

Solving the Euler equations in 2D using a finite-volume method

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Abstract

A cell-centered finite volume method (FVM) has been used to simulate an inviscid flow in a channel with a bump. The solver uses the Advection Upstream Splitting Method (AUSM) and a first-order Roe scheme to find the inviscid fluxes across the cell faces. Results have been compared for both the schemes. Additionally, a higher-order state reconstruction has been done at the interfaces for more accurate flux calculations. This has been done using the Monotonic Upstream-centered Scheme for Conservation Laws (MUSCL) and has been used with the AUSM scheme.

Keywords: Channel flow, AUSM, Roe scheme, MUSCL, Finite-volume method.

1 Introduction

The cell-centered finite volume method (FVM) has been a widely used and well-validated method for computationally solving flow equations over various domains (both, internal and external). One of the main keypoints that differentiate FVMs from finite-difference methods (FDM) is the requirement to solve the equations in their conservative form, rather than a non-conservative differential form. Therefore, traditionally FVMs have had a better capability in handling a wide range of flows as singularities in the equations can be handled better if they are in their conservative form. But major additional steps that come up in FVMs are the handling of numerical fluxes as cell-centered FVMs solve average values within every cell and leads to numerical discontinuities at every interface. The information from one cell to the next cell has to travel in the form of fluxes through their included interface. There have been several approaches to formulating these fluxes. For doing so, the practice of solving a Riemann problem is widely done at each interface. The Godunov method gives the exact solution to the Riemann problem but runs into various problems while handling non-linear problems such as the Euler equations. For such problems, approximate solutions to the Riemann problem are used as shown by Roe.²

Furthermore, it is a common practice to formulate the fluxes at the interface as some function of the neighboring cell states, using a flux vector-splitting method as used by Liou and Steffen³ to formulate the Advection Upstream Splitting Method (AUSM). It is an extension to the splitting scheme given by van Leer.⁴

Here, 2D Euler equations have been solved by FVM for a flow within a channel and over a bump. Three flow Mach numbers have been considered, subsonic, transonic and supersonic. For the flux calculations first order schemes such as AUSM has been used and an approximate Riemann problem solution based on the Roe scheme have been used. The flux calculation has also been extended to higher order state reconstruction at the faces using the Monotonic Upstream-centered Scheme for Conservation Laws (MUSCL).

The following section, Computational Methodology, discusses the set-up of the problem, followed by the discretization method and the flux scheme used. Following this section, the Results section compares the solutions plots for the different dissipation values and gives an insight to the

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2. P.L. Roe, "Approximate Riemann Solvers, Parameter Vectors, and Difference Schemes," *Journal of Computational Physics*, October 1981, [https://doi.org/10.1016/0021-9991\(81\)90128-5](https://doi.org/10.1016/0021-9991(81)90128-5).

3. Meng-Sing Liou and Christopher J. Steffen Jr., "A New Flux Splitting Scheme," *Journal of Computational Physics*, July 1993, <https://doi.org/10.1006/jcph.1993.1122>.

4. Eighth International Conference on Numerical Methods in Fluid Dynamics, *Flux-Vector Splitting for the Euler Equation* (Springer, 1982), https://doi.org/10.1007/3-540-11948-5_66.

numerical scheme used. Finally, in the Conclusion section, we understand the effectiveness of using the schemes and effect of the higher order schemes.

2 Computational methodology

2.1 Problem Solved

The 2D Euler equations in conservative form are given as follows:

$$\mathbf{q}_t + \mathbf{f}_x + \mathbf{f}_y = \mathbf{0} \quad (1)$$

where,

$$\mathbf{q} = \begin{bmatrix} \rho \\ \rho u \\ \rho v \\ \rho e_T \end{bmatrix}, \quad \mathbf{f} = \begin{bmatrix} \rho u_n \\ \rho u^2 + p \\ \rho v^2 + p \\ \rho u h_0 \end{bmatrix},$$

$x_y = \frac{\partial x}{\partial y}$, e_T = specific total energy and h_0 = specific total enthalpy

2.2 Discretization

The discretized form for solving using FVM was as follows:

$$\frac{\bar{q}_{i,j}^{n+1} - \bar{q}_{i,j}^n}{\Delta t} + \frac{\sum_k \tilde{F}_k}{\Delta V} = 0 \quad (2)$$

where, $\tilde{F}_k = A_k \bar{F}_k$ in which, \bar{F}_k is the flux vector of each face, qualitatively similar to the \bar{f} defined above and A_k is the corresponding face area.

