New 2 and 3-dimensional non-oscillatory central finite volume methods on staggered Cartesian grids

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Abstract

We present a 3D finite volume generalization of the 1-dimensional Lax-Friedrichs and Nessyahu-Tadmor schemes for hyperbolic equations on Cartesian grids. The non-oscillatory central difference scheme of Nessyahu and Tadmor, in which the resolution of the Riemann problem at the cell interfaces is by-passed thanks to the use of the staggered Lax-Friedrichs scheme, is extended here to a two-step, tri-dimensional non-oscillatory centered scheme in finite volume formulation.

Piecewise linear cell interpolants using several van Leer-type limiting techniques to estimate the gradient (van Leer, van Albada, SuperBee, MinMod), lead to a non-oscillatory spatial resolution of order superior to 1. The fact that the expected second-order resolution is not fully attained in 3D is investigated first by considering an alternate dual grid (in 2D), and by using the original van Albada limiter in 3D.

Numerical results for a linear advection problem with continuous and discontinuous initial conditions in 2D and 3D show the accuracy and stability of the method. A comparison is made between the 2D Arminjon-Stanescu-Viallon and Jiang-Tadmor formulations and the new one. A new simple projection method is used for the gradients in the new 2D scheme. We also include results for the 3D Euler system (channel with a forward facing step).
Introduction

The non-oscillatory central difference scheme of Nessyahu and Tadmor is a Godunov-type scheme for one-dimensional hyperbolic conservation laws in which the resolution of the Riemann problems at the cell interfaces is by-passed thanks to the use of the staggered Lax-Friedrichs scheme. Piecewise linear MUSCL-type cell interpolants and slope limiters lead to an oscillation-free second-order resolution.

In earlier papers [4] [5] [3] we presented a two-dimensional finite volume method generalizing the one-dimensional Lax-Friedrichs [14] and Nessyahu-Tadmor [15] difference schemes for hyperbolic conservation laws to unstructured triangular grids, while in [2] we constructed a corresponding extension in the case of 2-dimensional Cartesian grids. These “central” finite volume methods share with the Lax-Friedrichs and Nessyahu-Tadmor schemes the two advantages of avoiding the resolution of Riemann problems at the cell interfaces, and being non-oscillatory.

In [12], Jiang and Tadmor have presented a slightly different extension to two-dimensional Cartesian grids; the difference between both extensions for rectangular grids is that in [2], the MUSCL (Monotonic Upstream-centered Scheme for Conservation Laws) variable extrapolation (“reconstruction technique”) [21],[22] is applied both to the vector of conservative variables and to the fluxes, while in [12], it is only used for the conservative variables. All these extensions to 2-dimensional problems lead to non-oscillatory, essentially second-order accurate finite volume methods.

In this paper, we first present a finite volume extension of the first-order accurate Lax-Friedrichs scheme to three-dimensional Cartesian grids, and show that the new scheme is monotone under an appropriate CFL condition. We then construct the corresponding extension of the Nessyahu-Tadmor difference scheme to a non-oscillatory finite volume method for 3-dimensional staggered Cartesian grids, which should in principle be second-order accurate.

As our first numerical experiments indicated orders between 0.8 and 1.3, we investigated two avenues to try and recover quasi-second order accuracy. First in three dimensions, if we apply the simplified form of the van Albada limiter in the computation of the predictor, and any reasonable limiter for the gradient of the piecewise linear reconstruction at time \( t^n \), we obtain accuracy orders between 0.7 and 1.86 depending on the choice of the limiter for the reconstruction step. Alternately, if we apply, as suggested by Venkatakrishnan [23], the original van Albada limiter [18], both to the reconstruction at time \( t^n \) and to the predictor at \( t^{n+1/2} \), we obtain second-order resolution, in the \( L_1 \) norm, for an appropriate choice of the parameter \( K \) (see [23] and section (1.7)). Unfortunately, we then lose strict monotonicity preservation, and get small oscillations in computationally sensitive regions e.g. near the supersonic inlet, as was the case in [23]. This is caused by the fact that the scheme leaves the reconstruction gradients unlimited or nearly unlimited in the near-constant regions as observed by Venkatakrishnan [23]. Choosing the smallest value of \( K \) compatible with second-order accuracy allows the limiter to remain active nearly everywhere except in regions where the solution is truly constant, and thus leads to nearly perfectly monotonous profiles where needed.

Our second field of investigation, as regards the recovery of second-order accuracy and preservation of monotonicity, was related to our earlier construction [2] for 2-dimensional Cartesian grids of a finite volume generalization of the Lax-Friedrichs (LF) and Nessyahu-Tadmor (NT) schemes. This generalization, for Cartesian grids, is not the direct transposition to rectangular grids of our earlier generalization of the LF and NT schemes to unstructured triangular grids [4] [5]; indeed it features cells, both for the original and dual grid, with sides parallel to the \( x,y \) axes, while for the direct analogous to our triangular scheme, the alternate quadrilateral cells \( L_{nm} \) of the dual grid should in fact be \( oblique \), with sides making 45 degree angles with the sides of the original cell \( C_{ij} \) (see figure (2)). With this new “oblique” version of our 2-dimensional finite volume extension of the Lax-Friedrichs and Nessyahu-Tadmor schemes to staggered Cartesian grids, we obtain second-order accuracy and monotonicity preservation, provided that we use standard limiters, without having to apply the original or modified van Albada limiter. In a forthcoming paper ([1]), we extend this new “oblique dual grid” approach to the corresponding 3-dimensional finite volume method with oblique dual cells, and show that we again obtain second order accuracy and monotonicity preservation without having to resort to special limiters such as the original van Albada limiter, and simply applying the
3-dimensional analog of the standard limiters successfully used with both 2-dimensional Cartesian versions (non oblique dual rectangular cells [2], oblique dual cells as described in subsection 2 of this paper) of the 2-dimensional finite volume method for Cartesian grids.

The organization of this paper is as follows. In section 1, we describe our 3-dimensional central finite volume methods of first and second order accuracy, which generalize the LF and NT schemes, respectively. The dual cells used here have sides parallel to the coordinate axes, as in [2]. The first-order accurate Lax-Friedrichs-type finite volume method is shown to be monotone under an appropriate CFL condition. We also describe the limiters used in this paper. In section 2, we construct the direct analogous, for Cartesian grids, of our finite volume generalization of the NT scheme for unstructured triangular grids. This finite volume method has oblique dual cells. In section 3, we present numerical experiments for the 3-dimensional method of section 1 and for the 2-dimensional oblique dual cell method of section 2, with, in the latter case, a comparison with our earlier 2-dimensional finite volume method with dual cell boundaries parallel to the axes, and with the Jiang-Tadmor variant of that method.

1 3-dimensional finite volume extensions of the Lax-Friedrichs and Nessyahu-Tadmor schemes

In this section we briefly expose how to construct a 3D finite volume generalization of the Nessyahu-Tadmor (1D) scheme on Cartesian grids. A 2D extension (ASV$^5$) on quadrilaterals was first proposed in 1995 by Arminjon et al. [2]. The ASV finite volume method differs from the Nessyahu-Tadmor scheme because not only the dependent variable vector $\vec{U}$ but also the fluxes are reconstructed using a MUSCL technique. Four items are needed to obtain the ASV formulation (on structured uniform hexahedrals), for each time step.

- A centered monotone scheme (which implies that this scheme is only first-order)
- A reconstruction of the solution with the MUSCL method to obtain second-order accuracy
- A predictor for the fluxes
- A precise CFL condition to prevent wave interaction between the original and dual grid

The first item is defined with the help of a staggered Lax-Friedrichs-type (LF) finite volume scheme. This scheme uses a uniform fully structured hexahedral grid for the dual grid. We also need to prove that this LF-type scheme is monotone. The reconstruction is then performed here both for the point values $\vec{U}_{ijk}$ and for the fluxes (see [2]). A predictor (in time) is then constructed by using a Taylor expansion combined with the conservation law (1). Finally, the CFL condition is computed using analytic and geometric arguments inspired by Hindmarsh et al. [9] and [5].

