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A CARTESIAN GRID EMBEDDED BOUNDARY METHOD FOR THE COMPRESSIBLE NAVIER–STOKES EQUATIONS

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We present an unsplit method for the time-dependent compressible Navier–Stokes equations in two and three dimensions. We use a conservative, second-order Godunov algorithm. We use a Cartesian grid, embedded boundary method to resolve complex boundaries. We solve for viscous and conductive terms with a second-order semiimplicit algorithm. We demonstrate second-order accuracy in solutions of smooth problems in smooth geometries and demonstrate robust behavior for strongly discontinuous initial conditions in complex geometries.

1. Introduction

In this paper, we present an unsplit method for the time-dependent compressible Navier–Stokes equations in two and three dimensions. This algorithm is an extension of the algorithm in [9] to flows with viscous and thermal diffusion. The Navier–Stokes equations contain parabolic terms that arise from conductivity and viscosity. There are several methods to advance these terms. In [10], for example, a kinetic energy equation is evolved to get a stable approximation to the viscous term in the energy equation. This solution is elegant but also difficult to extend to multiple dimensions. We use a conservative, semiimplicit method in which the hyperbolic terms are advanced explicitly and the parabolic terms advanced implicitly. This approach to the compressible Navier–Stokes equations has been used without embedded boundaries [3; 30; 16; 14; 11]. Our algorithm follows the basic outline in the mapped grid algorithm presented in [30], in which the velocity and temperature evolution are split. They use a Crank–Nicolson time evolution with the energy-momentum coupling term treated explicitly. We use a hybrid approach to energy-momentum coupling. Also, since Crank–Nicolson has been shown to be marginally

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stable in certain cases [23], we use the L_0 -stable algorithm presented in [31] for elliptic coupling. We present our changes to the [31] algorithm that were necessary to make the linear equations tractable in the presence of small cells. This algorithm has been implemented with adaptive mesh refinement (AMR) as described in [4; 2]. All our cut cells are refined to the finest level, reducing all coarse-fine interactions (such as refluxing and coarse-fine interpolation) to exactly those described in [30].

Dragojlovic et al. [12] present a two-dimensional algorithm for viscous, conducting compressible flow with embedded boundaries. They use a split hyperbolic scheme, explicit updates of the viscous state and the (formally inconsistent) extended state algorithm developed in [24]. Our algorithm uses an unsplit scheme (as seen in [8; 25; 1]) and works in two and three dimensions. Ghias et al. [13] present an immersed boundary method to solve the same set of equations for subsonic applications. Hartmann et al. [15] present a cut-cell method that uses a form of cell merging to achieve small-cell stability. Berger et al. [5] survey a wide variety of these algorithmic permutations. We use redistribution (first presented by Chern et al. [7]) for small-cell stability. We use the (formally consistent) approach in [9] to construct extended states. To evaluate viscous fluxes at the embedded boundary we use the ray-casting algorithm developed in [18] for Poisson's equation. Also, for increased stability, we treat the viscous stress and conductivity terms implicitly.

This algorithm is suitable for use in applications where compressibility is important and the geometries are complex. Our target application is flow inside of capillary tubes in laser wakefield particle accelerators. In these accelerators, the pressure and temperature is driven very high along the axis of a capillary tube. The resulting flow produces a low density core through which lasers are shot. The capillary is connected to fill tubes which are used to fill the capillary with gas [27; 19; 20; 29]. We present a simplified version of this problem as our example to demonstrate robustness while acknowledging that other physics in these problems (such as ionization and magnetization) are very important. We drive a capillary tube with a large pressure pulse to demonstrate the stability of the algorithm under extreme conditions. The geometric configuration is derived from the experimental set-up described in [29].

There are of course many regimes for which the compressible Navier Stokes equations are relevant. The regime of interest for this algorithm has substantial compressibility effects (including shocks) as well as substantial viscous effects. We are also interested in time-accurate (as opposed to steady state calculations). For algorithm validation, we run several examples which demonstrate the efficacy of the algorithm in this regime.

First we present convergence tests demonstrating second-order solution error accuracy in two and three dimensions. For these tests, we use a smooth, subsonic ($M = 0.5$) flow inside a sphere. This demonstrates that, even with compressibility

effects, we get the expected convergence rate for smooth problems.

Next, for more quantitative validations, we present a boundary layer calculation and a viscous shock reflection calculation. Charest et al. [6] present a low-Mach-number algorithm for steady state calculations. They present a boundary layer calculation that reproduces the behavior of the similarity solution which emerges from analysis (a Blasius boundary layer profile). We present a similar run which also reproduces Blasius behavior. This demonstrates that the algorithm has correct boundary layer behavior.

Glaz et al. [14] present a comparison between inviscid calculations of shock reflections and experimental results. They show a case where viscous effects cause substantial changes in the reflection pattern. We present both viscous and inviscid calculations of the same problem and show good agreement with their results. This demonstrates, that even in this very complex, time-dependent flow, we compare well with experiment.

2. Notation

Cartesian grids with embedded boundaries are useful to describe finite-volume representations of solutions to partial differential equations in the presence of irregular boundaries. In Figure 1, the gray area represents the region excluded from the solution domain. The underlying description of space is given by rectangular control volumes on a Cartesian mesh $\Upsilon_i = [(i - \frac{1}{2}\mathbf{v})h, (i + \frac{1}{2}\mathbf{v})h]$, $\mathbf{i} \in \mathbb{Z}^D$, where D is the dimensionality of the problem, h is the mesh spacing, and \mathbf{v} is the vector whose entries are all one. Given an irregular domain Ω , we obtain control volumes $V_i = \Upsilon_i \cap \Omega$ and faces $A_{i \pm e^d/2}$ which are the intersection of the boundary of ∂V_i with the coordinate planes $\{\mathbf{x} : x_d = (i_d \pm \frac{1}{2})h\}$. We also define A_i^B to be the intersection of the boundary of the irregular domain with the Cartesian control volume: $A_i^B = \partial\Omega \cap \Upsilon_i$. For ease of exposition, we will assume here that there is only one control volume per Cartesian cell. The algorithm described here has been generalized to allow for boundaries whose width is less than the mesh spacing.

To construct finite-volume methods using this description, we will need several quantities derived from these geometric objects.

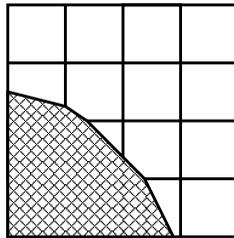


Figure 1. Illustration of cut cells. The shaded area is outside the solution domain.

