelve-subdomain problem with randomly distributed interfaces.

Key words. high-order finite-difference, numerical stability, interface conditions, summation-by-parts

Subject classification. Applied and Numerical Mathematics

1. Introduction. Higher order and spectral schemes are ideally suited for resolving problems for which high resolution is essential. Computational aeroacoustics (CAA) and computational electromagnetics (CEM) are two such fields that require high accuracy to resolve the vastly disparate length and time scales involved. High-order (spectral) schemes easily outperform low-order schemes on simple problems in which the physical domain is smoothly mapped onto the computational space. The spatial convergence rates of these schemes allow satisfactory results on relatively coarse grids.

At least two fundamental obstacles presently limit the use of high-order schemes. The first one is the lack of nonlinear robustness exhibited by high-order formulations. Under resolved features in the solution and inappropriate numerical and physical boundary conditions are the primary causes. A second limitation is the difficulty in applying high-order formulations to complex geometries. Often, the generation of a grid around a complex configuration is the most difficult aspect of the solution procedure. Further constraint of the grids so that they are smooth to higher order (necessary to attain design accuracy for high-order methods) severely complicates grid generation around complex configurations.

Many high-order practitioners advocate a fully unstructured approach to grid generation. This approach simplifies the grid-generation procedure considerably for complex configurations. Finite-element techniques are an example of the fully unstructured schemes that are routinely used on complex geometries. An alternative to fully unstructured methods is the semistructured approach, in which the solution domain

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is broken into the union of piecewise smooth subdomains. Each subdomain is discretized with a stable formulation, and the resulting multiple domains are patched together globally. This technique has been successfully used by Kopriva [1], and more recently by Hesthaven and Gottlieb [2].

The approach for designing interface conditions developed in this work is equally valid for the unstructured and semi-structured approaches in multiple spatial dimensions. The interface conditions are determined entirely by accurate left and right state data along the interface, and do not depend on the source of the data. For simplicity, however, we focus on the interface matching conditions necessary to maintain stability and accuracy in one spatial dimension. We demonstrate the technique for both spectral and high-order formulations.

In section 2, we define and describe semidiscrete operators that satisfy the SBP convention. In section 3, we introduce new interface boundary conditions for multiple domains. In section 4, we show that the new conditions are conservative across interfaces. In section 5, we consider specific examples of the stability and accuracy of finite-difference schemes. In section 6, we present the conclusions. Finally, in the appendix we present the stencils used for fourth- and sixth-order finite-difference schemes.

2. Spatial Discretizations. The stable interface conditions presented in this work are valid for spatial discretizations of arbitrary accuracy. To achieve this generality, the spatial discretizations must be of a specific form. Fortunately, most numerical schemes can be put into the required form with only minor modifications. To be more precise we consider discrete spatial derivative operators with the following properties:

2.1. First-derivative properties.

1. The first derivative operator defining the numerical derivative $\mathbf{u}_x = [(\frac{\partial u}{\partial x})_0, \dots, (\frac{\partial u}{\partial x})_N]^T$ is

$$(2.1) \quad \begin{aligned} P\mathbf{u}_x - Q\mathbf{u} &= 0 \\ P\mathbf{v}_x - Q\mathbf{v} &= P\mathbf{T}_e, \end{aligned}$$

where $\mathbf{u} = [u_0(t), u_1(t), \dots, u_N(t)]^T$, $\mathbf{v} = [v(x_0, t), \dots, v(x_N, t)]^T$ and $\mathbf{v}_x = [(\frac{\partial v}{\partial x})_0, \dots, (\frac{\partial v}{\partial x})_N]^T$. (The vector \mathbf{v} is the exact solution.) The truncation error \mathbf{T}_e satisfies $|\mathbf{T}_e| = O(\Delta x)^m$ where the quantity Δx is defined as the maximum distance between any two neighboring grid points.

2. The matrix P is symmetric and positive definite $(\Delta x)p I \leq P \leq (\Delta x)q I$ where p and q are independent of N with $p > 0$ and $q > 0$.
3. The matrix Q is nearly skew symmetric and satisfies the property $Q + Q^T = D$, where the diagonal matrix D has the form $d_{i,i} = [-1, 0, \dots, 0, 1]$ for $i = 0, 1, \dots, N$. Furthermore, $Q_{0,0} = -\frac{1}{2}$ and $Q_{N,N} = \frac{1}{2}$.

A spatial operator in the form of equation (2.1), which satisfies properties 1 through 3, is referred to as an SBP operator [3]. All SBP operators automatically lead to an energy estimate for periodic solutions to the linear advection-diffusion equation. In the finite-domain case, an energy estimate exists when an SBP operator is combined with specific boundary treatments.

Discretization operators that satisfy the SBP framework are remarkably general. Kreiss and Scherer [3] first suggested the use of SBP spatial operators in the context of second-order central-difference schemes. In Olsson [4][5][6] and Strand [7], high order finite difference operators are constructed based on spatial operators of SBP type. These resulting schemes are strictly stable which means that the growth rate of the analytic and semi-discrete solution is identical.

The precise properties of the matrices P and Q provide a constructive means of formulating boundary closures. A discretization begins with a parameterization of several points near the boundary of the required

where

$$(2.5) \quad \begin{aligned} Su|_0 &= v_x(x_0) + O(\Delta x)^r \\ Su|_N &= v_x(x_N) + O(\Delta x)^r \end{aligned}$$

The matrix S is the identity matrix (scaled by the grid spacing) where a discrete representation of the first derivative replaces the first and last rows.

4. The matrix P is that used in the first-derivative operator.

Explicit forms of the matrices S and M are given in the appendix for a second-order explicit discretization. In addition, the matrix S is presented up to sixth order.