1.1 Generic model equation

We consider the 3-dimensional system of conservation laws

$$\vec{U}_t + \vec{\nabla} \cdot \vec{F}(U) = \vec{U}_t + \vec{f}_x + \vec{g}_y + \vec{h}_z = 0$$

where

$$\vec{F} = (\vec{f}, \vec{g}, \vec{h}) = ((f^1, f^2, ..., f^m)^T, (g^1, g^2, ..., g^m)^T, (h^1, h^2, ..., h^m)^T)$$

with the initial condition $\vec{U}(x, y, z, 0) = \vec{U}_0(x, y, z)$. System (1) will be assumed to be hyperbolic in the sense of the following definition.

$^5$Arminjon-Stanescu-Viallon
Definition 1.1 The system \( (1) \) is said to be hyperbolic if any linear combination of the \( m \times m \) Jacobian matrices \( A(\vec{U}), B(\vec{U}), C(\vec{U}) \) where

\[
\mathbf{A} = (A(\vec{U}), B(\vec{U}), C(\vec{U})) = \left( \frac{\partial \vec{f}}{\partial \vec{U}}, \frac{\partial \vec{g}}{\partial \vec{U}}, \frac{\partial \vec{h}}{\partial \vec{U}} \right)
\]

has \( m \) real eigenvalues \( \lambda^k(\vec{U}) \) and \( m \) linearly independent right and left eigenvectors \( \vec{r}^k(\vec{U}) \) and \( \vec{l}^k(\vec{U}) \) that satisfy

\[
\vec{r}^i \cdot \vec{l}^j = \delta_{ij}
\]

1.1.1 Notations

As the staggered Lax-Friedrichs and Nessyahu-Tadmor scheme are one-dimensional two step schemes using two staggered grids at alternate time steps, our three-dimensional finite volume extensions also use two staggered grids and two time steps.

For the first grid, the control volume or cell is the region \( C_{ijk} = [x_{i-\frac{1}{2}}, x_{i+\frac{1}{2}}] \times [y_{j-\frac{1}{2}}, y_{j+\frac{1}{2}}] \times [z_{k-\frac{1}{2}}, z_{k+\frac{1}{2}}] \), while for the second grid the cell will be defined as \( L_{ijk} = [x_i, x_{i+1}] \times [y_j, y_{j+1}] \times [z_k, z_{k+1}] \).

1.2 First time step, LF-type scheme

In the first (and further odd) time step of our 3-dimensional finite volume schemes, we start from initial (or previously obtained) cell average values \( \vec{U}_{C_{ijk}} \) for the cells of the first grid, and compute cell average values \( \vec{U}_{L_{ijk}} \) for the (staggered) dual cells of the second grid. This is done by integrating \( (1) \) on an extended control volume \( L_{ijk} \times [t^n, t^{n+1}] \), where \( L_{ijk} \) is a cell of the dual grid, and by assuming that \( \vec{U}_{L_{ijk}}(t^n + \Delta t) \) is piecewise constant on the cells of the second grid. The following equality is easily verified

\[
L_{ijk} = \bigcup_{lmn \text{ neighbour}} C_{lmn} \cap L_{ijk}
\]

where \( lmn \) means \( l = i, i+1; \ m = j, j+1; \ n = k, k+1 \) and in the sequel, \( \sum_{lmn} (\ ) \) stands for \( \sum_{lmn \text{ neighbour}} (\ ) \).

Using \( (1) \) and applying the divergence theorem to the flux term, we obtain

\[
\int_{t^n}^{t^{n+1}} \int_{L_{ijk}} \vec{U}_j \, dV \, dt + \int_{t^n}^{t^{n+1}} \int_{\partial L_{ijk}} \vec{F} \cdot \vec{n} \, dA \, dt = 0
\]

or

\[
\int_{L_{ijk}} \vec{U}(x,y,z,t^n+1) \, dV - \int_{L_{ijk}} \vec{U}(x,y,z,t^n) \, dV = - \int_{t^n}^{t^{n+1}} \int_{\partial L_{ijk}} \vec{F}(\vec{U}(x,y,z,t)) \cdot \vec{n} \, dA \, dt
\]

For a first-order accurate Lax-Friedrichs-type method, the R.H.S. is evaluated (according to the left rectangular quadrature formula) at time \( t^n \), using \( (3) \) and cell-average values:

\[
V(L_{ijk}) \vec{U}_{L_{ijk}}(t^{n+1}) - \sum_{lmn} V(C_{lmn} \cap L_{ijk}) \vec{U}_{lmn}(t^n) = -\Delta t \int_{\partial L_{ijk}} \vec{F}(\vec{U}^n) \cdot \vec{n} \, dA
\]

The function \( V(\ ) \) calculates the volume of the region common to the cell \( L_{ijk} \) (second grid) and cell \( C_{lmn} \) (first grid). Let us mention that the cells of either grid do not need to be cubic, in fact, they can be any rectangular prism in 3-space.
For a second-order accurate method generalizing the NT scheme, the R.H.S. of (5) is evaluated with the midpoint formula for integration with respect to time

$$\int_{t^n}^{t^{n+1}} \int_{\partial L_{ijk}} \vec{F}(\vec{U}(x, y, z, t)) \cdot \vec{n} \, dA \, dt \approx -\Delta t \int_{\partial L_{ijk}} \vec{F}(\vec{U}^{n+1/2}) \cdot \vec{n} \, dA$$

(7)

where $\vec{U}^{n+1/2}$ is some predicted value obtained with the help of (1). The corresponding calculations will be described in subsection (1.6).

1.3 Second time step, LF-type scheme

Once the first time step is completed, starting from the obtained dual cell values $\vec{U}_{L_{ijk}}$, we compute cell-average values $\vec{U}_{C_{ijk}}$ for the original cells $C_{ijk}$ of the first grid; the calculations are performed in a similar manner (see [2],[4],[5] for details in the 2-dimensional case).

1.4 fluxes

The computation of the flux for the scheme (R.H.S of (6)) is more complex than that of the conservative variables. Here, again, we assume that the fluxes are piecewise constant on $L_{ijk}$ and $C_{lmn}$ respectively.

We begin by labelling each of the faces of $\partial L_{ijk}$ and the corresponding orthonormal vectors:

$$\partial L_{ijk} = \partial L_{ijk}^1 \cup \partial L_{ijk}^2 \cup \partial L_{ijk}^3 \cup \partial L_{ijk}^4 \cup \partial L_{ijk}^5 \cup \partial L_{ijk}^6$$

$$\vec{n}_1 = \vec{n}_x \quad \vec{n}_2 = \vec{n}_y \quad \vec{n}_3 = \vec{n}_z$$

(8)

$$\vec{n}_4 = -\vec{n}_1 \quad \vec{n}_5 = -\vec{n}_2 \quad \vec{n}_6 = -\vec{n}_3$$

(9)

$$\vec{n}_x = (1, 0, 0), \quad \vec{n}_y = (0, 1, 0), \quad \vec{n}_z = (0, 0, 1)$$

(10)

Figure 1: Intersection of Cell $L_{ijk}$ and Cell $C_{lmn}$

Again, one can derive an expression similar to (3) for the faces. Each one of the faces can be written as the union of 4 subfaces resulting from the intersection with the $C_{lmn}$ of the first grid:

$$\partial L_{ijk}^q = \bigcup_{l_{mn} \text{ neighbour}} C_{lmn} \cap \partial L_{ijk}^q$$
where the notation neighbour now means that only four of the eight index combinations corresponding to
the notation \( \cup \) \( _{lmn} \) neighbour give rise to a non-empty intersection set. If \( A() \) is a function that calculates the
area of a given surface, the first step of the staggered LF scheme can be written in the following compact
form:

\[
\vec{U}_{Lijk}(t^n + \Delta t) = \frac{1}{V(L_{ijk})} \left( \sum_{lmn \text{ neighbour}} V(C_{lmn} \cap L_{ijk}) \vec{U}_{lmn}(t^n) - \Delta t \sum_{q=1}^{6} \sum_{lmn \text{ neighbour}} A(C_{lmn} \cap \partial L_{ijk}^q) \vec{F}_{lmn} \cdot \vec{n}_q \right)
\] (11)

where \( \vec{F}_{lmn} = \vec{F}(\vec{U}_{lmn}(t^n)) \)

1.5 Monotonicity condition

Let us recall here some basic definitions to make the article self-contained.

