Case C1.4: Vortex trasport by uniform flow

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I. Discretization, iterative method and hardware

See the appendices.

II. Case summary

This classic test case aims at characterizing the solver’s ability to preserve vorticity in an inviscid flow. The unsteady 2D Euler equations govern the simulation, which consists in a 2D vortex transported by a uniform flow across a rectangular computational domain of dimensions \((x, y) = (0, L_x) \times (0, L_y)\). The initial configuration of the vortex, centered in \((x_c, y_c)\) and superimposed onto the uniform (infinity) flow, is given by the following equations:

\[
\begin{align*}
\rho_0 &= \rho_\infty \left( \frac{T_0}{T_\infty} \right)^{1/(\gamma - 1)}, & u_0 &= U_\infty + \delta u, & v_0 &= \delta v, & T_0 &= T_\infty - \delta T \\
\delta u &= -\left( \beta U_\infty \right) \frac{y - y_c}{R} e^{-r^2/2} \\
\delta v &= \left( \beta U_\infty \right) \frac{x - x_c}{R} e^{-r^2/2} \\
\delta T &= 0.5 C_p \left( \beta U_\infty \right)^2 e^{-r^2}
\end{align*}
\]

where

\[
C_p = \frac{\gamma}{\gamma - 1} R_{gas}, \quad r = \sqrt{\left( x - x_c \right)^2 + \left( y - y_c \right)^2}/R
\]

The variable \(R\) represents the vortex characteristic radius while \(\beta\) defines its strength. Furthermore, \(\gamma = 1.4\) is the constant specific heat ratio, \(R_{gas} = 287.87 \text{ J/(kg K)}\) is the gas constant and \(U_\infty = M_\infty \sqrt{\gamma R_{gas} T_\infty}\) and \(\rho_\infty = p_\infty/(T_\infty R_{gas})\) are the velocity and density of the unperturbed flow, respectively.

The “slow vortex” configuration, for which \(M_\infty = 0.05\), \(\beta = 1/50\) and \(R = 0.005\), is computed on regular grids. Moreover, we assume \(L_x = L_y = 0.1 \, \text{m}\), \(x_c = y_c = 0.05 \, \text{m}\), \(p_\infty = 10^5 \, \text{Pa}\) and \(T_\infty = 300 \, \text{K}\). The flow is periodic in both directions, therefore no boundary closure is needed and the internal discretization is used throughout the domain. As the problem is almost linear, no artificial dissipation was necessary for this case.

The solution is advanced in time with the classical RK4 scheme up to 50 periods \((t_{final} = 50 U_\infty/L_x)\). For this case the time step is dictated by stability requirements. Reducing the time step by a factor of two did not change the results significantly. Since the vortex should be transported without distortion, the initial solution can be used to assess the accuracy of the algorithm, hence the the norm of the error, \(L_2^{\text{error}}\), can be
defined as:

\[ L^2_{\text{error}} = \left( \frac{1}{N} \sum_{i=1}^{N} \frac{\text{error}_i^2 \cdot |J_i|}{\sum_{i=1}^{N} |J_i|} \right)^{1/2}, \quad (1) \]

with

\[ \text{error}_i = \phi_i^{\text{final}} - \phi_i^{\text{initial}}, \quad i = 1, 2, ..., N, \quad (2) \]

where \( \phi \) is either the x-velocity \( u \) or the y-velocity \( v \), \( N \) is the total number of grid points, and \( |J_i| \) is the determinant of the metric Jacobian. The solution is computed on four structured uniform grids containing \( 33^2, 65^2, 129^2 \) and \( 257^2 \) vertices, respectively.

![Figure 1. \( L^2 \) norms of the velocity components \((u, v)\) as function of the grid size.](image1)

![Figure 2. \( L^2 \) norms of the velocity components \((u, v)\) as function of the work units. The quadruple precision 8th order case is not shown to avoid excessive scaling.](image2)
Figure 1 and 2 show the $L^2$ error of the two velocity components as a function of $h$ and work units. The 2\textsuperscript{nd} order scheme does not show the design accuracy, in fact the error is almost flat. This indicates that the asymptotic range has not been reached and therefore the 2\textsuperscript{nd} order scheme requires much finer grids. Note also that there is no difference between the error on the second finest and finest grid for the 8\textsuperscript{th} order scheme when running in double precision. It turns out that the reason for this behavior is finite arithmetic. When running the solver in quadruple precision, the error goes down according to design accuracy of the scheme. Therefore, a quadruple precision computation is approximately two orders of magnitude slower than the corresponding double precision computation and has therefore only be carried out for the 8\textsuperscript{th} order scheme.