3. Interface Boundary Conditions for Multiple Domains. Consider the linear advection-diffusion equation

$$(3.1) \quad \frac{\partial U}{\partial t} + a \frac{\partial U}{\partial x} = \epsilon \frac{\partial^2 U}{\partial x^2}, \quad |x| \leq 1, \quad t > 0.$$

Suppose that the equation is discretized by a multi-domain technique such that the interval is divided arbitrarily into two subintervals $-1 \leq x \leq x_i$ and $x_i \leq x \leq 1$. On each subinterval, a discretization is used that satisfies the SBP properties 1 through 3. We propose implementing the interface boundary conditions by using a penalty treatment of the form

$$(3.2) \quad \begin{aligned} P_l \mathbf{u}_t + a Q_l \mathbf{u} &= \epsilon R_l \mathbf{u} + \sigma_1 \mathbf{e}_{\mathbf{l}_i} (u|_{x=x_i} - v|_{x=x_i}) + \sigma_2 \epsilon \mathbf{e}_{\mathbf{l}_i} [(D_l \mathbf{u})|_{x=x_i} - (D_r \mathbf{v})|_{x=x_i}] \\ P_r \mathbf{v}_t + a Q_r \mathbf{v} &= \epsilon R_r \mathbf{v} + \sigma_3 \mathbf{e}_{\mathbf{r}_i} (v|_{x=x_i} - u|_{x=x_i}) + \sigma_4 \epsilon \mathbf{e}_{\mathbf{r}_i} [(D_r \mathbf{v})|_{x=x_i} - (D_l \mathbf{u})|_{x=x_i}] \end{aligned}$$

where \mathbf{u} is a vector of length M $\mathbf{u} = [u_0(t), u_1(t), \dots, u_M(t)]^T$ defined in the left domain at the points $\mathbf{x}_{\mathbf{L}} = [x_0 = -1, x_1, \dots, x_M = x_i]^T$ and $\mathbf{e}_{\mathbf{l}_i} = [0, \dots, 0, 1]^T$ is of dimension M . In the right domain, $\mathbf{v} = [v_0(t), v_1(t), \dots, v_N(t)]^T$ is defined at the points $\mathbf{x}_{\mathbf{R}} = [x_0 = x_i, x_1, \dots, x_N = 1]^T$ and $\mathbf{e}_{\mathbf{r}_i} = [1, 0, \dots, 0]^T$ is of dimension N .

The second-derivative matrices $P_l^{-1} R_L$ and $P_r^{-1} R_r$, as well as the first-derivative matrices $P_l^{-1} Q_l$ and $P_r^{-1} Q_r$, are defined as in section 2. The matrices D_l and D_r are any operators that approximate the first derivative to $O(\Delta x)^m$. The obvious first choice would be to use $P_l^{-1} Q_l$ and $P_r^{-1} Q_r$, but this choice is not essential for accuracy or stability. (In equation (3.2) we have ignored the physical boundary conditions at $x = 1$ and $x = -1$ for the sake of simplicity.)

THEOREM 3.1. *Consider the scheme (3.2) for the advection-diffusion equation (3.1). If the matrices P_l, Q_l, P_r, Q_r, R_L and R_r satisfy the first and second derivative properties of section 2 and*

$$(3.3) \quad \sigma_3 = \sigma_1 - a, \quad \sigma_4 = \sigma_2 + 1, \quad \sigma_1 \leq \frac{a}{2} - \epsilon \left[\frac{\sigma_2^2}{4\alpha_r} + \frac{\sigma_4^2}{4\alpha_L} \right],$$

then (3.2) is stable.

In the proof which follows, we have without loss of generality considered only the interface terms, and ignored the terms that arise at the physical boundaries. We assume that the physical boundary conditions are implemented by stable and accurate numerical procedures. (See Hesthaven and Gottlieb [2] for a possible implementation).

PROOF: The proof is based on a simple energy estimate. By premultiplying the equations by the vectors \mathbf{u}^T and \mathbf{v}^T , respectively, and adding we obtain

$$\frac{d}{dt} [\|\mathbf{u}\|_{P_l}^2 + \|\mathbf{v}\|_{P_r}^2] = 2\mathbf{u}^T (\epsilon R_l - a Q_l) \mathbf{u} + 2\mathbf{v}^T (\epsilon R_r - a Q_r) \mathbf{v}$$

$$\begin{aligned}
& + 2\sigma_1 u_i(u_i - v_i) + 2\epsilon\sigma_2 u_i[(D_l u)_i - (D_r v)_i] \\
& + 2\sigma_3 v_i(v_i - u_i) + 2\epsilon\sigma_4 v_i[(D_r v)_i - (D_l u)_i]
\end{aligned}$$

where $\|u\|_{P_l}^2 = \mathbf{u}^T P_l \mathbf{u}$, and we have defined $u_i, v_i, (D_l u)_i$, and $(D_l v)_i$ as $u|_{x=x_i}, v|_{x=x_i}, (D_l \mathbf{u})|_{x=x_i}$, and $(D_r \mathbf{v})|_{x=x_i}$, respectively. The second-derivative properties of section 2 lead to

$$(3.4) \quad \mathbf{u}^T R_l \mathbf{u} \leq -\alpha_l (D_l u)_i^2 + u_i (D_l u)_i$$

$$(3.5) \quad \mathbf{v}^T R_r \mathbf{v} \leq -\alpha_r (D_r v)_i^2 - v_i (D_r v)_i$$

where the constants α_l and α_r are positive.