Definition 1.2 A general scheme for the model equation (1) can be written

\[
\vec{U}_{Lijk}^{n+1} = \mathcal{H}(\vec{U}_{C_{i-P,j-Q,k-R}}, \ldots , \vec{U}_{C_{i+P',j+Q',k+R'}})
\] (12)

Our finite volume scheme is a subcase of the form:

\[
\vec{U}_{Lijk}^{n+1} = \mathcal{H}(\vec{U}_{C_{i,j,k}}, \ldots , \vec{U}_{C_{i+1,j+1,k+1}}) = \mathcal{H}\{\vec{U}_{lmn}(t^n), (lmn \text{ neighbour})\}
\] (13)

Definition 1.3 The scheme (13) is monotone if \( \mathcal{H} \) is a monotone increasing function of each of its arguments.

\[
0 \leq \frac{\partial \mathcal{H}}{\partial \vec{U}_{lmn} \text{ (lmn neighbour)}}
\]
or equivalently if all eigenvalues of the matrix \( \frac{\partial \mathcal{H}}{\partial \vec{U}_{lmn} \text{ (lmn neighbour)}} \) are non negative.

Theorem 1.1 Under the CFL-like condition (22), the 3-dimensional finite volume extension (11) of the
LF scheme for structured uniform staggered grids is monotone.

Proof of 1.1 From (11) we find

\[
\frac{\partial \mathcal{H}}{\partial C_{lmn}} = \frac{V(C_{lmn} \cap L_{ijk})}{V(L_{ijk})} \mathcal{T} - \Delta t \sum_{q=1}^{6} \frac{A(C_{lmn} \cap \partial L_{ijk}^q)}{V(L_{ijk})} \vec{A}_{lmn} \cdot \vec{n}_q
\] (14)

for \( lmn \text{ neighbour} \) . To simplify the calculations, we introduce the following vector :

\[
\vec{K} = \sum_{q=1}^{6} \frac{A(C_{lmn} \cap \partial L_{ijk}^q)}{V(L_{ijk})} \vec{n}_q = (K_x, K_y, K_z)
\] (15)

we then have

\[
\sum_{q=1}^{6} \frac{A(C_{lmn} \cap \partial L_{ijk}^q)}{V(L_{ijk})} \vec{A}_{lmn} \cdot \vec{n}_q = \vec{A}_{lmn} \cdot \sum_{q=1}^{6} \frac{A(C_{lmn} \cap \partial L_{ijk}^q)}{V(L_{ijk})} \vec{n}_q = \vec{A}_{lmn} \cdot \vec{K} = \mathcal{M}
\] (16)

Here, \( \mathcal{M} = \mathcal{M}(l, m, n) \) represents a weighted sum of the 3 jacobian matrices \( A, B, C \). By our assumption
on the hyperbolicity of system (1), the matrix \( \mathcal{M} \) has real eigenvalues \( \lambda_s \) \( (1 \leq s \leq m) \) and can be
diagonalised by a similarity transformation (see also Hirsch Vol.II, pp.176-191 for the Euler equations). By
definition (1.3), monotonicity of our finite volume extension of the Lax-Friedrichs scheme will be established
if we can prove, in view of (14) - (16), that
\[ W = \frac{V(C_{lmn} \cap L_{ijk})}{V(L_{ijk})} I - \Delta t M \]  

is positive semi-definite. By the spectral theorem, the eigenvalues of \( W \) are the numbers

\[ \frac{V(C_{lmn} \cap L_{ijk})}{V(L_{ijk})} - \Delta t \lambda_s(M) \]

where \( \lambda_s(M) \) is an eigenvalue of \( M \). We therefore need

\[ \Delta t \lambda_s(M) \leq \frac{V(C_{lmn} \cap L_{ijk})}{V(L_{ijk})} \]

where \( s = 1, \ldots, m \) lmn neighbour

or finally

\[ \Delta t \max_{1 \leq s \leq m} \{ |\lambda_s(M)| \} \leq \frac{V(C_{lmn} \cap L_{ijk})}{V(L_{ijk})} \]

Remark 1.1 It might appear at first sight that (20) does not have the dimensional form of a CFL condition; but the matrix \( M \) is equal to the matrix product \( \vec{A}_{lmn} \cdot \vec{K} \) where \( \vec{K} \) is obviously \( O(h^{-1}) \) by (15), where \( h \) is the mesh size in the \( x, y, z \) directions, for the case of a uniform rectangular grid. The eigenvalues \( \lambda_s(M) \) are thus \( O(h^{-1}) \) and (20) does have the dimensions of a CFL condition.

Remark 1.2 For a perfectly uniform hexahedral grid the factor \( \frac{V(C_{lmn} \cap L_{ijk})}{V(L_{ijk})} \) reduces to \( \frac{1}{5} \), and the vector \( \vec{K} \) takes the simpler form \( \vec{K} = \frac{1}{3} \sum_{q=1}^{6} \delta(C_{lmn} \cap L_{ijk}) \vec{n}_q \), so that the monotonicity condition (20) is not very restrictive (here \( \delta(\cdot) \) represents a “Kronecker-like” set function which is equal to zero when the argument is the empty set, otherwise its value is 1); we then obtain the condition

\[ \frac{\Delta t}{h} \max_s \{ |\lambda_s(\vec{U}_{lmn})| \} \leq \frac{1}{2} \]

However this is only the monotonicity condition for a usual (non staggered) grid, the parameter \( h \) in (21) must be divided by two for obvious geometric reasons (since we are using staggered grids). The final monotonicity condition for the finite volume Lax-Friedrichs scheme can thus be written as

\[ \frac{\Delta t}{h} \max_s \{ |\lambda_s(\vec{U}_{lmn})| \} \leq \frac{1}{4} \]

Remark 1.3 In the scalar case, we can therefore expect the numerical solution \( u_h \) to converge to the unique solution of (1) satisfying the entropy condition, as \( h \to 0 \), if it does converge, see [13, 6, 7]. In the 2-dimensional case of unstructured triangular grids, we have proved [3], for a scalar linear hyperbolic equation, that the numerical solution satisfies a discrete maximum principle, and converges weakly* to a weak solution of (1). The proof can be directly extended to the case of 2-dimensional Cartesian grids, again for a scalar linear hyperbolic equation.
1.6 A second order finite volume method via MUSCL cellwise interpolation

In the one-dimensional Nessyahu-Tadmor difference scheme, the first order accuracy of the Lax-Friedrichs scheme was raised to second order by the use of van Leer’s MUSCL \([1, 2]\) cellwise piecewise linear interpolation of the conservative variables \(\text{but not of the flux functions}\), and a midpoint rule approach for the time discretisation of the flux integral, implemented with a predictor-corrector treatment of each time step. In the 3-dimensional finite volume generalization, we apply the same strategy to both the conservative variables \(\vec{u}\) and the flux functions \(\vec{f}, \vec{g}, \vec{h}\), following our 2-dimensional generalization of the NT scheme to Cartesian grids \([2]\). Let \(M_{C_{lmn}}^{\vec{u}}\) (resp. \(M_{C_{lmn}}^{\vec{f}}, M_{C_{lmn}}^{\vec{g}}, M_{C_{lmn}}^{\vec{h}}\)) denote the piecewise trilinear interpolant of the conservative variable \(\vec{u}\) (resp. of the flux functions \(\vec{f}, \vec{g}, \vec{h}\)) on cell \(C_{lmn}^\ast\):

\[
M_{C_{lmn}}^{\vec{u}}(\vec{R}) = \bar{U}_{C_{lmn}} + \nabla \bar{U}_{C_{lmn}} \cdot (\vec{R} - \bar{R}_{C_{lmn}})
\]

where \(\nabla \bar{U}_{C_{lmn}}\) is some limited approximate gradient of the interpolant, to be defined later (with similar definitions for \(M_{C_{lmn}}^{\vec{f}}, \ldots\) etc). \(\bar{R}\) is the position vector of an arbitrary point contained in \(C_{lmn}^\ast\), and \(\bar{R}_{C_{lmn}}\) denotes the centroid of the same cell.