- Volume fractions κ and area fraction α :

$$\kappa_i = \frac{|V_i|}{h^D}, \quad \alpha_{i+e_s/2} = \frac{|A_{i+e_s^d/2}|}{h^{D-1}}, \quad \alpha_i^B = \frac{|A_i^B|}{h^{D-1}}.$$

- The centroids of the faces and of A_i^B ; and \mathbf{n} , the average of outward normal of $\partial\Omega$ over A_i^B .

$$\mathbf{x}_{i+e^d/2} = \frac{1}{|A_{i+\frac{1}{2}e^d}|} \int_{A_{i+e^d/2}} \mathbf{x} dA - (\mathbf{i} + \mathbf{e}^d/2)h,$$

$$\mathbf{x}_i^B = \frac{1}{|A_i^B|} \int_{A_i^B} \mathbf{x} dA - \mathbf{i}h, \quad \mathbf{n}_i = \frac{1}{|A_i^B|} \int_{A_i^B} \mathbf{n} dA.$$

Here D is the dimension of space and $1 \leq d \leq D$. We assume we can compute all derived quantities to $O(h^2)$. With just these geometric descriptors, we can define a conservative discretization of the divergence operator. Let $\vec{F} = (F^1 \dots F^D)$ be a function of \mathbf{x} , then

$$\nabla \cdot \vec{F} \approx \frac{1}{|V_i|} \int_{V_i} \vec{F} dV = \frac{1}{|V_i|} \int_{\partial V_i} \vec{F} \cdot \mathbf{n} dA.$$

We discretize the divergence of the flux as

$$\kappa D(F)_i = \frac{1}{h} \left(\sum_{d=1}^D \sum_{\pm=+,-} \pm \alpha_{i \pm e^d/2} F^d(\mathbf{x}_{i \pm e^d/2}) + \alpha_i^B \mathbf{n}_i \cdot \vec{F}(\mathbf{x}_i^B) \right), \quad (1)$$

where (1) is obtained by replacing the normal components of the vector field \vec{F} with the values at the centroids. This converges to the exact divergence by the relation $D(F)_i = \nabla \cdot F + O(h/\kappa_i)$ in cells which intersect the embedded boundary and converges to $O(h^2)$ away from the boundary. The elliptic operators in this calculation all take the form

$$L(\phi) = a(\mathbf{x})\phi + D(F(\phi)).$$

We refer to a in this context as the identity coefficient.

3. System of equations

We are solving the compressible Navier–Stokes equations, given here in conservation form with hyperbolic terms to the left and elliptic terms to the right.

$$\begin{aligned}
\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho \mathbf{u}) &= 0, \\
\frac{\partial (\rho \mathbf{u})}{\partial t} + \nabla \cdot (\rho \mathbf{u} \mathbf{u} + p \mathbf{I}) &= \nabla \cdot \boldsymbol{\sigma}, \\
\frac{\partial (\rho E)}{\partial t} + \nabla \cdot (\rho \mathbf{u} E + \mathbf{u} p) &= \nabla \cdot (\boldsymbol{\sigma} \mathbf{u}) + \nabla \cdot (\boldsymbol{\xi}(\nabla T)).
\end{aligned} \tag{2}$$

In these equations, ρ is the mass density, \mathbf{u} is the velocity, $\boldsymbol{\xi}$ is the thermal conductivity, p is the pressure, and T is the temperature. The shear stress tensor $\boldsymbol{\sigma}$ is given by

$$\boldsymbol{\sigma} = \mu(\nabla \mathbf{u} + \nabla \mathbf{u}^T) + \lambda(\nabla \cdot \mathbf{u})\mathbf{I},$$

where μ and λ are the viscosity coefficients (typically $\lambda = -\frac{2}{3}\mu$). The total energy is given by $E = e + \frac{1}{2}|\mathbf{u}|^2$; the internal energy is given by $e = C_v T$ (where C_v is the specific heat at constant volume). The fluid is assumed to be an ideal gas ($p = C_v(\gamma - 1)\rho T$).

4. Algorithm description

We define $U = (\rho, \rho \mathbf{u}, \rho E)$ and we define $L^H(U) = \nabla \cdot F$, the divergence of the hyperbolic flux. The flux is given by

$$F = \begin{pmatrix} \rho \mathbf{u} \\ \rho \mathbf{u} \mathbf{u} + p \mathbf{I} \\ \rho \mathbf{u} E + \mathbf{u} p \end{pmatrix}.$$

The divergence and the fluxes are computed in the same way as in [9]. To summarize, a Taylor series extrapolation is done to produce second-order (in both space and time) approximations to the fluxes at the centroids of the faces. A conservative approximation to the divergence ($D^c(F)$) is computed using (1). Ideally, we would use $D^c(F)$ for our hyperbolic divergence. The difficulty with this approach is that the CFL (Courant–Friedrichs–Lewy) stability constraint on the time step for an algorithm using the conservative divergence for an explicit update is at best

$$\Delta t = O\left(\frac{h}{v_i^{\max}}(\kappa_i)^{1/D}\right),$$

where v_i^{\max} is the magnitude of the maximum wave speed for the i -th control volume. This is the well-known small-cell problem for embedded boundary methods. Instead, we compute a stable, nonconservative approximation to the divergence ($D^{nc}(F)$) using an extended state where necessary and ignoring the embedded boundary. This extended state is extrapolated from the interior. The effective divergence is

$$L^H(U) = \kappa D^c(F) + (1 - \kappa)D^{nc}(F).$$

The mass difference (δM) between using L^H and using only the conservative divergence $D^c(F)$ is given by

$$\delta M = \kappa(1 - \kappa)(D^c(F) - D^{nc}(F)).$$

This mass difference is redistributed to neighboring cells. The redistribution algorithm is described in [7]. This hybrid formulation preserves conservation and allows this algorithm to be stable using a time step constraint based on full cells. We compute our time step as follows:

$$\Delta t = \frac{C_F h}{W^{\max}}, \quad (3)$$

where W^{\max} is the maximum wave speed in the problem and C_F is the Courant number ($0 < C_F < 1$).

Define L^v to be the elliptic terms in the system of equations

$$L^v(U)_i = \begin{pmatrix} 0 \\ L^m(\mathbf{u})_i \\ L^k(T)_i + L^d(\mathbf{u})_i \end{pmatrix}.$$

The term L^m is the discretization of the viscous stress term ($L^m \nabla \cdot \sigma$) and is described in Section 5.3. The term $L^k \approx \nabla \cdot \xi \nabla T$ is a discretization of the heat conduction term and is described in Section 5.2. The term $L^d \approx \nabla \cdot (\sigma \mathbf{u})$ is the viscous heating term and is described in Section 5.1.