By using the first-derivative properties of section 2 and equations (3.4) and (3.5) and neglecting the physical boundary terms leads to

$$(3.6) \quad \frac{d}{dt} [\|u\|_{P_l}^2 + \|v\|_{P_r}^2] \leq \mathbf{w}_i^T B \mathbf{w}_i$$

where $\mathbf{w}_i = [u_i, v_i, (D_l u)_i, (D_r v)_i]$, and the boundary matrix B defined by

$$(3.7) \quad B = \begin{bmatrix} (-a + 2\sigma_1) & -(\sigma_1 + \sigma_3) & \epsilon(1 + \sigma_2) & -\epsilon\sigma_2 \\ -(\sigma_1 + \sigma_3) & a + 2\sigma_3 & -\epsilon\sigma_4 & \epsilon(-1 + \sigma_4) \\ \epsilon(1 + \sigma_2) & -\epsilon\sigma_4 & -2\epsilon\alpha_l & 0 \\ -\epsilon\sigma_2 & \epsilon(-1 + \sigma_4) & 0 & -2\epsilon\alpha_r \end{bmatrix}$$

Straightforward (though tedious) algebra shows that conditions (3.3) yield a non-positive definite matrix B , thus proving stability. Details are presented in Appendix I.

In practice, the values of σ_1 through σ_4 are determined as follows. The parameters α_r and α_l are functions from the numerical method and the chosen grid. The viscous contribution in the constraint equation $\sigma_1 \leq \frac{a}{2} - \epsilon[\frac{\sigma_2^2}{4\alpha_r} + \frac{\sigma_4^2}{4\alpha_l}]$ is minimized for $\sigma_2 = \frac{-\alpha_r}{\alpha_r + \alpha_l}$, yielding the expression $\sigma_1 \leq \frac{a}{2} - \epsilon[\frac{1}{4(\alpha_r + \alpha_l)}]$. The value σ_1 determines the dissipation at the interface, and also influences the effective CFL of the numerical scheme. Values of σ_1 in the range $-1 \leq \sigma_1 \leq 0$ provide a compromise between adequate levels of dissipation, and acceptable numerical efficiency.

We have shown that the linking of two domains at an interface with the interface conditions prescribed in Theorem 3.1 is stable in a semidiscrete sense for specific values of the penalty parameters σ_1 through σ_4 . The basic methodology can be extended to an arbitrary number of subdomains without complication. The only constraint is that the numerical method must satisfy the SBP framework. The methodology does not rely on subdomain size and does not require the same SBP operator to be used in each domain. In principle, a finite-difference operator of any order, as well as spectral operators on subdomains of arbitrary size, can be linked together in a stable manner. Practical details on how to chose σ_1 through σ_4 are included in the results section (Section 6).

In section 2, we presented the general form of second-derivative operators appropriate for this work. We then noted two specific derivative operators that satisfy this form. We now show that both choices for the matrices R_l (and R_r) suggested in section 2 satisfy conditions (3.4) and (3.5) of Theorem 3.1. We start with the first option (i.e. $R_l = Q_l P_l^{-1} Q_l$). In this case, the first derivative matrix in (3.2) is $D_l = P_l^{-1} Q_l$. Thus, the quantity $\mathbf{u}^T R_l \mathbf{u}$ becomes

$$\begin{aligned}
\mathbf{u}^T Q_l P_l^{-1} Q_l \mathbf{u} &= \mathbf{u}^T Q_l P_l^{-1} P_l P_l^{-1} Q_l \mathbf{u} \\
&= -(P_l^{-1} Q_l \mathbf{u})^T P_l (P_l^{-1} Q_l \mathbf{u}) + u_i (P_l^{-1} Q_l u)_i
\end{aligned}$$

where we have used the SBP property $Q + Q^T = D$, and have ignored the physical boundary contribution.

We recall now that $P_l \geq (\Delta x)p_l$ so that

$$\begin{aligned} \mathbf{u}^T R_l \mathbf{u} &= \mathbf{u}^T Q_l P_l^{-1} Q_l \mathbf{u} \\ &\leq -(\Delta x)p_l |(D_l \mathbf{u})|^2 + u_i (D_l u)_i \end{aligned}$$

Thus, (3.4) is satisfied with $\alpha_l = (\Delta x)p_l$. A similar result holds for R_r with $\alpha_r = (\Delta x)p_r$.

The second choice presented in section 2 for the second-derivative operator $P^{-1}R_l$ is of the form of equation (2.3):

$$P^{-1}R_l = P^{-1}(-S^T M + D)S$$

For the purpose of proving stability, we relate the two matrices $D_l = S$. (In actuality, only the first and last rows satisfy $D_l = S$. They are, however, the only portions of the matrices that enter the proof.)

$$\begin{aligned} \mathbf{u}^T R_l \mathbf{u} &= -(S\mathbf{u})^T M S\mathbf{u} + U_i (S\mathbf{u})_i \\ &\leq -(\Delta x) m |S\mathbf{u}|^2 + U_i (S\mathbf{u})_i. \end{aligned}$$

Thus, (3.4) is satisfied with $\alpha_l = (\Delta x) m$.

4. Conservation at the Interface. The Lax-Wendroff theorem [12] addresses the complexities encountered in solving nonlinear conservation laws. The theorem states that a convergent numerical approximation $U_l(x, t)$, computed with a consistent and **conservative** method, converges to a weak solution of the conservation law. Note that discrete conservation is necessary to satisfy the conditions of the theorem.

A heuristic definition of conservation (commonly encountered by practitioners) describes how the numerical flux function “telescopes” across a domain to the boundaries. The total quantity of a conserved variable in any region changes only as a result of the flux through the boundaries of the region. We, however, rely on a broader definition of conservation motivated by the original proof of the Lax-Wendroff theorem. We demand that the numerical flux telescope across the domain, and that all moments of the flux against an arbitrary test function telescope across the domain. This additional constraint demands an equivalence between the weak forms of the continuous and discrete operators.