Replacing in \((5)\) all the cellwise constant variables by their linear reconstruction \((23)\), integrating in time with the help of the midpoint rule for the flux integral, we obtain

\[
V(L_{ijk})\bar{U}_{L_{ijk}}^{n+1} = \int_{L_{ijk}} M_{L_{ijk}}^{\vec{u}} dV - \Delta t \int_{\partial L_{ijk}} (M_{L_{ijk}}^{\bar{u}+1/2} n_x + M_{L_{ijk}}^{\bar{u}+1/2} n_y + M_{L_{ijk}}^{\bar{u}+1/2} n_z) dA
\]

where \(M_{L_{ijk}}^{\bar{u}+1/2}\) and \(M_{L_{ijk}}^{\bar{u}+1/2}\) are the piecewise linear interpolants of \(\bar{U}\) and \(\bar{f}^{n+1/2} = \bar{f}(\bar{U}(t^{n+1/2}))\), respectively, in cell \(L_{ijk}\); \(\bar{U}(t^{n+1/2})\) is a predicted value defined below.

We now use the definitions of the control volumes \(C_{ijk}\) and \(L_{ijk}\) and introduce the following notations:

\[
\Delta x_i = x_{i+1/2} - x_{i-1/2}, \quad \Delta y_j = y_{j+1/2} - y_{j-1/2}, \quad \Delta z_k = z_{k+1/2} - z_{k-1/2}
\]

\[
V_{ijk} = \Delta x_i \Delta y_j \Delta z_k, \quad A_{ij} = \Delta x_i \Delta y_j
\]

For a uniform and isotropic grid, \(\Delta x_i = \Delta y_j = \Delta z_k = h\).

For simplicity, from now on, \(\bar{U}_{ijk}\) will mean \(\bar{U}_{C_{ijk}}\) i.e. the value of \(\bar{U}\) at the centroid of the cell on the first grid. Decomposing the limited gradient into its Cartesian parts gives:

\[
\nabla \bar{V}_{ijk} = (\bar{V}_{ijk,x}^{\lim}, \bar{V}_{ijk,y}^{\lim}, \bar{V}_{ijk,z}^{\lim})
\]

where for any function \(\bar{V}\), \(\bar{V}^{\lim}\) denotes the limited vector \(\bar{V}\). Our complete ASV-type 3d finite volume generalization of the NT scheme can now be derived and is given in the appendix. The intermediate-time fluxes are found with the help of a predictor step:

\[
\bar{U}_{ijk}^{n+1/2} = \bar{U}_{ijk}^n + \frac{\Delta t}{2} \bar{U}_{ijk,t}^n(t^n)
\]

Using the conservation equation \((1)\), \((26)\) can be written in a form which is easy to evaluate, replacing the gradients by their limited numerical approximations:

\[
\bar{U}_{ijk}^{n+1/2} = \bar{U}_{ijk}^n - \frac{\Delta t}{2} (\bar{f}_{ijk,x}^{\lim} + \bar{g}_{ijk,y}^{\lim} + \bar{h}_{ijk,z}^{\lim})
\]

where \(\bar{f}_{ijk,x}^{\lim}\) denotes the limited derivative of \(\bar{f}\) in the \(\hat{x}\) direction.

Here, an alternate choice for the predictor can be made, using the known Jacobian matrices of the flux function:

\[
\bar{U}_{ijk}^{n+1/2} = \bar{U}_{ijk}^n - \frac{\Delta t}{2} (A \cdot \bar{U}_{ijk,x}^{\lim,n} + B \cdot \bar{U}_{ijk,y}^{\lim,n} + C \cdot \bar{U}_{ijk,z}^{\lim,n})
\]
From a computational viewpoint, some inconveniences arise from the fact that (28) is very costly to evaluate even for the non-conservative Jacobians (which have a simpler form).

For each of these two predictors, two approaches to define the predicted values of the dependent variables at time $t_{n+1}$ can be used. By a proper transformation (see [10], 1990 and [7], 1996), one can pass from the conservative to the physical variables (when dealing with the Euler equations), make the prediction at $t_{n+1/2}$ and then use it to reconstruct the fluxes at $t_{n+1/2}$, to finally obtain $\vec{U}^{n+1}$. The other approach is to use the predictor on the conservative variables, and then rebuild the fluxes with the intermediate-time approximations.

1.7 Limiters

To give a short description of the MUSCL cell interpolation considered here with the various limiters mentioned earlier in this section and used in our numerical experiments, we follow the notation used in [23] and [16], and consider in the 1-dimensional case for simplicity piecewise linear reconstructions defined, at the cell interface $x_{i+1/2}$, by

\[
\begin{align*}
    u_{i+1/2}^L &= u_i + \frac{1}{2} \psi(\theta_i)(u_i - u_{i-1}), \\
    u_{i-1/2}^R &= u_i - \frac{1}{2} \psi(\frac{1}{\theta_i})(u_{i+1} - u_i)
\end{align*}
\]

where

\[
\theta_i = \frac{\Delta+u_i}{\Delta-u_i} \equiv \frac{u_{i+1} - u_i}{u_i - u_{i-1}}
\]

and $\psi(\theta)$ is the (slope) limiter function; we take $\psi(r) \equiv 0$ for $r \leq 0$. In this paper, we have used the following limiters.

(i) Simplified van Albada\(^6\)

\[
\psi_{VA}(\theta) = \frac{\theta^2 + \theta}{\theta^2 + 1}
\]

With this limiter, (29) takes the form

\[
\begin{align*}
    u_{i+1/2}^L &= u_i + \frac{1}{2} \frac{\theta^2 + \theta}{\theta^2 + 1}(u_i - u_{i-1})
\end{align*}
\]

with $\theta_i = \frac{\Delta+u_i}{\Delta-u_i} \equiv \frac{\Delta+}{\Delta-}$ for simplicity.

We note here that if we use a cell interface value defined from a centered approximation of the slope

\[
\begin{align*}
    u_{i+1/2}^L &= u_i + \frac{1}{2} R(\theta_i) \frac{u_{i+1} - u_{i-1}}{2}
\end{align*}
\]

van Albada’s simplified limiter must now be written in the form

\[
R(\theta_i) = \frac{2\theta_i}{1 + \theta_i^2}
\]

It can be easily verified that both (33) and (34) can be written as

\[
\begin{align*}
    u_{i+1/2}^L &= u_i + \frac{1}{2} \frac{\Delta_+^2 + \Delta_- + \Delta_-^2 \Delta_+}{\Delta_+^2 + \Delta_-^2}
\end{align*}
\]

(ii) Original van Albada\(^7\)

\(^6\)SVA: Simplified van Albada limiter. (We should mention here that in [23], the original and simplified van Albada limiters have been exchanged. Van Albada’s original limiter really is defined by (37), as communicated to the first author by B. van Leer.)

\(^7\)OVA: Original van Albada without Threshold.
This choice corresponds to modifying equation (36) as follows:

\[ u_{i+1/2}^L = u_i + \frac{1}{2} \frac{(\Delta_x^2 + \epsilon^2)\Delta_- + (\Delta_x^2 + \epsilon^2)\Delta_+}{\Delta_x^2 + \Delta_-^2 + 2\epsilon^2} \]  

(37)

where \( \epsilon^2 \) is made proportional to \( (\Delta x)^3 \), by setting \( \epsilon^2 = (K\Delta x)^3 \) where \( K \) is a constant. Notice that at an extremum of \( u \) or in a region where \( u \) is nearly constant, \( \epsilon^2 \approx O(h^3) \) will dominate \( \Delta_-^2, \Delta_+^2 \) by Taylor’s theorem, leading to a nearly unlimited value \( u_{i+1/2}^L \) (i.e. \( \psi \approx 1 \)) in (29), while if \( K \) is very large the same situation will prevail. It has been verified by Venkatakrishnan that for a fixed value of \( K \), the scheme will tend to be second-order accurate, but will not be monotonicity preserving for an advection problem. This difficulty arises from the fact that the scheme resumes to the unlimited form (34) with \( R(\theta) = 1 \) in the near-constant regions (see [23] for details). By restricting the action of the original van Albada limiter (37) to the neighbourhood of true extrema only, while using the simplified form (32)-(36) elsewhere, Venkatakrishnan obtained a globally second-order accurate scheme (for all usual norms) which is monotonicity preserving (the limiter is then the original van Albada limiter at a true extremum, but the simplified van Albada in near-constant regions). In this paper, for simplicity, we chose the following limiting strategy: we use the original van Albada limiter (37) everywhere except when both \( \tilde{\Delta}_+ = u_{i+1}^\text{max} - u_i \) and \( \tilde{\Delta}_- = U(x_{i+1/2}) - u_i \) are below a certain tolerance threshold, in which case we take the limiter equal to 1 (no limitation). This variant of the original van Albada limiter will be noted OVAT (Original van Albada with Threshold).