4.1. Outline. We begin with the state at time $U^n = U(n\Delta t)$, we advance the solution as follows.

1. Compute U^* , the solution advanced explicitly using only hyperbolic terms.

$$U_i^* = U_i^n - \Delta t L_i^H(U^n).$$

This produces the final value of density ($\rho^{n+1} = \rho^*$). From U^* , we compute \mathbf{u}^* and T^* , the intermediate values of velocity and temperature (which exclude the effects of conduction and viscosity).

2. Compute L_0 -stable approximations to the momentum diffusion $L^m(U) = \nabla \cdot \sigma$ by advancing the diffusion equation

$$\rho \frac{\partial \mathbf{u}}{\partial t} = L^m(\mathbf{u})$$

using the method described in Section 5:

$$\mathbf{u}^{n+1} = G_{L^m}(\rho^{n+1})\mathbf{u}^*.$$

The symbol G is defined in (6). The stable approximation to $L^m(\mathbf{u})$ is calculated as

$$(L^m(\mathbf{u}))^{n+\frac{1}{2}} = \rho^{n+1} \frac{\mathbf{u}^{n+1} - \mathbf{u}^*}{\Delta t},$$

giving us the final value of momentum:

$$(\rho \mathbf{u})^{n+1} = (\rho \mathbf{u})^* + \Delta t (L^m(\mathbf{u}))^{n+\frac{1}{2}}.$$

The operator L^m is described in Section 5.3.

3. Using the value of \mathbf{u} calculated above, calculate the viscous dissipation of energy ($L^d \approx \nabla \cdot (\sigma \mathbf{u})$) as described in Section 4.2. We then update the energy with the term

$$(\rho E)^{**} = (\rho E)^* + \Delta t L^d(\mathbf{u}^{n+1}).$$

From E^{**} , we compute the intermediate value of temperature T^{**} .

4. Compute L_0 -stable approximations to the conduction term

$$L^k(T) = \nabla \cdot \xi \nabla T$$

by advancing the diffusion equation

$$\rho C_v \frac{\partial T}{\partial t} = L^k(T)$$

using the method described in Section 5:

$$T^{n+1} = G_{L^m}(\rho^{n+1} C_v) T^{**},$$

where G is described in Equation (6). The stable approximation to $L^k(T)$ is computed by

$$(L^k(T))^{n+\frac{1}{2}} = \rho^{n+1} C_v \frac{T^{n+1} - T^{**}}{\Delta t},$$

giving us the final value of energy:

$$(\rho E)^{n+1} = (\rho E)^{**} + \Delta t (L^k(T))^{n+\frac{1}{2}}.$$

The operator $L^k(T)$ is described in Section 5.2.

4.2. Viscous dissipation calculation. To avoid small-cell instabilities, we split up the $L^d(U)$ into conservative and nonconservative approximations much as we did with L^H . The conservative approximation to $L^{d,c} = \nabla \cdot (\sigma \mathbf{u})$ is described in Section 5.1. The nonconservative form of the operator is given by the volume-weighted average of the neighbor's conservative operator evaluations. Define $N(i)$ to be the set of cells reachable from i by a unit monotone path. The nonconservative approximation of L^d is

$$L^{d,nc}(\mathbf{u})_i = \frac{\sum_{j \in N(i)} (\kappa L^{d,c}(\mathbf{u}))_j}{\sum_{j \in N(i)} \kappa_j}.$$

We use a linear combination of conservative and nonconservative versions of the divergence to advance the solution:

$$L^d(U) = \kappa L^{d,c}(\mathbf{u}) + (1 - \kappa)(L^{d,nc}(\mathbf{u})).$$

To preserve conservation, we compute the energy difference between this version and the conservative version:

$$\delta E = \Delta t \kappa (1 - \kappa) (L^{d,c} - L^{d,nc}).$$

We push this energy correction δE into the solution implicitly. First we set a right hand side $R = 0$ and redistribute δE into the cells of R that can be reached by a unit monotone path (as described in [7]). We then solve for a temperature difference that can account for this energy using the conduction operator

$$(\rho^{n+1} C_v I - \Delta t L^k) \delta T = \Delta t R.$$

This change in temperature is interpreted as an increment to the energy as follows:

$$(\delta E)^{**} = \rho^* C_v \delta T.$$

We add $(\delta E)^{**}$ into E^{**} .

5. Stable parabolic discretizations

Twizell et al. [31] present a second-order L_0 -stable algorithm to advance the constant coefficient heat equation. Given the equation

$$\frac{\partial \phi}{\partial t} = \nu L \phi, \tag{4}$$

their time advance takes the form

$$\phi^{n+1} = (I - \mu_1 L)^{-1} (I - \mu_2 L)^{-1} (I + \mu_3 L) \phi^n, \tag{5}$$

where μ_1, μ_2, μ_3 are constants. In the present algorithm we have two parabolic equations of the form

$$a \frac{\partial \phi}{\partial t} = L(\phi),$$

where $a = a(\mathbf{x}) > 0$ is the identity coefficient. Define the operator $M(\phi) = L(\phi)/a$. In the case of our viscous operator (Section 5.3) $M^m = L^m(\mathbf{u})/\rho$ and the case of conduction (Section 5.2), $M^k(T) = L^k(T)/(\rho C_v)$. In both cases, the denominators are positive and restricted away from zero. In each case, a naive interpretation of (5) yields

$$\phi^{n+1} = (I - \mu_1 M)^{-1} (I - \mu_2 M)^{-1} (I + \mu_3 M) \phi^n.$$

This is problematic in the presence of small cells because this would involve dividing by the volume fraction to evaluate M (see (1)) and volume fractions here can be arbitrarily close to zero. Using the matrix identity $(AB)^{-1} = B^{-1}A^{-1}$, we refactor the preceding equation, obtaining

$$\phi^{n+1} = G_L(a)\phi = (\kappa a I - \mu_1 \kappa L)^{-1}(\kappa a)(\kappa a I - \mu_2 \kappa L)^{-1}(\kappa a I + \mu_3 \kappa L)\phi^n. \quad (6)$$

This is the implicit advance we use for stable discretizations of $L^m(\mathbf{u})$ and $L^k(T)$.