We begin by discussing conservation in a single domain. Consider the nonlinear equation $U_t + F_x = 0$ on $-1 \leq x \leq 1$ and $t \geq 0$. Note that in the linear case $F = aU$ and we obtain (3.1) with $\epsilon = 0$. To obtain the weak form of this equation we multiply by an arbitrary test function $\phi(x, t)$ that vanishes on the boundaries. By integrating with respect to space and time we obtain an integral statement of the original differential equation:

$$\int_{-1}^1 \phi U dx \Big|_0^t - \int_0^t \int_{-1}^1 (U \phi_t + F \phi_x) dx d\tau = 0$$

Now consider the semidiscrete equation given by $P\mathbf{U}_t + Q\mathbf{F} = 0$. Here, we have replaced the spatial derivative F_x in the continuous case with an SBP derivative operator of order $(\Delta x)^r$. By multiplying by the discrete vector $\phi(x_j) = \phi^T$ (the discrete analog of integration) and integrating with respect to time, we obtain

$$\phi^T P\mathbf{U} \Big|_0^t - \int_0^t (U^T P \phi_t + F^T Q \phi) d\tau = 0$$

Thus, the semi-discrete operator satisfies a weak form similar to that of the continuous operator, and asymptotically approaches the continuous operator in the limit of infinite spatial resolution. The special form of the

P and Q matrices present in the SBP operators enables the semidiscrete operator to mimic the conservation property of the continuous operator.

The equivalence between the continuous and semi-discrete operators is more more complicated for multiple domains. The conservation property of the SBP operator does not necessarily apply at an interface boundary. Under very mild restrictions, however, the SBP interface operators telescope out to the physical boundaries, as does the continuous operator. Because conservation is only necessary for the advection terms in the advection-diffusion equation, we set $\epsilon = 0$ (see equation (3.1)) and prove conservation for a two-domain discretization. We prove conservation for a general nonlinear flux. Note that the penalty parameters for this nonlinear case are designated $\hat{\sigma}_1$ and $\hat{\sigma}_3$. The resulting conservation condition obtained in the nonlinear case is slightly different from that obtained in the linear analysis. This difference results from different scalings of the penalty parameters.

THEOREM 4.1. *Assume the nonlinear equation $\frac{\partial U}{\partial t} + \frac{\partial F(U)}{\partial x} = 0$ is valid on the interval $-1 \leq x \leq 1$, $t > 0$, divided arbitrarily into two subintervals $-1 \leq x \leq x_i$ and $x_i \leq x \leq 1$. On each subinterval, a discretization is used that satisfies the SBP framework, and boundary conditions are imposed via penalties in the form*

$$(4.1) \quad \begin{aligned} \mathbf{u}_t + P_l^{-1} Q_l \mathbf{F}(\mathbf{u}) &= \hat{\sigma}_1 P_l^{-1} \mathbf{e}_{l_i} [F(u(x_i)) - F(v(x_i))] \\ \mathbf{v}_t + P_r^{-1} Q_r \mathbf{F}(\mathbf{v}) &= \hat{\sigma}_3 P_r^{-1} \mathbf{e}_{r_i} [F(v(x_i)) - F(u(x_i))] \end{aligned}$$

where $\mathbf{u} = [u_0(t), u_1(t), \dots, u_M(t)]^T$ is defined in the left domain at the points $\mathbf{x}_L = [x_0 = -1, x_1, \dots, x_M = x_i]^T$, and $\mathbf{e}_{l_i} = [0, \dots, 0, 1]^T$ is of dimension M , with similar definitions on the right domain. The discretization is conservative provided that the stability condition $\hat{\sigma}_3 = \hat{\sigma}_1 - 1$ is satisfied.

PROOF: For multiple domains, we proceed as shown previously in the single-domain case. Multiplying equations (4.1) by the vectors $\phi^T P_l$ and $\phi^T P_r$, respectively, yields the set of equations

$$\begin{aligned} \phi^T P_l \mathbf{u}_t + \phi^T Q_l \mathbf{F}(\mathbf{u}) &= \hat{\sigma}_1 \phi(x_i) (F(u(x_i)) - F(v(x_i))) \\ \phi^T P_r \mathbf{v}_t + \phi^T Q_r \mathbf{F}(\mathbf{v}) &= \hat{\sigma}_3 \phi(x_i) (F(v(x_i)) - F(u(x_i))) \end{aligned}$$

Using the properties of Q_l and Q_r we get

$$\begin{aligned} \phi^T P_l \mathbf{u}_t - \mathbf{F}^T Q_l \phi + \phi(x_i) F(u(x_i)) &= \hat{\sigma}_1 \phi(x_i) (F(u(x_i)) - F(v(x_i))) \\ \phi^T P_r \mathbf{v}_t - \mathbf{F}^T Q_r \phi - \phi(x_i) F(v(x_i)) &= \hat{\sigma}_3 \phi(x_i) (F(v(x_i)) - F(u(x_i))) \end{aligned}$$

By integrating with respect to time and making use of the fact that ϕ is continuous at the interface, we get

$$\begin{aligned} \phi^T P_l \mathbf{u}|_0^t + \phi^T P_r \mathbf{v}|_0^t &= \int_0^t (\mathbf{u}^T P_l \phi_t + \mathbf{F}^T Q_l \phi) d\tau \\ &+ \int_0^t (\mathbf{v}^T P_r \phi_t + \mathbf{F}^T Q_r \phi) d\tau \\ &+ \int_0^t \phi_i F(u(x_i)) (\hat{\sigma}_1 - \hat{\sigma}_3 - 1) d\tau \\ &+ \int_0^t \phi_i F(v(x_i)) (\hat{\sigma}_3 - \hat{\sigma}_1 + 1) d\tau \end{aligned}$$

Obviously, the condition $\hat{\sigma}_3 = \hat{\sigma}_1 - 1$ eliminates the interface terms from the expression and leaves the desired weak form of the semidiscrete equation. Thus, the theorem is proved.