(iii) van Leer
In van Leer’s notation we consider piecewise linear interpolants

\[ \tilde{u}(x) = u_i + \delta_i \frac{(x-x_i)}{\Delta x}, \quad x_{i-1/2} \leq x \leq x_{i+1/2} \]  

(38)

with \( \theta_i \) as in (31) and

\[ \delta_i = R(\theta_i) \hat{\delta}_i \quad \text{and} \quad \hat{\delta}_i \equiv \frac{u_{i+1} - u_{i-1}}{2} \]  

(39)

the coefficient \( R(\theta_i) \) is a "slope limiter” related to Sweby’s [17] flux limiter \( \phi(r) \) by [7]

\[ \phi_{VL}(r) = \frac{R(\theta)(1 + \theta)}{2\theta} \quad \text{with} \quad r_i = \frac{\Delta_- u_i}{\Delta_+ u_i} = \frac{1}{\theta_i} \]  

(40)

We then have [7, 8] the corresponding forms of van Leer’s limiters

\[ R(\theta) = \frac{4\theta}{(1 + \theta)^2}, \quad \theta \geq 0 \]  

(41)

and

\[ \phi_{VL}(r) = \begin{cases} \frac{2r}{1+r} & r \geq 0 \\ 0 & r < 0 \end{cases} \]  

(42)

(iv) MinMod
In Sweby’s notation, the minmod limiter is defined by

\[ \phi_{MM}(r) = \begin{cases} \min(r, 1) & r \geq 0 \\ 0 & r < 0 \end{cases} \]  

(43)

in the slope limiter notation

\[ \text{MM}(\Delta_+, \Delta_-) = \frac{1}{2} \{ \text{sgn}(\Delta_+) + \text{sgn}(\Delta_-) \} \min\{ |\Delta_-|, |\Delta_+| \} \]  

(44)

(v) Superbee

\[ \phi_{SB}(r) = \max\{ 0, \min(2r, 1), \min(r, 2) \} \]  

(45)
2 2-dimensional Cartesian grid scheme with oblique dual cells ASCV

Notations
We consider equation (1) with a flux function \( \vec{F}(\vec{U}) = (\vec{f}(\vec{U}), \vec{g}(\vec{U})) \) in some rectangular region \( \Omega \) of the \( x-y \) plane, with initial condition
\[
\vec{U}(x, y, t = 0) = \vec{U}_0(x, y) \quad (x, y) \in \Omega
\]  
(46)
We shall describe the scheme in the case of a uniform rectangular grid which partitions \( \Omega \) into \( M^2 \) squares of equal area; the extension to arbitrary rectangular grids is straightforward, except for the initialization, which requires numerical integration, and the programming part of the resolution, where we have to proceed as in the case of a fully unstructured grid. If \( h = \Delta x = \Delta y = x_{i+1/2} - x_{i-1/2} = y_{j+1/2} - y_{j-1/2} \) is the mesh size, the nodes of the first grid are the points \( (x_i, y_j) = (ih, jh) \) for \( 0 \leq i, j \leq M \). We introduce a numbering of these nodes, noted
\[
a_i \quad 1 \leq i \leq (M + 1)(M + 1)
\]  
(47)
and consider, for an arbitrary node \( a_i \), the finite volume cell \( C_i \) defined, for the first grid, as the square (quadrilateral in the case of an arbitrary rectangular grid) obtained by joining the centroids of the four squares adjacent to \( a_i \) (figure (2)). For the second (dual) grid, the nodes are the midpoints \( m_{ij} \) of sides such as \( a_i, a_j \) of the original grid. The corresponding oblique dual cell \( L_{ij} \) is defined by joining the points \( a_i, a_j \) to the centroids \( I_{ij} \) and \( S_{ij} \) of the squares sharing the common side \( a_i a_j \).

![Cell C_i and Cell L_{ij}](image)

Let \( \vec{U}_i^n \cong \vec{U}(a_i, t^n) \) and \( \vec{U}_{ij}^{n+1} \cong \vec{U}(m_{ij}, t^{n+1}) \) denote the nodal or cell average values in the first and second grid at time \( t^n \) and \( t^{n+1} \), respectively \((n\ even)\). In the first step, we compute the approximation \( \vec{U}_{ij}^{n+1} \) of the problem (1)-(46) at the centroid of the oblique cell \( L_{ij} \), using a finite volume approach.

First time step
Integrating (1) in the extended cell \( L_{ij} \times [t^n, t^{n+1}] = [(L_{ij} \cap C_i) \cup (L_{ij} \cap C_j)] \times [t^n, t^{n+1}] \), we obtain
\[
\int_{t^n}^{t^{n+1}} \iint_{L_{ij}} \vec{U}_i dA dt = - \int_{t^n}^{t^{n+1}} \iint_{L_{ij}} \nabla \cdot \vec{F} dA dt \tag{48}
\]
or using Gauss’ divergence theorem
\[
\iint_{L_{ij}} \vec{U}(x, y, t^n+1) dA = \iint_{L_{ij} \cap C_i} \vec{U}(x, y, t^n) dA + \iint_{L_{ij} \cap C_j} \vec{U}(x, y, t^n) dA - \int_{t^n}^{t^{n+1}} \int_{\partial L_{ij}} \vec{f} n_x + \vec{g} n_y d\sigma dt \tag{49}
\]
\(^8\) Arminjon, St-Cyr, Viallon

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where the left-hand side defines the new cell value $\bar{U}_{ij}^{n+1}$

$$A(L_{ij})\bar{U}_{ij}^{n+1} \approx \iint_{L_{ij}} \bar{U}(x, y, t^{n+1})dA$$  \hspace{2cm} (50)

The first two integrals in the right-hand side of (49) can be written as

$$\left( \iint_{(L_{ij} \cap C_i)_S} + \iint_{(L_{ij} \cap C_i)_I} \right) \bar{U}(x, y, t^n)dA + \left( \iint_{(L_{ij} \cap C_j)_S} + \iint_{(L_{ij} \cap C_j)_I} \right) \bar{U}(x, y, t^n)dA$$  \hspace{2cm} (51)

where $(L_{ij} \cap C_i)_S = \text{triangle } a_im_{ij}S_{ij}$, $(L_{ij} \cap C_i)_I = \text{triangle } a_iI_{ij}m_{ij}$

$(L_{ij} \cap C_j)_S = \text{triangle } a_jS_{ij}m_{ij}$, $(L_{ij} \cap C_j)_I = \text{triangle } a_jm_{ij}I_{ij}$

Instead of splitting the integral on $L_{ij}$ into the contributions of four triangles, as above, we can split it into only two triangles without changing the accuracy of the approximation. The decomposition into four triangles will be used for the R.H.S of (48).