5.1. Viscous heating operator. The viscous heating operator flux is an approximation to the shear stress dotted with the velocity ($F^h = \sigma \cdot \mathbf{u}$):

$$F^h = (\mu(\nabla \mathbf{u} + \nabla \mathbf{u}^T) + \lambda I \nabla \cdot \mathbf{u}) \cdot \mathbf{u}. \quad (7)$$

We compute the shear stress as described in Section 5.3. To get face-centered velocities, we average from neighboring cells:

$$\mathbf{u}_{i+e^d/2} = \frac{1}{2}(\mathbf{u}_{i+e^d} + \mathbf{u}_i).$$

At embedded boundaries and domain boundaries we set this flux to zero because the no slip condition requires that $\mathbf{u}|_{\partial\Omega} = 0$. We then can find the conservative discretization of the operator $L^{d,c}$ as given by (1).

5.2. Conductivity operator. Our operator for heat conduction

$$L^k(T) = \nabla \cdot (\xi \nabla T)$$

is an extension to variable coefficients of the operator described by Schwartz et al. [28]. The flux at face centers for the discretization in (1) is given by

$$F_{i+e^d/2}^T = \xi_{i+e^d/2} \frac{T_{i+e^d} - T_i}{\Delta x}.$$

Since we are representing thermally insulated embedded boundaries, $F_B^T = 0$. Given these fluxes, discretization of the operator is given by (1).

5.3. Viscous stress operator. For viscous diffusion, we first calculate the cell-centered gradient of the solution using centered differences:

$$\frac{\partial u^{d1}}{\partial x_{d2}} = \frac{u_{i+e^{d2}}^{d1} - u_{i-e^{d2}}^{d1}}{2\Delta x}.$$

The face centered gradient uses this gradient for tangential gradients and differences normal gradients directly:

$$(\nabla \mathbf{u})_{i+e^{d/2}}^{d'} = \begin{cases} (1/h)(\mathbf{u}_{i+e^d} - \mathbf{u}_i) & \text{if } d = d', \\ \frac{1}{2}((\nabla \mathbf{u})_{i+e^d}^{d'} + (\nabla \mathbf{u})_i^{d'}) & \text{if } d \neq d', \end{cases}$$

where

$$(\nabla \mathbf{u})_i^d = \frac{1}{2h} (\mathbf{u}_{i+e^d} - \mathbf{u}_{i-e^d}).$$

We then construct the flux at the face using the appropriate gradients:

$$F^v = \mu(\nabla \mathbf{u} + \nabla \mathbf{u}^T) + \lambda I \nabla \cdot \mathbf{u}. \quad (8)$$

At the embedded boundary, we have a physical boundary condition that $\mathbf{u} = 0$. Define a local coordinate system rotated to align with the normal to the embedded boundary \hat{n} and the tangent plane (\hat{t}^1, \hat{t}^2) . The Jacobian J of this rotational transformation is given by

$$J = \begin{pmatrix} \hat{n} \\ \hat{t}^1 \\ \hat{t}^2 \end{pmatrix}.$$

The transformation between a vector in Cartesian space (v) and a vector in rotated space (v^R) is given by

$$v^R = Jv.$$

We start by treating each component of the velocity as a scalar ϕ . To create our boundary flux, we use the Johansen extrapolation [18] to compute the normal gradient of ϕ , $(\nabla \phi^{R,n})$. We set the tangential components of the gradient of ϕ to zero (a consequence of the no-slip condition). So, in the rotated frame $(\nabla \phi)^R = (\nabla \phi^{R,n}, 0, 0)$. We then compute the Cartesian gradient of ϕ :

$$\nabla \phi = J^{-1}(\nabla \phi)^R.$$

We then construct the boundary flux using (8). Given these fluxes, discretization of the operator is given by (1).

5.4. Performance implications of implicit parabolic discretization. The time step constraint for the present algorithm is given by (3). Since we are advancing our elliptic terms implicitly, this adds no additional time step constraint. Suppose we were to advance (4) explicitly:

$$\phi^{n+1} = \phi^n + \nu \Delta t_{\text{exp}} L(\phi^n). \quad (9)$$

In the absence of cut cells, the stability constraint on this method is

$$\Delta t_{\text{exp}}^{\text{noeb}} < \frac{\Delta x^2}{2D\nu}.$$

where D is the dimensionality of the problem. For the conductivity operator at constant density with constant coefficients, this relationship is exact with $\nu = \xi/(\rho C_v)$.

To illustrate the performance tradeoff in the design decision to use the implicit discretization, we compare the number of operator evaluations required to advance the solution. Define N_{exp} to be the number of operator evaluations needed to advance the solution to a fixed time t_f using (9):

$$N_{\text{exp}} = \frac{t_f}{\Delta t_{\text{exp}}}.$$

Define N_{imp} to be the number of operator evaluations needed to advance the solution to a fixed time t_f using (6). We solve our elliptic equations using multigrid and we measure how many times the operators are applied. This puts the implicit method in the worst light possible because coarse and fine applications of the operator (through multigrid) are counted the same.

The problem in Section 7 is the target application for this algorithm. When we run this problem with 4 levels of refinement for a final time of $0.7 \mu\text{s}$ (which accounts for 3 steps at the coarsest level and 48 total steps at all levels), the conductivity operator is called $N_{\text{imp}} = 600$ times. For these parameters, the time step restriction for the explicit advance is $\Delta t_{\text{exp}}^{\text{noeb}} = 2.63 \cdot 10^{-10}$, so an explicit advance would call the operator $N_{\text{exp}} = 2665$ times. For problems with less resolution or lower viscosity, this performance tradeoff can easily flip and make the explicit method more efficient. In the shock-boundary layer calculation presented in Section 9, for example, $\Delta t_{\text{exp}}^{\text{noeb}} > \Delta t$, which means that the explicit parabolic advance for this case presents no addition time step constraint in the absence of embedded boundaries.

With embedded boundaries, however, the time step constraint for the explicit advance (Equation (9)) is far more severe. If κ_{min} is the smallest volume fraction in the domain, the true time step constraint for the explicit advance is given by

$$\Delta t_{\text{exp}} < \frac{\Delta x^2 (\kappa_{\text{min}})^{2/D}}{2D\nu}.$$

In this context, let us reconsider the shock-boundary layer calculation for a final time of $0.57 \mu\text{s}$ (and all other parameters described in Section 9), which is one time step at the coarsest level and 97 time steps at all levels. The smallest volume fraction at the finest level of this calculation is $\kappa_{\text{min}} = 3.83 \cdot 10^{-7}$, which means that $\Delta t_{\text{exp}} = 4.26 \cdot 10^{-15}$ and the number of operator evaluations required for stability is given by $N_{\text{exp}} = 1.34 \cdot 10^8$. The number of operator evaluations we count for our implicit algorithm is $N_{\text{imp}} = 37536$. Clearly, in the presence of small cells, the implicit advance is the more efficient algorithm to advance our elliptic terms.