5. Accuracy of Boundary Conditions. A significant obstacle in dealing with high-order finite-difference schemes is the formulation of stable stencils near the boundaries. A uniformly high-order approximation should be maintained if possible up to the boundary. In most high-order formulations, ensuring uniform accuracy up to the boundaries is difficult when numerical stability must be maintained. Fortunately, Gustafsson [13] showed that difference approximations to mixed hyperbolic parabolic equations admit global design accuracy when a finite number of points (independent of N) are closed with stencils that are less accurate by 1 order. For example, a fourth-order interior discretization will asymptotically recover fourth-order L_2 accuracy with third-order closures near the boundaries.

In this section, we confirm that the *physical* boundary conditions must be imposed with at least the design accuracy in the context of interface boundary conditions. We begin by inspecting equation (3.2) and by defining the truncation error as that error committed by substituting the *exact solution* into the scheme. Denote by \mathbf{V}_l and \mathbf{V}_r the projection of the exact solution in the two domains. Substituting the exact solution into the first equation in (3.2) yields

$$P_l \mathbf{T}_{1e} = P_l \frac{\partial \mathbf{V}_l}{\partial t} + a Q_l \mathbf{V}_l - \epsilon R_l \mathbf{V}_l + \sigma_1 \mathbf{e}_{1i} (\mathbf{V}_{l,x=x_i} - \mathbf{V}_{r,x=x_i}) + \sigma_2 \mathbf{e}_{1i} ((D_l \mathbf{V}_l)|_{x=x_i} - (D_r \mathbf{V}_r)|_{x=x_i})$$

with a similar expression in the right domain. The differentiation matrices are accurate to the design order of the method. Thus, the first three terms to the right of the equality, reduce to the truncation error of the spatial approximation. (Except for a finite number of points that are lower by 1 order near the interfaces and the physical boundary). Examining the truncation error from the penalty terms, we observe that V is smooth across the interface, and $\mathbf{V}_{l_i} - \mathbf{V}_{r_i} = 0$. Thus, we only need that $D_l \mathbf{V}_{l,x=x_i}$ and $D_r \mathbf{V}_{r,x=x_i}$ approximate the first derivative to the design order of accuracy. The exact nature of the solution error near the boundaries is extremely complicated due to the points treated less accurately in that vicinity. More details will be presented in a future work on this subject. We show by numerical example, however, that order reduction occurs when the interface derivative is treated with less than design accuracy. (See Table 5).

5.1. Uniform Grid.. Now we demonstrate that the physical boundary conditions must be imposed with accuracy of at least design order to maintain global design accuracy. This condition is a natural consequence of the overall dependence of the solution on the boundary conditions. The test problem we use is the Burgers' equation

$$(5.1) \quad \frac{\partial U}{\partial t} + U \frac{\partial U}{\partial x} = \epsilon \frac{\partial^2 U}{\partial x^2} \quad -1 \leq x \leq 1, t > 0,$$

with the exact solution

$$(5.2) \quad U(x, t) = -a \tanh\left(a \frac{x - ct}{2\epsilon}\right) + c, -\infty < x < \infty, t < 0.$$

The solution of (5.1) requires imposition of boundary conditions at each end of the physical domain. We choose Robin boundary conditions of the form

$$\alpha u(-1, t) - \beta \frac{\partial u}{\partial x}|_{-1} = g_{-1}(t); \quad \gamma u(1, t) - \delta \frac{\partial u}{\partial x}|_1 = g_1(t)$$

such that the problem is mathematically well-posed. (See Hesthaven and Gottlieb [2] for the constraints on α, β, γ , and δ). The physical boundary conditions were imposed in penalty form, as described in the work of Hesthaven and Gottlieb [2]. The time-advancement scheme is the five-stage fourth-order low-storage Runge-Kutta scheme. The time step was chosen to ensure that the temporal error in the formulation was

small relative to the spatial error. The simulation is run to a physical time of $T = 1$, and the viscosity is determined by the value $\epsilon = 5 \cdot 10^{-1}$.

Tables 1 to 4 show the results of a grid-refinement study on a single domain with a fourth-order explicit interior scheme. The accuracy of the boundary closure and of the physical boundary condition are parameters in the study. Table 2 shows the results of the refinement study with a uniformly fourth-order-accurate scheme (4,4-4-4,4) with the derivative term in the Robins' boundary conditions approximated to $O(\Delta x^4)$. We note

TABLE 5.1
L₂ Solution Errors: Convergence rate of uniformly fourth-order scheme

N	$LOG_{10}error$	Rate
33	-3.847	
65	-4.082	2.31
129	-5.239	3.84
257	-6.486	4.14
513	-7.731	4.14
1025	-8.960	4.87

that the convergence rate in Table 1 is fourth order and that the design accuracy is achieved.

Table 2 shows the second study in which boundary closure accuracy is relaxed by one order. The resulting scheme (3,3-4-3,3) is third order locally at each boundary and fourth order in the interior. (Both the inviscid and viscous stencils are reduced by one order of accuracy near the boundaries.) The physical boundary condition is still approximated to $O(\Delta x^4)$.

TABLE 5.2
L₂ Solution Errors: Convergence rate of fourth-order scheme with third-order closure at boundaries.

N	$LOG_{10}error$	Rate
33	-3.694	
65	-4.797	3.66
129	-5.971	3.90
257	-6.117	3.81
513	-7.276	3.85
1025	-9.455	3.92

We note that the convergence rate in Table 2 asymptotes to fourth order and that the absolute levels of error are comparable to those obtained using the (4,4-4-4,4) scheme. Again, design accuracy is achieved.

Table 3 shows the third study, in which boundary closure accuracy is relaxed by two orders. The resulting scheme (2,2-4-2,2) is second order locally at each boundary and fourth order in the interior. (Only the viscous terms are reduced by two orders of accuracy near the boundaries.) The physical boundary condition is still approximated to $O(\Delta x^4)$.