For a uniform Cartesian grid we have, to second-order accuracy:

$$\iint_{L_{ij}} \bar{U}(x, y, t^n)dA \cong \{ \bar{U}(x_i + \frac{\Delta x}{3}, y_i)A(L_{ij} \cap C_i) + \bar{U}(x_j - \frac{\Delta x}{3}, y_j)A(L_{ij} \cap C_j) \}$$  \hspace{2cm} (52)

To obtain second-order accuracy and preserve monotonicity of the solution, we introduce van Leer’s (MUSCL) piecewise linear interpolant, defined at a node $a_i$ (in analogy with (23),(25)) by

$$\tilde{U}_i(x, y, t^n) = \bar{U}_i^n + \bar{U}_i^{\text{lim},n,\frac{x-x_i}{\Delta x}} + \bar{U}_i^{\text{lim},n,\frac{y-y_i}{\Delta y}}$$  \hspace{2cm} (53)

Combining (52) and (53) now gives

$$\iint_{L_{ij}} \bar{U}(x, y, t^n)dA \cong \tilde{U}_i^n A(L_{ij} \cap C_i) + \tilde{U}_j^n A(L_{ij} \cap C_j) + \frac{1}{3} \bar{U}_i^{\text{lim},n} A(L_{ij} \cap C_i) - \frac{1}{3} \bar{U}_j^{\text{lim},n} A(L_{ij} \cap C_j)$$  \hspace{2cm} (54)

In the flux integral, the integration with respect to time can be approximated, to second-order accuracy, with the midpoint rule; the right-hand side of (48) is thus approximated by

$$\Delta t \iint_{\partial L_{ij}} \left( \tilde{f}(\bar{U}(x, y, t^{n+1/2}))n_x + \tilde{g}(\bar{U}(x, y, t^{n+1/2}))n_y \right)d\sigma$$  \hspace{2cm} (55)

which requires a prediction at the intermediate time $t^{n+1/2}$, both for $\bar{U}$ and for the fluxes, at the cell interfaces; for example, we need a prediction at the midpoint $a_{ij}^-$ of line segment $a_iI_{ij}$ (fig.2) to comply with the AV formulation [4], [3], [5]; this had not been applied in [2], where one had used a MUSCL reconstruction of the fluxes combined with an evaluation, at time $t^{n+1/2}$, performed directly at the vertices. We feel it is the extension of this procedure which is, in 3 dimensions, responsible (at least in part) for the loss of accuracy mentioned in the introduction, even though all steps of the approximations are in fact time - and - space second - order accurate.

We will use the following prediction at $a_{ij}^-$, considered as the value along line segment $a_iI_{ij}$ :

$$\bar{U}_{a_iI_{ij}}^{n+\frac{3}{2}} = \tilde{U}_i(a_{ij}^-, t^n) + \frac{\Delta t}{2} \tilde{U}_i(a_{ij}^-, t^n)$$  \hspace{2cm} (56)

In view of (1), we get

$$\bar{U}_{a_iI_{ij}}^{n+\frac{3}{2}} = \tilde{U}_i(a_{ij}^-, t^n) - \frac{\Delta t}{2}(A(\tilde{U}_i(a_{ij}^-, t^n))\bar{U}_{i,x}^{\text{lim}} + B(\tilde{U}_i(a_{ij}^-, t^n))\bar{U}_{i,y}^{\text{lim}})$$  \hspace{2cm} (57)
We note \( \tilde{f}(U_{n+1/2}^{a_i I_{ij}}) \) the corresponding predicted flux computed at the midpoint \( a_{ij}^- \) of \( a_{ij} \), at time \( t^{n+1/2} \), with the help of (57). For a uniform square grid, the quadrilateral cell \( L_{ij} \) is in fact a square rotated by 45 degrees. The unit vectors normal to the boundary of \( L_{ij} \) (fig. 2) are

\[
\hat{n}_1 = -\frac{\sqrt{2}}{2}(1,1), \quad \hat{n}_2 = \frac{\sqrt{2}}{2}(1,-1),
\]

\[
\hat{n}_3 = \frac{\sqrt{2}}{2}(1,1), \quad \hat{n}_4 = -\frac{\sqrt{2}}{2}(-1,1)
\]

By (55), the right-hand side of (48) is thus approximated as follows:

\[
\int_{t^n}^{t^{n+1}} \int_{\partial L_{ij}} (f\hat{n}_x + g\hat{n}_y) d\sigma dt = \int_{t^n}^{t^{n+1}} \sum_{q=1}^{4} \int_{\partial L_{ij}^n} (f\hat{n}_x,q + g\hat{n}_y,q) d\sigma dt =
\]

\[
\Delta t \frac{\sqrt{2}}{2} ((-f(U_{n+1/2}^{a_i I_{ij}}) - g(U_{n+1/2}^{a_i I_{ij}}))l(a_{i} I_{ij}) + (f(U_{n+1/2}^{a_i I_{ij}}) - g(U_{n+1/2}^{a_i I_{ij}}))l(a_{j} I_{ij}) + (f(U_{n+1/2}^{a_{i} S_{ij}}) + g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}) + (-f(U_{n+1/2}^{a_{i} S_{ij}}) + g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}))
\]

For the rotated square cell \( L_{ij} \), we have

\[
l(a_{i} I_{ij}) = l(\partial_{ij}^1) = \frac{\sqrt{2}}{2} h
\]

The first time step then takes the following form:

\[
\tilde{U}_{n+1}^{L_{ij}} A(L_{ij}) = \frac{h^2}{4} ((U_n^{i} + \tilde{U}_n^{j}) + \frac{1}{3} (U_{lim}^{i,x} - \tilde{U}_{lim}^{j,x}))
\]

\[
- \frac{1}{2} \frac{1}{h} \Delta t ((-f(U_{n+1/2}^{a_i I_{ij}}) - g(U_{n+1/2}^{a_i I_{ij}}))l(a_{i} I_{ij}) + (f(U_{n+1/2}^{a_i I_{ij}}) - g(U_{n+1/2}^{a_i I_{ij}}))l(a_{j} I_{ij})
\]

\[
+ (f(U_{n+1/2}^{a_{i} S_{ij}}) + g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}) + (-f(U_{n+1/2}^{a_{i} S_{ij}}) + g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}))
\]

hence:

\[
\tilde{U}_{n+1}^{L_{ij}} = \frac{1}{2} (U_n^{i} + \tilde{U}_n^{j}) + \frac{1}{6} (U_{lim}^{i,x} - \tilde{U}_{lim}^{j,x})
\]

\[
- \Delta t \frac{1}{h} ((-f(U_{n+1/2}^{a_i I_{ij}}) - g(U_{n+1/2}^{a_i I_{ij}}))l(a_{i} I_{ij}) + (f(U_{n+1/2}^{a_i I_{ij}}) - g(U_{n+1/2}^{a_i I_{ij}}))l(a_{j} I_{ij})
\]

\[
+ (f(U_{n+1/2}^{a_{i} S_{ij}}) + g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}) + (-f(U_{n+1/2}^{a_{i} S_{ij}}) + g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}))
\]

For a dual grid cell \( L_{ik} \) such that \( a_{i}a_{k} \) is parallel to the \( y \) axis, some minor changes are necessary and we obtain

\[
\tilde{U}_{n+1}^{L_{ik}} = \frac{1}{2} (U_n^{i} + \tilde{U}_n^{k}) + \frac{1}{6} (U_{lim}^{i,y} - \tilde{U}_{lim}^{k,y})
\]

\[
- \Delta t \frac{h}{1} ((-f(U_{n+1/2}^{a_i I_{ij}}) + g(U_{n+1/2}^{a_i I_{ij}}))l(a_{i} I_{ij}) + (f(U_{n+1/2}^{a_i I_{ij}}) + g(U_{n+1/2}^{a_i I_{ij}}))l(a_{j} I_{ij})
\]

\[
+ (-f(U_{n+1/2}^{a_{i} S_{ij}}) - g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}) + (-f(U_{n+1/2}^{a_{i} S_{ij}}) - g(U_{n+1/2}^{a_{i} S_{ij}}))l(a_{j} S_{ij}))
\]

\[
\text{2.1 Second time step}
\]

For the second (and further even) time step, we consider the original cell \( C_i \) (fig. 2) and the time interval \( t^{n+1} \rightarrow t^{n+2} \) \((n \text{ even})\). Proceeding as for the first step, we obtain

\[
\tilde{U}_{n+2}^{C_i} A(C_i) = \int \int C_i \tilde{U}(x, y, t^{n+2}) dA \cong \int \int C_i \tilde{U}(x, y, t^{n+1}) dA - \Delta t \int \int_{C_i} (f(x, y, t^{n+3/2}) \nu_x + g(x, y, t^{n+3/2}) \nu_y) d\sigma dA
\]