6. Convergence tests

To test the convergence rate of the algorithm we start with an initial condition of flow within a sphere (or a circle in two dimensions). All tests are done using

Richardson extrapolation which means that an average of a finer solution is used as an exact solution. Define A^{h-2h} to be a volume-weighted averaging operator. Given S_f to the set of fine volumes which cover a coarse volume i ,

$$A^{h-2h}(f)_i = \frac{\sum_{i_f \in S_f} \kappa_{i_f} f_{i_f}}{\sum_{i_f \in S_f} \kappa_{i_f}}.$$

U_h is defined to be our solution on a grid with resolution h . For an exact solution U^e , we use $U_{2h}^e = A^{h-2h}(U_h)$ and the error is given by

$$\epsilon^h = U^h(t) - U^e(t). \quad (10)$$

The order of convergence ϖ is estimated by

$$\varpi = \frac{\log(\|\epsilon^{2h}\|/\|\epsilon^h\|)}{\log 2}. \quad (11)$$

We compute the convergence rates using compute using L_∞ , L_1 , and L_2 norms (all these norms are defined in [9]). The geometry of the test is a sphere with radius in the center of a domain of length L . The initial condition of the tests is given by an axisymmetric Gaussian disturbance $f(r) = \exp(-30(r/r_0 - 0.5)^2)$. The maximum Mach number is set to $M = 0.5$. Define (x, y, z) to be Cartesian coordinates in a coordinate system whose origin is the sphere center. Define the distance $r = (x^2 + y^2 + z^2)^{1/2}$. The velocity is given by $\mathbf{u} = (-Mf(r)y/r_0, Mf(r)x/r_0)$ in two dimensions and $\mathbf{u} = (Mf(r)(z - y)/r_0, Mf(r)(x - z)/r_0, Mf(r)(y - x)/r_0)$ in three dimensions. Define v to be the magnitude of the velocity vector. The density and pressure are given by $\rho = \gamma(1 + v^2/r)$, $p = (1 + v^2/r)$. See Table 1 for other solution parameters.

Solution error is a measure of the convergence rate of the solution run to a fixed time. All refinements were advanced to a fixed time $t_f = 32 \mu\text{s}$. The finest solution was advanced 64 time steps with $\Delta t = 0.5 \mu\text{s}$. Each successively coarser solution was advanced half as many steps with twice as big a time step. This results in a Courant number (C_F , see (3)) of approximately 0.1 for full cells. The results of the solution error test are given in Tables 2 and 3. We demonstrate second-order accuracy in all norms.

$$\begin{aligned} \mu &= 2.1 \cdot 10^{-5} \text{ kg/(m s)} & L &= 1.0 \cdot 10^{-2} \text{ m} \\ \lambda &= -1.4 \cdot 10^{-5} \text{ kg/(m s)} & r_0 &= 4.5 \cdot 10^{-3} \text{ m} \\ C_v &= 3.00 \cdot 10^2 \text{ J/(kg K)} & \gamma &= \frac{7}{5} \\ \xi &= 1.7 \cdot 10^{-2} \text{ W/(m K)} \end{aligned}$$

Table 1. Initial condition set-up for the convergence tests. See text for variable definitions.

norm	variable	$e_{4h \rightarrow 2h}$	ϖ	$e_{2h \rightarrow h}$
L_∞	ρ	$1.103 \cdot 10^{-2}$	1.980	$2.796 \cdot 10^{-3}$
	$(\rho \mathbf{u})_x, (\rho \mathbf{u})_y$	$2.740 \cdot 10^{-3}$	1.822	$7.748 \cdot 10^{-4}$
	ρE	$2.006 \cdot 10^{-2}$	1.978	$5.092 \cdot 10^{-3}$
L_1	ρ	$2.519 \cdot 10^{-3}$	1.982	$6.377 \cdot 10^{-4}$
	$(\rho \mathbf{u})_x, (\rho \mathbf{u})_y$	$3.645 \cdot 10^{-4}$	1.822	$1.031 \cdot 10^{-4}$
	ρE	$4.560 \cdot 10^{-3}$	1.978	$1.158 \cdot 10^{-3}$
L_2	ρ	$3.900 \cdot 10^{-3}$	1.978	$9.903 \cdot 10^{-4}$
	$(\rho \mathbf{u})_x, (\rho \mathbf{u})_y$	$6.795 \cdot 10^{-4}$	1.824	$1.920 \cdot 10^{-4}$
	ρE	$7.066 \cdot 10^{-3}$	1.973	$1.801 \cdot 10^{-3}$

Table 2. Solution error convergence rates in two dimensions using the L_∞ -, L_1 - and L_2 -norms for $h = \frac{1}{1024}$ cm.

norm	variable	$e_{4h \rightarrow 2h}$	ϖ	$e_{2h \rightarrow h}$
L_∞	(ρ)	$3.536 \cdot 10^{-2}$	1.979	$8.968 \cdot 10^{-3}$
	$(\rho \mathbf{u})_x, (\rho \mathbf{u})_y, (\rho \mathbf{u})_z$	$7.406 \cdot 10^{-3}$	1.814	$2.107 \cdot 10^{-3}$
	(ρE)	$6.887 \cdot 10^{-2}$	1.978	$1.748 \cdot 10^{-2}$
L_1	(ρ)	$4.167 \cdot 10^{-3}$	1.986	$1.053 \cdot 10^{-3}$
	$(\rho \mathbf{u})_x, (\rho \mathbf{u})_y, (\rho \mathbf{u})_z$	$4.503 \cdot 10^{-4}$	1.805	$1.289 \cdot 10^{-4}$
	(ρE)	$7.767 \cdot 10^{-3}$	1.983	$1.965 \cdot 10^{-3}$
L_2	(ρ)	$7.931 \cdot 10^{-3}$	1.982	$2.007 \cdot 10^{-3}$
	$(\rho \mathbf{u})_x, (\rho \mathbf{u})_y, (\rho \mathbf{u})_z$	$1.060 \cdot 10^{-3}$	1.810	$3.024 \cdot 10^{-4}$
	(ρE)	$1.495 \cdot 10^{-2}$	1.980	$3.790 \cdot 10^{-3}$

Table 3. Solution error convergence rates in three dimensions using the L_∞ -, L_1 - and L_2 -norms for $h = \frac{1}{1024}$ cm.