We note that the convergence rate in Table 3 asymptotes to third order, which is a reduction in global accuracy of one order. This behavior is consistent with Gustafsson's [13] theory, specifically, that global solution accuracy allows a finite number of stencils to be reduced by one order of accuracy.

Table 4 shows the final study, in which boundary closure accuracy is uniformly fourth-order accurate (4,4-4-4,4). The physical boundary condition is approximated to $O(\Delta x^3)$, however. The convergence rate in

TABLE 5.3

L_2 Solution Errors: Convergence rate of fourth-order scheme with second-order closure at boundaries.

N	$LOG_{10}error$	Rate
33	-2.974	
65	-4.074	3.65
129	-5.519	4.80
257	-6.284	2.54
513	-7.048	2.54
1025	-7.898	2.82

Table 4 asymptotes to third order, which is a reduction in global accuracy by one order.

TABLE 5.4

L_2 Solution Errors: Convergence rate of uniformly fourth-order scheme, using third-order accurate boundary conditions.

N	$LOG_{10}error$	Rate
33	-3.004	
65	-4.002	3.32
129	-4.764	2.53
257	-5.636	2.90
513	-6.531	2.97
1025	-7.898	2.82

This series of tests on the single domain indicates the need to impose the *physical* boundary condition with design accuracy. However, closing the near boundary stencils with an accuracy that is one order less than the design interior accuracy appears to be sufficient. A similar conclusion was reached with a second-order-accurate scheme (1-2-1) and second-order physical boundary conditions.

We now demonstrate by numerical example that these results generalize to the case of multiple domains. Table 5 shows a grid-refinement study that compares one and eight spatial domains. The numerical test problem is the previously described Burgers' equation using a value of $\epsilon = 10^{-2}$. The numerical scheme used in both cases is the (3,3,3,3-4-3,3,3,3) scheme with physical boundary conditions imposed to an accuracy of $O(\Delta x^4)$.

TABLE 5.5

L_2 Solution Errors: Convergence rate of fourth-order scheme with third-order closure at interfaces, on multiple domain problem.

N	1 domains		8 domains	
	$LOG_{10}error$	Rate	$LOG_{10}error$	Rate
97	-2.148		-2.125	
193	-3.016	2.88	-3.143	3.38
385	-4.214	3.98	-4.485	4.45
769	-5.372	3.85	-5.656	3.38
1537	-6.505	3.76	-6.866	4.02
3063	-7.664	3.85	-8.055	3.95

We note that the convergence rate in Table 5 asymptotes to fourth order, for both the one- and eight-domain cases. This example demonstrates that design accuracy is achieved with multiple domains so long as the physical boundary conditions are imposed with design accuracy and the numerical closures near the interfaces are at most one order of accuracy less than the design accuracy of the interior scheme.

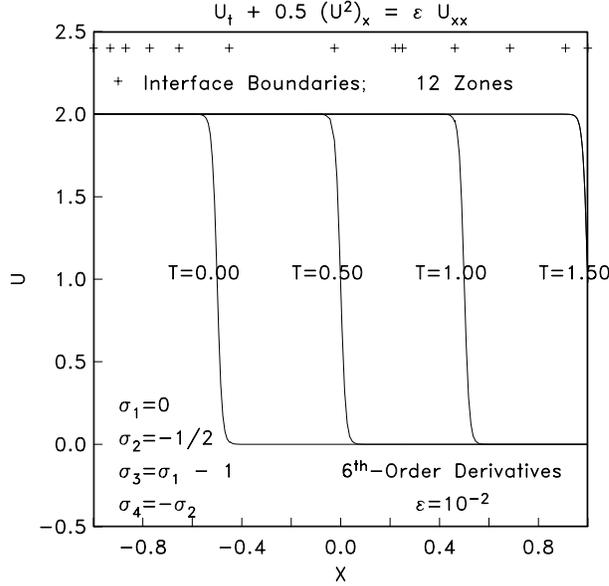


FIG. 5.1. *The Burgers equation solved using a sixth-order scheme with randomly generated interface points.*

5.2. Nonuniform Domain. The final problem we solve is the nonlinear Burgers’ equation with unequally spaced subdomains and a sixth-order scheme. Details of the numerical discretization are included in the appendix. The Burgers’ equation in the form of equation (5.1) is solved throughout the domain with a viscosity parameter of $\epsilon = 10^{-2}$. The domain is divided into 12 subdomains, each with the same number of points and a uniform local discretization. The domain interfaces are placed randomly throughout the domain. The ratio of maximum to minimum subdomain size is about 15:1. Figure 1 shows the solution at four different times. The “symbols” at the top of the figure show the positions of the 11 interface points. The profiles are smooth and monotone for this discretization. Figure 2 shows the logarithm of the solution error plotted as a function of space on the sequence of five grids.

This problem demonstrates the stability and accuracy of the new interface treatments. The discretizations asymptote to a convergence rate of sixth order on the sequence of grids. Table 6 shows the convergence rate of the calculations, for two different values of the parameter ϵ . The steep gradients are resolved to high-order on all grids for $\epsilon = 10^{-2}$. For $\epsilon = 2 \cdot 10^{-3}$, the two coarsest grids are not yet achieving high-order accuracy, and two-point grid oscillations exist in the solution. Further reduction of ϵ causes numerical instability, emanating from the interface location, as the gradients pass the interface. Increasing the robustness of the interface conditions for marginally resolved/discontinuous cases is the focus of current research.

6. Conclusions. We focus on high-order finite difference schemes, which satisfy the summation-by-parts (SBP) discretization framework. We show stable and conservative interface treatments of arbitrary spatial accuracy for the linear advection-diffusion equation. Problems with multiple domains and abruptly changing mesh sizes are considered.