The first integral in the R.H.S. can be decomposed into 8 parts, and obvious symmetries in the case of a uniform square grid reduce this to 4 parts:

\[
\int \int_{C_i} \tilde{U}(x, y, t^{n+1}) dA = \frac{h^2}{4} ((\tilde{U}_c^{n+1} - \frac{1}{6} \tilde{U}_{lim,n+1}^{c,x}) + (\tilde{U}_d^{n+1} - \frac{1}{6} \tilde{U}_{lim,n+1}^{d,y})
\]

\[
+ (\tilde{U}_a^{n+1} + \frac{1}{6} \tilde{U}_{lim,n+1}^{a,x}) + (\tilde{U}_b^{n+1} + \frac{1}{6} \tilde{U}_{lim,n+1}^{b,y})
\]

\[
\text{12}
\]
The second step of the scheme can therefore be written as

\[ \nu_a = (-1,0) , \nu_b = (0,-1) , \nu_c = (1,0) , \nu_d = (0,1) \]

we obtain

\[ \int_{\partial C_i} (f\nu_x + g\nu_y) \, d\sigma \cong \frac{\Delta t}{2} \left( f(\hat{U}_{ci}^{n+\frac{3}{2}}) + f(\hat{U}_{ci}^{n+\frac{1}{2}}) - f(\hat{U}_{ci}^{n+\frac{1}{2}}) - f(\hat{U}_{ci}^{n+\frac{3}{2}}) \right) + g(\hat{U}_{di}^{n+\frac{3}{2}}) + g(\hat{U}_{di}^{n+\frac{1}{2}}) - g(\hat{U}_{di}^{n+\frac{1}{2}}) - g(\hat{U}_{di}^{n+\frac{3}{2}}) \]

The second step of the scheme can therefore be written as

\[ \hat{U}_{ci}^{n+2} A(C_i) = \frac{h^2}{4} (\hat{U}_{ci}^{n+1} + \hat{U}_{ci}^{n+1} + \hat{U}_{di}^{n+1} + \hat{U}_{di}^{n+1} + \frac{h^2}{2} \hat{U}_{c,i}^{n+1} - \hat{U}_{d,i}^{n+1} - \hat{U}_{a,i}^{n+1} - \hat{U}_{b,i}^{n+1} - \hat{U}_{c,i}^{n+1}) - \frac{\Delta t}{2} (f(\hat{U}_{ci}^{n+\frac{3}{2}}) + f(\hat{U}_{ci}^{n+\frac{1}{2}}) - f(\hat{U}_{ci}^{n+\frac{1}{2}}) - f(\hat{U}_{ci}^{n+\frac{3}{2}}) + g(\hat{U}_{di}^{n+\frac{3}{2}}) + g(\hat{U}_{di}^{n+\frac{1}{2}}) - g(\hat{U}_{di}^{n+\frac{1}{2}}) - g(\hat{U}_{di}^{n+\frac{3}{2}})) \] (63)

### 3 Numerical experiments

Numerical experiments for the methods presented in sections 1-2 will be described in the next two subsections.

#### 3.1 3-dimensional ASV formulation

##### 3.1.1 Linear advection

We consider the scalar advection equation

\[ u_t + au_x + bu_y + cu_z = 0 \]

with initial condition \( u(x,y,z,0) = u_0(x,y,z) \equiv f_0^{test} \) where \( (a,b,c) \) will be chosen equal to \((1,1,1)\) here.

We consider the cases of a discontinuous and of a continuous initial function :

\[ f_0^{test 1}(x,y,z) = \begin{cases} 0.5 & \text{if } (x,y,z) \in [\frac{7}{20}, \frac{12}{20}]^3 \\ 0 & \text{else} \end{cases} \]

\[ f_0^{test 2}(x,y,z) = \cos(2\pi x) \cos(2\pi y) \cos(2\pi z) \]

The spatial computational domain for both tests is the cube \([0,1] \times [0,1] \times [0,1]\). In table 1, \( L^1 \) denotes the error calculated with the \( L^1 \) norm, and \( N \) is the number of intervals in each spatial direction. In tables 1 and 2, SVA/VL means that the simplified van Albada limiter was used for the prediction and the van Leer limiter for the reconstruction, with similar signification for SVA/MM, SVA/SB (i.e. predictor / reconstruction).

Table 1 shows the results for the discontinuous initial function; we note that in three dimensions, the limiters which are smoother give better accuracy, as has generally been observed in the literature [23, 11]. These results give some insight into the difficulties encountered in 3-dimensional calculations. We observe that the second grid refinement is much more efficient than the first; the order \( O_{L^1} \) indicated in the last column is the average of the two orders obtained from each grid refinement.

Tables 2 and 3 present a test with a continuous initial function. We see that with the smoother simplified van Albada and van Leer limiters, second-order accuracy is approached while it is even slightly exceeded with the original van Albada limiter.

Figure 3 shows the effect of varying the parameter \( K \) in \( \epsilon^2 = (K\Delta x)^3 \). For small values of \( K \) (say less than 3) the order increases with \( K \), but we found that monotonicity is no longer preserved if \( K \) grows beyond a particular value which depends on the smoothness of the initial function; smoother data allow higher values of \( K \) before encountering monotonicity breaches.
3.1.2 3-Dimensional Euler equations

We considered the classical test problem [24] of supersonic flow in a channel with a forward facing step, extended here to a three-dimensional geometry corresponding to the computational domain \([0,3] \times [0,1] \times [0,1]\) (in dimensionless units). The length, width and height of the step are \(2.4\), \(1\) and \(0.2\), respectively.

Figure 4 (left) shows the results obtained (in the unstationary mode) with the simplified van Albada limiter, at time \(t = 3.05\). Although the density contours are smooth, the forward shock is not very sharp and the reflected ones even weaker.

Figure 4 (right) presents the results obtained with the original van Albada limiter, with \(K = 3\). The forward and reflected shocks are somewhat sharper, but we are still far from the sharp structure observed e.g. in [24, 2]. This is not only the consequence of some difficulties which might have been encountered by our method, but can also be attributed, at least in part, to the fact that we are dealing with three-dimensional shocks.

<table>
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<tr>
<th>Limiters</th>
<th>(L^1, N = 16)</th>
<th>(L^1, N = 32)</th>
<th>(L^1, N = 64)</th>
<th>(O_{L^1})</th>
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</thead>
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<td>OVA (K=3)</td>
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<td>0.003764</td>
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<td>0.36</td>
</tr>
<tr>
<td>SVA/SB</td>
<td>0.010967</td>
<td>0.010627</td>
<td>0.006385</td>
<td>0.36</td>
</tr>
</tbody>
</table>

Table 1: \(L^1\) - error and accuracy order, discontinuous problem (test 1) 3D case

<table>
<thead>
<tr>
<th>Limiter</th>
<th>(L^1, N = 16)</th>
<th>(L^1, N = 32)</th>
<th>(L^1, N = 64)</th>
<th>(O_{L^1})</th>
</tr>
</thead>
<tbody>
<tr>
<td>OVA</td>
<td>0.032720</td>
<td>0.009580</td>
<td>0.002794</td>
<td>1.77</td>
</tr>
<tr>
<td>SVA/VL</td>
<td>0.023744</td>
<td>0.006406</td>
<td>0.001813</td>
<td>1.86</td>
</tr>
<tr>
<td>SVA/MM</td>
<td>0.156609</td>
<td>0.099203</td>
<td>0.05793</td>
<td>0.72</td>
</tr>
<tr>
<td>SVA/SB</td>
<td>0.152270</td>
<td>0.099262</td>
<td>0.057661</td>
<td>0.70</td>
</tr>
</tbody>
</table>

Table 2: \(L^1\) - error and accuracy order, continuous problem (test 2), 3D case

<table>
<thead>
<tr>
<th>(N)</th>
<th>(L^1(K=0))</th>
<th>(O_{L^1})</th>
<th>(L^1(K=0.75))</th>
<th>(O_{L^1})</th>
<th>(L^1(K=1.5))</th>
<th>(O_{L^1})</th>
</tr>
</thead>
<tbody>
<tr>
<td>8</td>
<td>0.095168</td>
<td>0.032114</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>16</td>
<td>0.032720</td>
<td>1.54</td>
<td>0.009070</td>
<td>1.82</td>
<td>0.007669</td>
<td>1.94</td>
</tr>
<tr>
<td>32</td>
<td>0.009580</td>
<td>1.77</td>
<td>0.002444</td>
<td>1.89</td>
<td>0.001876</td>
<td>2.03</td>
</tr>
</tbody>
</table>

Table 3: continuous problem (test 2) OVA limiter, 3D case
Figure 3: $L^1$ error Original van Albada limiter with $K \in [0.25, 2.25]$}

Figure 4: Density $t=3.05$, 61 contours, Grid 240x10x80, simplified van Albada (left) and OVAT with $K=3$ (right)

### 3.2 New 2D ASCV formulation

We considered initial values problems for the advection equation

$$u_t + au_x + bu_y = 0 \quad 0 \leq x, y \leq 1$$

where we chose $a = b = 1$ here, with the initial functions

$$f_{test 1}^0(x, y) = \begin{cases} 0.5 & \text{if } (x - 1/2)^2 + (y - 1/2)^2 < 1/8 \\ 0 & \text{elsewhere} \end{cases}$$

and $$f_{test 2}^0(x, y) = \sin(\pi(x + y))$$

Table 4 shows the results obtained with the new “oblique dual cell” ASCV finite volume method, using the simplified van Albada limiter (SVA). $N$ is the number of intervals in each spatial direction, and $N_t$ is the number of double time steps. On the right, for comparison’s sake, we have given the results obtained with the former ASV method (with cell boundaries parallel to the $x$ or $y$ axes), using van Leer’s limiter (VL).