7. Capillary tube simulation

Our target application is the flow inside of capillary tubes in laser wakefield particle accelerators. We present a simplified version of this problem as our robustness calculation while acknowledging that other physics in these problems (such as ionization and magnetization) are very important. Refer to Figure 2. The main tube (C) and the fill tubes (A) are filled with gas. The experimentalists drive the core pressure p_{core} along the axis of the tube to a high value using electrical charge, leaving the density constant. The resulting flow causes the core to expand and create a low density, high energy core. In the experiment, the laser (B) is shot through

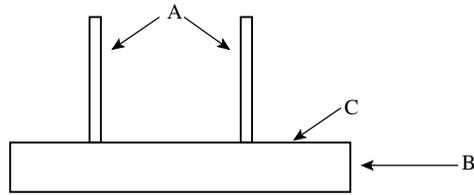


Figure 2. Illustration of wakefield accelerator. The main tube (C) and fill tubes (A) are filled with gas. The pressure and temperature are initialized to high values up along the axis of the tube. The resulting flow causes this region to expand and create a low density, high-energy core. A laser (B) is shot through this core.

this low density core. Ideally this core should be cylindrical and have a relatively flat density profile. There is some concern in the community, however, that the fill tubes can alter the core shape before the laser is shot.

We present both two- and three-dimensional runs that are meant to approximate to this problem. For a computational geometry we intersect a 200 micron diameter main tube with a perpendicular fill tube 50 microns in diameter. Figure 3 shows the

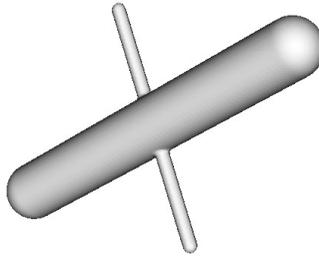


Figure 3. Geometric configuration of the three-dimensional example. The core tube's diameter is 200 microns; the filler tube's diameter is 50 microns. The core tube's length is 1.2 mm; the filler tube's length is 0.85 mm.

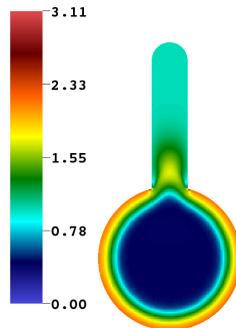


Figure 4. Two-dimensional plot of $\log \rho$ after $35 \mu s$. The base grid is 256^2 and there are 2 levels of refinement, all by a factor of 2. This means the effective grid resolution is 1024^2 . Though the density profile in the core is relatively flat, the core shape is no longer circular.

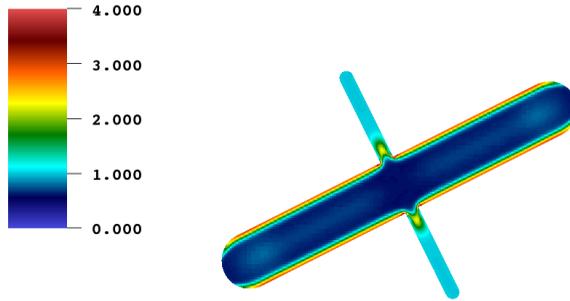


Figure 5. Two-dimensional plot of $\log \rho$ after $35 \mu\text{s}$. The base grid is 128×64 and there are 3 levels of refinement, all by a factor of 2. This means the effective grid resolution is 1024×512 . Though the density profile in the core is relatively flat, the filler tube has distorted the profile.

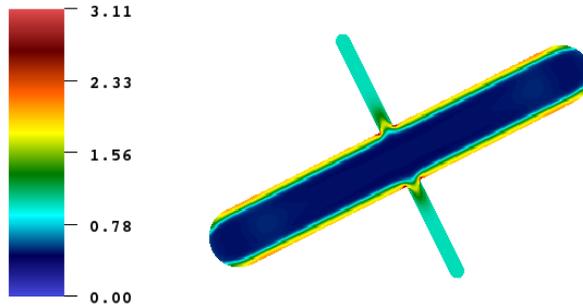


Figure 6. Axial slice through three-dimensional run plot of $\log \rho$ after $50 \mu\text{s}$. This is a one-level calculation with resolution $256 \times 128 \times 128$. Though the density profile in the core is relatively flat, the filler tube has distorted the profile.

geometric configuration. Both are filled with argon at 1 Pa, 1 kg/m^3 . We initialize the core pressure to be $p_{\text{core}} = 20 \text{ Pa}$, leave the density constant and initialize the velocity everywhere to zero. The core diameter is 100 microns. Figure 4 (on the previous page) shows a two-dimensional run of the plane normal to the central tube cutting through a filler tube. We plot the logarithm of density after $35 \mu\text{s}$. Though the density profile in the core is relatively flat, the core shape is no longer circular. Figure 5 shows a two-dimensional run of the plane along the central tube cutting through a filler tube. We plot the logarithm of density after $35 \mu\text{s}$. Though the density profile in the core is relatively flat, the core shape is once again distorted by the presence of the filler tube. Figure 6 shows an axial slice through a three-dimensional run after $50 \mu\text{s}$ and shows a similar result. To be clear, since we do not include any source terms for the effects of ionization or magnetization, this is greatly simplified approximation. We have, however, managed to show that purely hydrodynamic effects can distort the shape of the low density core.

$$\begin{aligned}
 p_{\text{outside}} &= 1.0 \text{ Pa} & \mu &= 2.1 \cdot 10^{-5} \text{ kg/(m s)} \\
 p_{\text{core}} &= 20.0 \text{ Pa} & \lambda &= -1.4 \cdot 10^{-5} \text{ kg/(m s)} \\
 \rho_{\text{outside}} &= 1.0 \text{ kg/m}^3 & C_v &= 3.00 \cdot 10^2 \text{ J/(kg K)} \\
 \rho_{\text{core}} &= 1.0 \text{ kg/m}^3 & \xi &= 1.7 \cdot 10^{-2} \text{ W/(m K)} \\
 & & \gamma &= \frac{5}{3}
 \end{aligned}$$

Table 4. Initial condition set-up for capillary tube problem. The initial velocity is zero.

8. Boundary layer similarity solution

Given a semiinfinite flat plate in a flow field at zero incidence to the flow, the velocity profile over the plate can be calculated using a similarity solution in the absence of thermal and compressibility effects. Define x to be the distance along the plate and y to be the distance from the surface of the plate and ν to be the kinematic viscosity and $U = Mc$ is the incident velocity (refer to Figure 7). The similarity variable η is given by

$$\eta = y \sqrt{\frac{U}{\nu x}}. \quad (12)$$

This reduces the equations to a nonlinear ordinary differential equation, the solution of which is the familiar Blasius boundary layer. See Schlichting [26] or White [32] for a full exposition of this derivation. Charest et al. [6] present a low-Mach-number algorithm for steady state calculations. In this calculation, they present a boundary layer calculation that reproduces the behavior Blasius layer. Berger et al. [5] present a wide variety of these calculations.