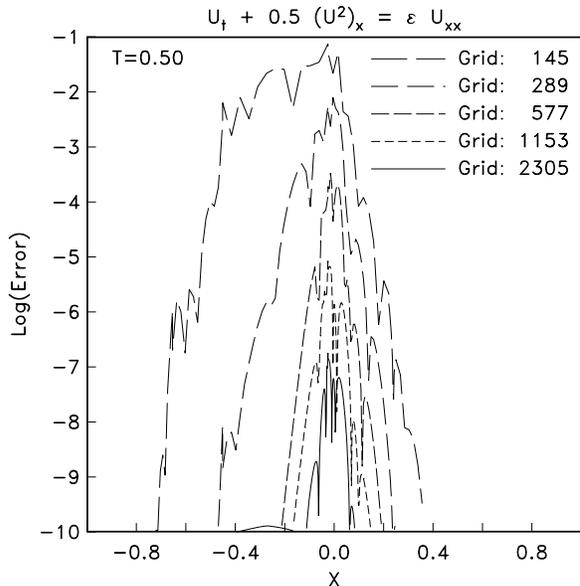


FIG. 5.2. Errors obtained from Burgers equation solved on a sequence of grids with a sixth-order scheme.

TABLE 5.6

L_2 Solution Errors: Convergence of sixth-order scheme with twelve subdomains and interfaces distributed randomly.

N	$\epsilon = 10^{-2}$		$\epsilon = 2 \cdot 10^{-3}$	
	$LOG_{10}error$	Rate	$LOG_{10}error$	Rate
145	-3.090		-1.376	
289	-4.641	5.15	-1.865	1.62
577	-5.915	4.22	-3.053	3.95
1153	-7.520	5.33	-4.574	5.05
2305	-9.370	6.15	-5.834	4.18

Finite-difference operators are shown to admit design accuracy (p^{th} -order global accuracy) when $p - 1^{th}$ -order stencil closures are used near boundaries if the physical boundary conditions are imposed with p^{th} -order accuracy. Finite-difference operators of up to sixth order are constructed which satisfy the constraints of the new interface procedures.

Accurate sixth order calculations are achieved for the nonlinear Burgers equation on a twelve subdomain problem having randomly distributed interfaces.

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Appendix I. Stability. Here we show the algebra involved in proving THEOREM 3.1. We begin by restating of the stability condition presented in equation (3.6), governing the total energy of the system:

$$(1) \quad \frac{d}{dt} [\|u\|_{P_l}^2 + \|v\|_{P_r}^2] \leq \mathbf{w}_i^T B \mathbf{w}_i$$

where $\mathbf{w}_i = [u_i, v_i, (D_l u)_i, (D_r v)_i]$, and the boundary matrix defined in equation (3.7) is defined by

$$(2) \quad B = \begin{bmatrix} (-a + 2\sigma_1) & -(\sigma_1 + \sigma_3) & \epsilon(1 + \sigma_2) & -\epsilon\sigma_2 \\ -(\sigma_1 + \sigma_3) & a + 2\sigma_3 & -\epsilon\sigma_4 & \epsilon(-1 + \sigma_4) \\ \epsilon(1 + \sigma_2) & -\epsilon\sigma_4 & -2\epsilon\alpha_l & 0 \\ -\epsilon\sigma_2 & \epsilon(-1 + \sigma_4) & 0 & -2\epsilon\alpha_r \end{bmatrix}$$

The stability of this matrix is easier to analyze if it is rotated with a similarity transformation. Define the new vector $\hat{\mathbf{w}} = S \mathbf{w}$ such that:

$$(3) \quad \hat{\mathbf{w}} = \frac{1}{\sqrt{2}} \begin{bmatrix} u_i - v_i \\ u_i + v_i \\ (D_l u)_i - (D_r v)_i \\ (D_l u)_i + (D_r v)_i \end{bmatrix} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & -1 \\ 0 & 0 & 1 & 1 \end{bmatrix} \begin{bmatrix} u_i \\ v_i \\ (D_l u)_i \\ (D_r v)_i \end{bmatrix}$$

The similarity rotation matrix has the property $S^T = S^{-1}$ as can easily be verified. The rotation matrix S can be used to transform the stability condition defined by equation (3.6) into the following equivalent condition:

$$(4) \quad \mathbf{w}_i^T M^i \mathbf{w}_i = \mathbf{w}_i^T S^T S M^i S^T S \mathbf{w}_i = \hat{\mathbf{w}}^T \hat{M}^i \hat{\mathbf{w}} \leq 0;$$

where

$$(5) \quad \hat{M}^i = \begin{bmatrix} 2(\sigma_1 + \sigma_3) & -(-\sigma_1 + \sigma_3 + a) & \epsilon(\sigma_2 + \sigma_4) & \epsilon \\ -(-\sigma_1 + \sigma_3 + a) & 0 & -\epsilon(-\sigma_2 + \sigma_4 - 1) & 0 \\ \epsilon(\sigma_2 + \sigma_4) & -\epsilon(-\sigma_2 + \sigma_4 - 1) & -\epsilon(\alpha_r + \alpha_l) & \epsilon(\alpha_r - \alpha_l) \\ \epsilon & 0 & \epsilon(\alpha_r - \alpha_l) & -\epsilon(\alpha_r + \alpha_l) \end{bmatrix}$$

To ensure negative definiteness, every sub-matrix in the matrix \hat{M}^i must be negative definite. We observe by inspection that $(\sigma_1 + \sigma_3) \leq 0$ is a necessary condition. Analyzing the 2×2 sub-matrices along the diagonal, we obtain the necessary conditions $(-\sigma_1 + \sigma_3 + a) = 0$, and $\epsilon(-\sigma_2 + \sigma_4 - 1) = 0$. Substituting the equalities $(-\sigma_1 + \sigma_3 + a) = 0$ and $(-\sigma_2 + \sigma_4 - 1) = 0$ into the matrix \hat{M}^i yields:

$$(6) \quad \hat{M}^i = \begin{bmatrix} 2(2\sigma_1 - a) & 0 & \epsilon(2\sigma_2 + 1) & \epsilon \\ 0 & 0 & 0 & 0 \\ \epsilon(2\sigma_2 + 1) & 0 & -\epsilon(\alpha_r + \alpha_l) & \epsilon(\alpha_r - \alpha_l) \\ \epsilon & 0 & \epsilon(\alpha_r - \alpha_l) & -\epsilon(\alpha_r + \alpha_l) \end{bmatrix}$$