For this discontinuous problem, both methods seem to be comparable. The slight advantage provided by the ASV method could be due to the reconstruction of the fluxes and the fact that van Leer’s limiter might lead to better results than the simplified van Albada limiter, as already observed for our 3-dimensional tests (tables 1 and 2).

For the second (continuous) test function, we have compared the new oblique dual cell method (table 5) with the results obtained by Jiang and Tadmor [12] for the same test (table 6) with a 2-dimensional
extension of the Nessyahu-Tadmor scheme similar to the ASV extension but without reconstruction of the fluxes.

The new ASCV method leads to better $L^1$ and $L^\infty$ error with the final grid, and to higher orders of accuracy. We note that we have performed 400 double time steps, while the number of time steps performed in [12] is not indicated.

<table>
<thead>
<tr>
<th>N</th>
<th>$N_t$</th>
<th>SVA $L^1$</th>
<th>$L^\infty$</th>
<th>van Leer $L^1$</th>
<th>$L^\infty$</th>
</tr>
</thead>
<tbody>
<tr>
<td>85</td>
<td>30</td>
<td>1.149E-2</td>
<td>2.615E-1</td>
<td>5.835E-3</td>
<td>2.929E-1</td>
</tr>
</tbody>
</table>

Table 4: test 1 new scheme ASCV (left) old scheme ASV (right)

<table>
<thead>
<tr>
<th>N</th>
<th>ASCV $L^1$</th>
<th>ASCV $L^\infty$</th>
<th>$O_L^1$</th>
<th>$O_L^\infty$</th>
</tr>
</thead>
<tbody>
<tr>
<td>40</td>
<td>2.352E-2</td>
<td>6.003E-2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>80</td>
<td>5.759E-3</td>
<td>2.303E-2</td>
<td>2.03</td>
<td>1.38</td>
</tr>
<tr>
<td>160</td>
<td>1.347E-3</td>
<td>8.759E-3</td>
<td>2.10</td>
<td>1.39</td>
</tr>
</tbody>
</table>

Table 5: test 2, 2D ASCV

<table>
<thead>
<tr>
<th>N</th>
<th>Jiang-Tadmor [12] $L^1$</th>
<th>Jiang-Tadmor $L^\infty$</th>
<th>$O_L^1$</th>
<th>$O_L^\infty$</th>
</tr>
</thead>
<tbody>
<tr>
<td>40</td>
<td>1.93E-2</td>
<td>4.91E-2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>80</td>
<td>5.70E-3</td>
<td>2.12E-2</td>
<td>1.76</td>
<td>1.21</td>
</tr>
<tr>
<td>160</td>
<td>1.55E-3</td>
<td>8.90E-3</td>
<td>1.88</td>
<td>1.25</td>
</tr>
</tbody>
</table>

Table 6: test 2, 2D NT

4 Concluding remarks

We have presented new 2 and 3-dimensional non-oscillatory central finite volume methods for staggered grids\(^9\). Starting from our two-dimensional finite volume generalization for Cartesian grids of the Nessyahu-Tadmor 1-dimensional finite difference scheme, we have extended it to a 3-dimensional finite volume method using two staggered Cartesian grids at alternate time steps. Our numerical experiments have shown that for continuous initial data, the method is non-oscillatory and second order accurate when coupled with the original van Albada limiter; slight under/overshoots, for discontinuous data, can arise for inadequate choices of the limiter parameter $K$.

Having observed these under/overshoots when using the original van Albada limiter, and orders of accuracy inferior to the theoretical (in the absence of discontinuities) second order predicted by the principle of the method when using the simplified van Albada limiter, and in an effort to simplify the computational complexity of our earlier 2-dimensional method, where the reconstruction step was applied both to the dependent variables and to the fluxes, we have investigated here the possibility of a direct transposition of our 2-dimensional finite volume method for staggered unstructured triangular grids to the case of 2-dimensional Cartesian grids. This new “oblique dual cell” method (ASCV) seems to converge faster in all norms than the Jiang-Tadmor 2-dimensional extension of the NT scheme, while having comparable

---

\(^9\)we will use the acronym NOCS for “Non-Oscillatory Central Scheme” for these methods
accuracy; on the other hand, its accuracy is comparable with the earlier 2-dimensional version (ASV), but computing times are reduced since we do not reconstruct the fluxes; the 2-dimensional ASCV method can therefore be considered as a slight improvement of both earlier methods (ASV and JT).

These considerations have encouraged us to proceed to three-dimensional computations on unstructured tetrahedric grids, using grid adaptation [1].

Acknowledgments

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Appendix

Using equation (11) and the reconstruction (23) the complete ASV second order scheme in 3D can be written:

First time step

(i) reconstruction of conservative variables

\[
M^T_{C_{lmn}}(\vec{R}_{ijk}) = \vec{U}_{C_{lmn}}^n + \frac{1}{2} \nabla \vec{U}_{C_{lmn}}^n \cdot (\vec{R}_{ijk} - \vec{R}_{C_{lmn}}) \tag{64}
\]

\(\vec{R}_{ijk}\) represents the center of cell \(L_{ijk}\) and \(\vec{R}_{C_{lmn}}\) the center of cell \(C_{lmn}\).

(ii) predictor

\[
\vec{U}^{n+1/2}_{ijk} = \vec{U}_{ijk}^n - \frac{\Delta t}{2} (A \cdot \vec{U}_{ijk,x}^{\text{lim},n} + B \cdot \vec{U}_{ijk,y}^{\text{lim},n} + C \cdot \vec{U}_{ijk,z}^{\text{lim},n}) \tag{65}
\]

(iii) reconstruction of the flux

\[
M^F_{C_{lmn}}(\vec{R}_{\partial L_{ijk}^q}) = \vec{F}_{C_{lmn}}^{n+1/2} + \nabla \vec{F}_{C_{lmn}}^{n+1/2} \cdot (\vec{R}_{\partial L_{ijk}^q} - \vec{R}_{C_{lmn}}) \tag{66}
\]

\(\vec{R}_{\partial L_{ijk}^q}\) is the center of face \(q\) of the \(L_{ijk}\) cell.

Using the facts that \(\Delta x = \Delta y = \Delta z = h\) and \(\frac{V(C_{lmn} \cap L_{ijk})}{V(L_{ijk})} = \frac{1}{8}\) and \(\frac{A(C_{lmn} \cap \partial L_{ijk}^q)}{V(L_{ijk})} = \frac{1}{4h}\), (11) can be re-written as:

(iv) corrector

\[
\vec{U}^{n+1}_{L_{ijk}} = \frac{1}{8} \sum_{\text{neighbour}} M^T_{C_{lmn}}(\vec{R}_{ijk}) - \frac{\Delta t}{4h} \sum_{q=1}^{6} \sum_{\text{neighbour}} M^F_{C_{lmn}} \frac{1}{4h} \sum_{\text{neighbour}} (\vec{R}_{\partial L_{ijk}^q}^n \cdot \vec{n}_q) \tag{67}
\]

Second time step

Repeat steps (i) to (iv) starting from the values on the staggered grid \(L_{ijk}\).
References