We cut a rectangular grid with a wedge of angle θ . Refer to Figure 8 and Table 5 for the initial and boundary conditions. The density and temperature are set to constants $\rho = \rho_0$, $T = T_0 = P_0/(RT_0)$. The velocity is set to $(U \cos \theta, U \sin \theta)$ everywhere. The velocity boundary conditions are inflow-outflow left to right (the

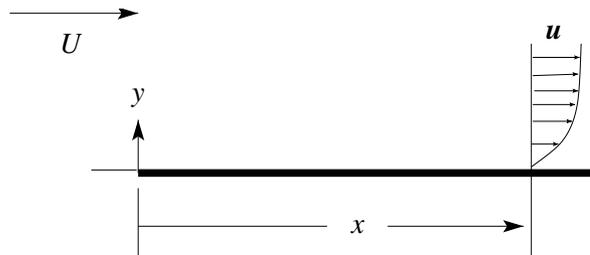


Figure 7. Formulation of semiinfinite flat plate boundary layer problem. U is the (constant) inflow velocity, x is the distance along the plate and y is the distance above the plate.

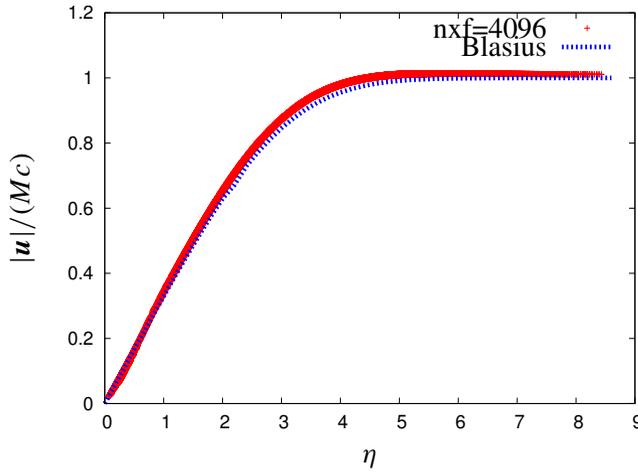


Figure 9. Results of boundary layer calculation as compared with the Blasius solution. This is a scatter plot of normalized velocity versus the similarity variable η at two different resolutions. The magnitude of solution velocity is $|\mathbf{u}|$, c is the sound speed and the similarity variable η is defined in (12). The Blasius solution is in blue. The solution with effective resolution of 4096×4096 is in red. We cast a ray from every point along the boundary where the local Reynolds number is in the range $5000 < \text{Re}_x < 15000$. We plot every point along every ray. The rays are 30 points long.

9. Shock reflection

Define M to be the Mach number of a shock propagating into a gas at rest. Glaz et al. [14] present a comparison between inviscid calculations of shock reflections and experimental results. They show a case where viscous effects cause substantial changes in the reflection pattern. For $M = 7.1$ shock reflection from a 49 degree wedge, they show that the Mach stem is much shorter in the experiment than in an inviscid calculation. The reason cited for this difference is that the viscosity of argon varies strongly with temperature and the temperature behind the shock is quite high (the initial temperature behind the diaphragm is 10265 K). The viscosity is approximated to vary with the Sutherland's power law (dynamic viscosity varies with $T^{3/2}$). We use the viscosity shown in Table 6. We compute this viscosity using the highest value given in [22] and extrapolating to the initial high temperature. The specific heat and conductivity of argon are left at the room temperature values. These approximations are sufficient to illustrate the phenomenon.

Refer to Figure 10 for an illustration of the initial conditions. Table 6 has the numerical values of the inputs. Both calculations have a 128×64 base grid with seven levels of adaptive mesh refinement, all by a factor of 2. This makes the effective resolution 16384×8192 . All embedded boundary cells are refined to the finest level. This gives resolution at the boundary layer $h = 9.1$ microns.

$$\begin{array}{ll}
 \rho_0 = 1.95 \cdot 10^3 \text{ Pa} & Y_3 = 7.50 \cdot 10^{-2} \text{ m} \\
 p_1 = 7.42 \cdot 10^5 \text{ Pa} & \mu = 1.21 \cdot 10^{-3} \text{ kg/(m s)} \\
 \rho_0 = 3.29 \cdot 10^{-2} \text{ kg/m}^3 & \lambda = -8.08 \cdot 10^{-4} \text{ kg/(m s)} \\
 \rho_1 = 3.61 \cdot 10^{-1} \text{ kg/m}^3 & C_v = 3.00 \cdot 10^2 \text{ J/(kg K)} \\
 X_1 = 9.0 \cdot 10^{-2} \text{ m} & \xi = 1.7 \cdot 10^{-2} \text{ W/(m K)} \\
 X_2 = 1.0 \cdot 10^{-1} \text{ m} & \theta = 49^\circ \\
 X_3 = 1.5 \cdot 10^{-1} \text{ m} & \gamma = \frac{5}{3}
 \end{array}$$

Table 6. Initial condition set-up for shock reflection problem. The initial velocity is zero. See Figure 10 for variable definitions.

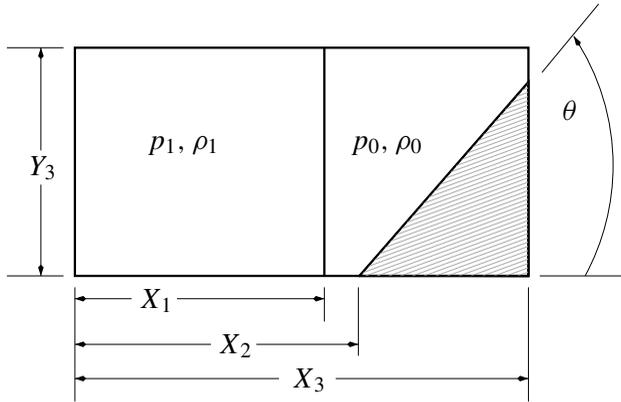


Figure 10. Shock tube set-up. The initial velocity is zero. The initial pressures and densities are tailored to make a $M = 7.1$ shock. See Table 6 for details.