A symmetric matrix can be rotated into diagonal form by an orthogonal matrix, making the condition of negative semi-definiteness

$$\hat{\mathbf{w}}^T \hat{U}^T D^i \hat{U} \hat{\mathbf{w}} \leq 0;$$

where \hat{U} is the orthogonal matrix that satisfies $\hat{U}^T D^i \hat{U} = \hat{M}^i$. Pre- and post- multiplication of \hat{M}^i by suitable rotation matrices $M_\lambda = R_1^T \hat{M}^i R_1$, yield the equivalent condition

$$\hat{\mathbf{w}}^T R_1^T \hat{U}^T D^i \hat{U} R_1 \hat{\mathbf{w}} \leq 0;$$

The matrix R_1 , chosen to yield a diagonal expression for the matrix M_λ is

$$(7) \quad L_1 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ L_{3,1} & 0 & 1 & 0 \\ L_{4,1} & 0 & L_{4,3} & 1 \end{bmatrix}$$

with

$$\begin{aligned} L_{3,1} &= \frac{-\epsilon(2\sigma_2 + 1)}{2(2\sigma_1 - a)} \\ L_{4,1} &= \frac{-2\epsilon(\alpha_r \sigma_2 - \alpha_l \sigma_2 + \alpha_r)}{\epsilon(4\sigma_2^2 + 4\sigma_2 + 1) + (4\sigma_1 - 2a)(\alpha_r + \alpha_l)} \\ L_{4,3} &= \frac{-\epsilon(2\sigma_2 + 1) + (-4\sigma_1 + 2a)(\alpha_r - \alpha_l)}{\epsilon(4\sigma_2^2 + 4\sigma_2 + 1) + (4\sigma_1 - 2a)(\alpha_r + \alpha_l)} \end{aligned}$$

The diagonal elements of M_λ are

$$\lambda_1 = 2(2\sigma_1 - a)$$

The symmetric P and nearly skew-symmetric Q matrices have the entries $A_1 = P^{-1}Q$, where

$$\begin{aligned}
(19) \quad p(1, 1) &= \frac{-(14400 r_2 + 302400 r_1 - 7420003)}{36288000} \\
p(1, 2) &= \frac{-(75600 r_3 + 1497600 r_2 + 11944800 r_1 - 59330023)}{21722800} \\
p(1, 3) &= \frac{-(9450 r_3 + 202050 r_2 + 1776600 r_1 - 7225847)}{340200} \\
p(1, 4) &= \frac{(900 r_2 + 18900 r_1 - 649)}{226800} \\
p(1, 5) &= \frac{(86400 r_3 + 1828800 r_2 + 15854400 r_1 - 66150023)}{3110400} \\
p(1, 6) &= \frac{(378000 r_3 + 7747200 r_2 + 65167200 r_1 - 279318239)}{188640000} \\
p(2, 2) &= \frac{(302400 r_3 + 6091200 r_2 + 49896000 r_1 - 210294289)}{7257600} \\
p(2, 3) &= \frac{(3780 r_3 + 82575 r_2 + 741825 r_1 - 2991977)}{34020} \\
p(2, 4) &= \frac{(5400 r_3 + 104400 r_2 + 810000 r_1 - 3756643)}{129600} \\
p(2, 5) &= \frac{-(529200 r_3 + 11107200 r_2 + 95508000 r_1 - 400851749)}{2419200} \\
p(2, 6) &= \frac{(86400 r_3 + 1828800 r_2 + 15854400 r_1 - 65966279)}{3110400} \\
p(3, 3) &= \frac{-(51300 r_3 + 1094400 r_2 + 9585000 r_1 - 39593423)}{64800} \\
p(3, 4) &= \frac{(120960 r_3 + 2584800 r_2 + 22680000 r_1 - 93310367)}{181440} \\
p(3, 5) &= \frac{(5400 r_3 + 104400 r_2 + 810000 r_1 - 3766003)}{129600} \\
p(3, 6) &= \frac{(900 r_2 + 18900 r_1 - 37217)}{226800} \\
p(4, 4) &= \frac{-(17100 r_3 + 364800 r_2 + 3195000 r_1 - 13184701)}{21600} \\
p(4, 5) &= \frac{(3780 r_3 + 82575 r_2 + 741825 r_1 - 2976857)}{34020} \\
p(4, 6) &= \frac{-(1890 r_3 + 40410 r_2 + 355320 r_1 - 1458223)}{68040} \\
p(5, 5) &= \frac{(302400 r_3 + 6091200 r_2 + 49896000 r_1 - 213056209)}{7257600} \\
p(5, 6) &= \frac{-(75600 r_3 + 1497600 r_2 + 11944800 r_1 - 54185191)}{21722800} \\
p(6, 6) &= \frac{-(14400 r_2 + 302400 r_1 - 36797603)}{36288000} \\
q(1, 1) &= \frac{(-1)}{2} \\
q(1, 2) &= \frac{(415800 r_3 + 8604000 r_2 + 72954000 r_1 - 283104553)}{32659200} \\
q(1, 3) &= \frac{(120960 r_3 + 2672640 r_2 + 24192000 r_1 - 100358119)}{6531840} \\
q(1, 4) &= \frac{-(25200 r_3 + 542400 r_2 + 4788000 r_1 - 19717139)}{403200}
\end{aligned}$$