### A. Background information for the SBP-SAT scheme

As is well-known, stability of a numerical scheme is a key property for a robust and accurate numerical solution. Proving stability for high-order finite-difference schemes on bounded domains is a highly non-trivial task. One successful way to obtain stability proofs is to employ so-called Summation-by-Parts (SBP) schemes with Simultaneous Approximation Terms (SAT) for imposing boundary conditions. With a simple example, we will briefly describe how stability proofs can be obtained.

Consider the scalar advection equation,

$$u_t + a u_x = 0, \quad 0 < x < 1, \quad 0 < t \leq T$$

$$a^+ u(0, t) = a^+ g(t) \quad a^- u(1, t) = a^- g(t)$$

where $a^+ = \max(a, 0)$ and $a^- = \min(a, 0)$. Furthermore, we augment the equation with initial data $u(x, 0) = f(x)$, bounded in $L^2$. To demonstrate well-posedness, we employ the energy method.

$$\|u\|^2_1 + a \int_0^1 u u_x \, dx = 0$$

$$\|u\|^2_1 \leq au^2(0, t) - au^2(1, t) \leq a^+ g(t)^2 - a^- g_r(t)^2$$

Integrating in time gives the bound

$$\|u(\cdot, T)\| \leq \|f\| + a^+ \int_0^T g(t)^2 \, dt - a^- \int_0^T g_r(t)^2 \, dt.$$  

(5)

For linear PDEs, such a bound is sufficient to prove well-posedness.

Next, we turn to the SBP-SAT semi-discretization of (3). To this end, we introduce the computational grid, $x_i = ih, \ i \in \{0, 1, 2, ..., N\}$ and $h > 0$ is the grid spacing. For the moment, we keep time continuous. With each grid point $x_i$, we associate a value $v_i(t)$, and define a grid function $v(t) = (v_0, v_1, v_2, ...)^T$. The SBP difference operator, $D$ is a matrix with the following properties: $D = P^{-1}Q$ where $P$ and $Q$ are two matrices; $P = P^T > 0$ and $Q + Q^T = B = \text{diag}(-1, 0, ..., 0, 1)$. The matrix $P$ can be used to define a weighted $l^2$ equivalent norm as $\|v\|^2 = v^T P v$. We will also need the vectors $e_0 = (1, 0, ..., 0)^T$ and $e_N = (0, ..., 0, 1)^T$.

Let $w$ denote a smooth function and define a grid function $\tilde{w} = (w(x_0), ..., w(x_N))^T$ and $\tilde{w}_x = (w_x(x_0), ..., w(x_N))^T$. It turns out that the SBP property precludes the accuracy of $D$ to be uniform in space. We have

$$D \tilde{w} = \tilde{w}_x + \tilde{T}$$

where $\tilde{T}$ is the truncation error. In general, it takes the form,

$$\tilde{T} = (O(h^s), ..., O(h^s), O(h^p), ..., O(h^p), O(h^s), ..., O(h^s)).$$

(6)

where $s < p$ and the lower accuracy is confined to a few (finite) number of points close to the boundary. SBP operators exist with various orders of accuracy, [1]. In particular, if $P$ is a diagonal matrix, there are SBP operators with $p$ even and $p \leq 8$, and $s = p/2$. If $P$ is allowed to have off-diagonal elements for a few points near the boundary $s = p - 1$ can be achieved.
Using the SBP operators, we now define a semi-discrete scheme for (3).
\[ v_t + a Dv = \sigma_l a^+ P^{-1} e_0 (v_0 - g_l(t)) + a^- \sigma_r P^{-1} e_N (v_N - g_r(t)) \]

The right-hand side are the SAT:s, which impose the boundary conditions weakly. (Originally proposed in [2].) \(\sigma_l, \sigma_r\) are two scalar parameters, to be determined by the stability analysis. Multiplying by \(2v^T P\), we obtain
\[
\|v\|_t^2 - a(v_0^2 - v_N^2) = 2\sigma_l a^+ v_0 (v_0 - g_l(t)) + 2a^- \sigma_r v_N (v_N - g_r(t))
\]