The flux at an interface, say $i + \frac{1}{2}, j$ that lies between the cells i, j and $i + 1, j$ is calculated specifically for the scheme at use.

2.3 Flux scheme

The flux given by the AUSM is as follows:

$$\tilde{F}_{i+\frac{1}{2},j} = A_{i+\frac{1}{2},j} (\rho_i C_{LS}^+ [1, u, v, h_0]_i^T + \rho_{i+1} C_{LS}^- [1, u, v, h_0]_{i+1}^T + (\tilde{D}_i p_i + \tilde{D}_{i+1} p_{i+1}) [0, n_x, n_y, 0]^T) \quad (3)$$

where, C_{LS} are the Liou-Steffen polynomials, which themselves are functions of the van Leer polynomials C_{VL} . The \tilde{D} 's are the dissipation coefficients. More on the formulations of C_{LS} and C_{VL} can be found in the papers by Liou and Steffen⁵ and van Leer,⁶ respectively.

The flux formulated using the Roe scheme is based on the dissipation due to the 4 wavespeeds (coming from the 4 equations being solved, i.e., \tilde{u}_n , \tilde{u}_n , $|\tilde{u}_n + \tilde{a}|$ and $|\tilde{u}_n - \tilde{a}|$), where \tilde{u}_n is the Roe-averaged normal velocity and \tilde{a} is the Roe-averaged local sound speed. At supersonic speeds, all the characteristic wavespeeds are positive whereas at subsonic speeds three characteristic wavespeeds are positive and one is negative. The total flux at a face is given as the average of the neighboring states and sum of the dissipation due to the four characteristic wavespeeds. The flux is as follows:

5. Liou and Jr., "A New Flux Splitting Scheme."

6. Eighth International Conference on Numerical Methods in Fluid Dynamics, *Flux-Vector Splitting for the Euler Equation*.

$$\tilde{F}_{i+\frac{1}{2},j} = \frac{1}{2} A_{i+\frac{1}{2},j} ([\rho u_n, \rho u + p, \rho v + p, \rho u_n h_0]_i^T + [\rho u_n, \rho u + p, \rho v + p, \rho u_n h_0]_{i+1}^T - \tilde{D}) \quad (4)$$

here, \tilde{D} will have 4 components, D_1 and D_2 corresponding to wavespeeds $|\tilde{u}_n|$, D_3 corresponding to wavespeed $|\tilde{u}_n - \tilde{a}|$ and D_4 corresponding to wavespeed $|\tilde{u}_n + \tilde{a}|$. The D 's are given as follows:

$$D_{1,2} = |\tilde{u}_n| [(\Delta\rho - \frac{\Delta p}{\tilde{a}^2})[1, \tilde{u}, \tilde{v}, \frac{\tilde{u}^2 + \tilde{v}^2}{2}]^T + [0, \Delta u - |\Delta u_n| n_x, \Delta v - |\Delta u_n| n_y, \tilde{u}\Delta u + \tilde{v}\Delta v - \tilde{u}_n\Delta u_n]^T] \quad (5)$$

$$D_3 = |\tilde{u}_n - \tilde{a}| [(\frac{\Delta p - \tilde{\rho}\tilde{a}\Delta u_n}{2\tilde{a}^2})[1, \tilde{u} - \tilde{a}n_x, \tilde{v} - \tilde{a}n_y, \tilde{h}_0 - \tilde{a}\tilde{u}_n]^T] \quad (6)$$

$$D_4 = |\tilde{u}_n + \tilde{a}| [(\frac{\Delta p + \tilde{\rho}\tilde{a}\Delta u_n}{2\tilde{a}^2})[1, \tilde{u} + \tilde{a}n_x, \tilde{v} + \tilde{a}n_y, \tilde{h}_0 + \tilde{a}\tilde{u}_n]^T] \quad (7)$$

More on the formulation of the Roe fluxes can be found in the book by Blazek.⁷

2.4 Mesh

A structured grid as shown in figure 1 has been used for the computation.