Figure 11 illustrates the Mach reflection problem. Figure 12 shows the viscous and inviscid calculations at the same scale after $9.61 \mu\text{s}$. The viscous calculation shows an interesting shock-boundary layer interaction, which is magnified in Figure 14. The shock reflects off of the boundary layer, creating a separation bubble. This is followed by a compression (from the reflected shock) and boundary layer reattachment. For steady shocks interacting with laminar boundary layers, this is the classical lambda shock phenomenon. Both Schlichting [26] and Liepmann et al. [21] explain this in detail and include a wealth of experimental images. This is also observed (albeit barely) in the experiment presented in [14]. The interferogram they show has only two or three density contours in that region which makes the feature difficult to see.

Figure 12 clearly shows that the viscous boundary layer has reduced the Mach stem substantially and a density stratification on the left. Recall that the problem is configured as a shock tube. The initial conditions are zero velocity with a discontinuity in pressure and density. As the shock moves to the right, a rarefaction

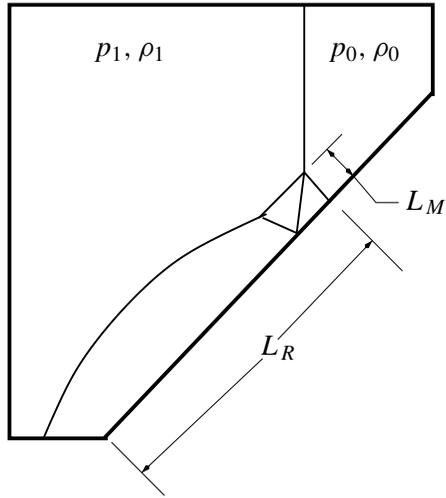


Figure 11. Shock reflection illustration. The ratio of the Mach stem length L_M to the shock distance L_R is the quantity of interest.

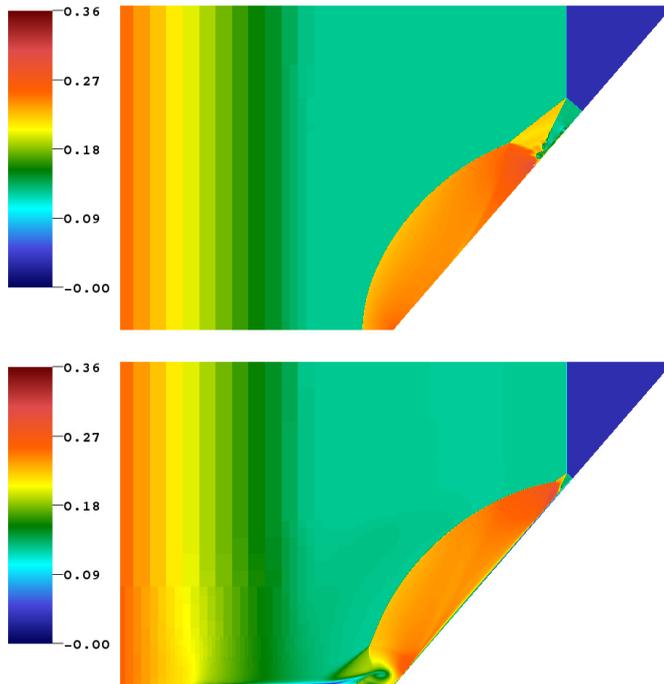


Figure 12. Mass density (kg/m^3) in the inviscid (top) and viscous (bottom) calculations of $M = 7.1$, with 49° shock reflection at $9.6 \mu\text{s}$. These calculations were run on a 128×64 base grid with seven levels of adaptive mesh refinement, all by a factor of 2. This makes the effective resolution 16384×8192 . All embedded boundary cells are refined to the finest level. This gives resolution at the boundary layer $h = 9.1$ microns.

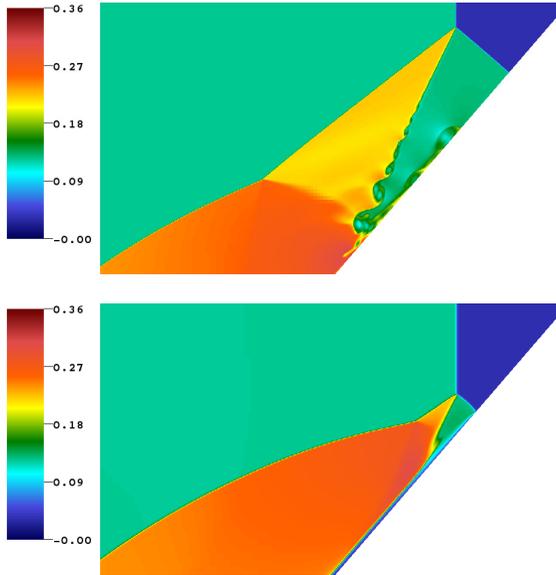


Figure 13. Mass density (kg/m^3) in the inviscid (top) and viscous (bottom) calculation of $M = 7.1$, with 49° shock reflection at $9.6 \mu\text{s}$, zoomed in to show the reflection pattern.



Figure 14. Magnification of the lambda shock-boundary layer pattern in the viscous calculation. Density (in kg/m^3) shown here.

fan moves to the left, producing this density variation. We show the two shock reflection patterns more closely in Figure 13. For a quantitative look at this reduction, we refer to the experimental and computational results in [14]. See Figure 11 for an illustration of the relevant lengths. The ratio of the Mach stem length L_M to the shock distance L_R is the quantity of interest:

$$R_m = \frac{L_M}{L_R}.$$

Glaz et al. report a value of $R_m = 0.07$ in their inviscid calculation and $R_m = 0.038$ for an experimental result (see Figure 10 in [14]). Our inviscid calculation has $R_m = 0.072$ and our viscous calculation has $R_m = 0.03$. We believe that our agreement is reasonable since not all the experimental set-up information is available (the time at which the interferogram is taken, for example, is not available). For more examples of this viscous effect, see Henderson et al. [17].

10. Conclusion

We have presented a stable, second-order method for solving the two- and three-dimensional compressible Navier–Stokes equations in the presence of complex geometries. This semiimplicit method advances parabolic terms implicitly and hyperbolic terms explicitly. This allows a time step controlled by the CFL constraint associated with the hyperbolic wave speeds. We demonstrate second-order accuracy for smooth initial conditions in smooth geometric configurations and robust behavior in the presence of strong discontinuities and geometric complexity that mimic the conditions in a plasma wakefield accelerator in the absence of magnetic or ionization effects. We also show good quantitative agreement with experimental results in a viscous shock reflection problem and a boundary layer problem.

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