For stability, it is sufficient to obtain a bound with \(g_l, g_r = 0\). In that case, it is easy to see that we must require \(\sigma_l \leq -1/2\) and \(\sigma_r \geq 1/2\) to obtain a bounded growth of \(\|v\|\). More generally, allowing boundary data to be inhomogeneous when deriving a bound leads to strong stability. (See [3]. The benefit of proving strong stability as opposed to stability is that less regularity in the boundary data is required.) For strong stability, it can be shown that \(\sigma_l, \sigma_r\) must satisfy \(\sigma_l < -1/2\) and \(\sigma_r < 1/2\), i.e., strict inequalities. As an example, the choice \(\sigma_l = -1, \sigma_r = 1\) leads to
\[
\|v\|_t^2 - a(v_0^2 - v_N^2) = -2a^+ v_0 (v_0 - g_l(t)) + 2a^- v_N (v_N - g_r(t))
\]
or
\[
\|v\|_t^2 \leq -a^+ (v_0 - g_l)^2 + a^+ g_l(t)^2 + a^- (v_N - g_r(t)^2) - a^- g_r^2
\]

If \(v_0 = g_l, v_N = g_r\), (8) is the same as (4), but this is not the case and the additional terms add a small damping to the boundary. Upon integration of (8) in time, an estimate corresponding to (5) is obtained. We also remark that the SAT terms are accurate as they do not contribute to a truncation error in the scheme. Furthermore, semi-discrete stability guarantees stability of the fully discrete problem obtained by employing Runge-Kutta schemes in time, [4].

The above example, demonstrates the general procedure for obtaining energy estimates for an SBP-SAT scheme. Naturally, for systems of PDEs, in 3-D with stretched and curvilinear multi-block grids, and with additional parabolic terms, the algebra for proving stability becomes more involved. However, the resulting schemes are still fairly straightforward to use. For the linearized Euler and Navier-Stokes equations, semi-discrete energy estimates have been derived. (See [5–7] and references therein.) Different boundary types, including far-field, walls and grid block interfaces are included in the theory. For flows with smooth solutions, linear stability implies convergence as the grid size vanishes. (See [8].)

B. Code description

Both a general code and specialized codes for some of the test cases (used in the 1st and 2nd high order workshop, see [9]) are available. The general code is a 3D code that can handle multiblock grids and can run on (massively) parallel platforms. For load balancing reasons the blocks are split during runtime in an arbitrary number of sub-blocks with a halo treatment of the newly created interfaces, such that the results are identical to the sequential algorithm.

The specialized codes assume a single block 2D grid and do not have parallel capabilities, hence they are relatively easy to modify for testing purposes. Due to the fact that these codes can only be used for one specific test case and the fact that the general purpose code can only handle 3D problems, the efficiency of the specialized codes is quite a bit higher than the general purpose code.

The discretization schemes used are finite difference SBP-SAT schemes, see section A, of order 2 to 5. Thanks to the energy stability property of these schemes no or a significantly reduced amount of artificial dissipation is needed compared to schemes which do not posses this (or a similar) property. This leads to a higher accuracy of the numerical solutions.

For the steady test cases the set of nonlinear algebraic equations is solved using the nonlinear solver library of PETSc [10]. This library requires the Jacobian matrix of the spatial residual, which is computed via dual numbers [11] and appropriate coloring of the vertices of the grid, for which the PETSc routines are used. Initial guesses are obtained via grid sequencing, where appropriate. The solution of the linear systems needed by PETSc’s nonlinear solution algorithm is obtained by Block ILU preconditioned GMRES.
Implicit time integration schemes of the ESDIRK type [12] are available, for which the resulting nonlinear systems are solved using a slightly adapted version of the steady state algorithm explained above. However, for the unsteady test cases considered, the Euler vortex and the Taylor-Green vortex, the time steps needed for accuracy are relatively small compared to the stability limit of explicit time integration schemes and therefore the explicit schemes are better suited for these cases. The available explicit schemes are the classical 4th order Runge Kutta scheme (RK4, [13]) and TVD Runge Kutta schemes up till 3rd order [14]. As the maximum CFL number of the RK4 scheme is significantly higher than the CFL number of the TVD Runge Kutta schemes, the RK4 scheme is used for the unsteady test cases mentioned above.

For the post processing standard commercially available software, such as Tecplot, and open-source software, such as Gnuplot, are used. Grid adaption has not been carried out.

C. Machines description

The results for the easy test cases have been obtained on a Linux work station running Ubuntu 10.04 with an Intel i7-2600 CPU running at 3.4 GHz, with 8 Mb of cache. The machine contains 16 Gb of RAM memory with an equivalent amount of swap. Running the Taubench on this machine led to a CPU time of 5.59 seconds (average over 4 runs).

The difficult test cases were run on up to 512 processors on the LISA machine of SARA, the Dutch Supercomputing Center and Hexagon, the Cray XE6 machine of the University of Bergen. Running the Taubench on these machines led to a CPU time of 10.3 and 10.8 seconds respectively (average over 4 runs).

References