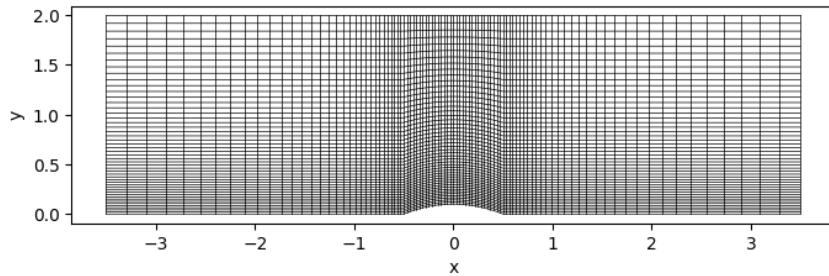


Figure 1: Mesh with 96 x 48 cells

2.5 Boundary conditions

The boundary conditions on the inlet had freestream Mach numbers of 0.4, 0.8 and 1.4 and had a static gauge pressure of 0.01 atm on the inlet. Both the top and bottom faces have been treated as slip walls.

2.6 Higher order state reconstruction

The states at the interfaces have been reconstructed from the cell-center values using the MUSCL technique, wherein the state at the left interface of a cell is the state at the interface subtracted by a higher order term. This higher order term is a function of the state at the cell and the state difference between itself and the neighboring cells acted upon by a limiter. The limiter is used to limit the slope of the reconstructed functions in order to minimize the oscillations in the solution. In this work, a *minmod* limiter has been used. Similarly the state on the right interface is reconstructed by adding the higher order term to the cell state values.

⁷. Jiri Blazek, *Computational Fluid Dynamics: Principles and Applications* (Elsevier, 2015), <https://doi.org/10.1016/C2013-0-19038-1>.

3 Results and Discussions

The various plots and graphs have been compared and inferred from in this section.

3.1 Flow variables for linear reconstruction

The pressure plots for all the cases are shown in figures 2 through 4. We see that both the schemes give similar results in case of the subsonic and transonic flows. However, when it comes to the supersonic flows, AUSM seems to not be able to develop the trailing shock completely.

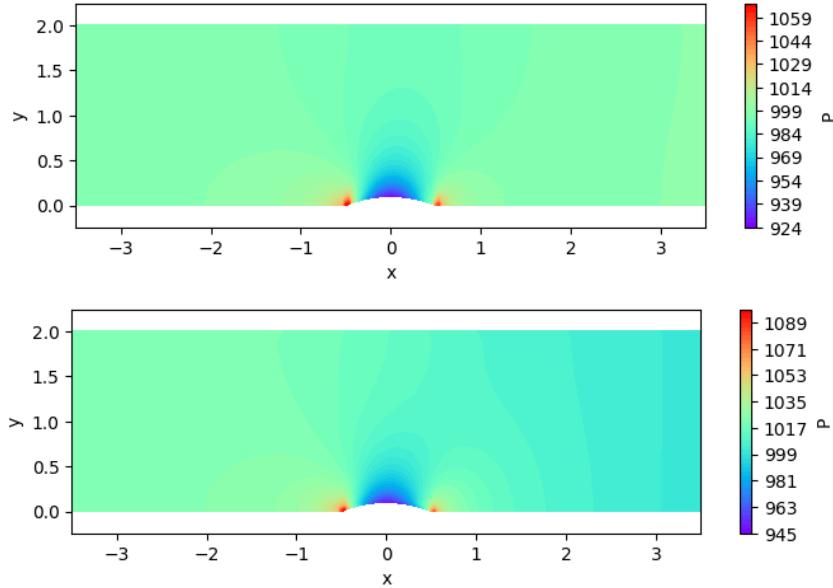


Figure 2: Pressure contours at $M=0.4$ computed using AUSM (top) and Roe (bottom) schemes

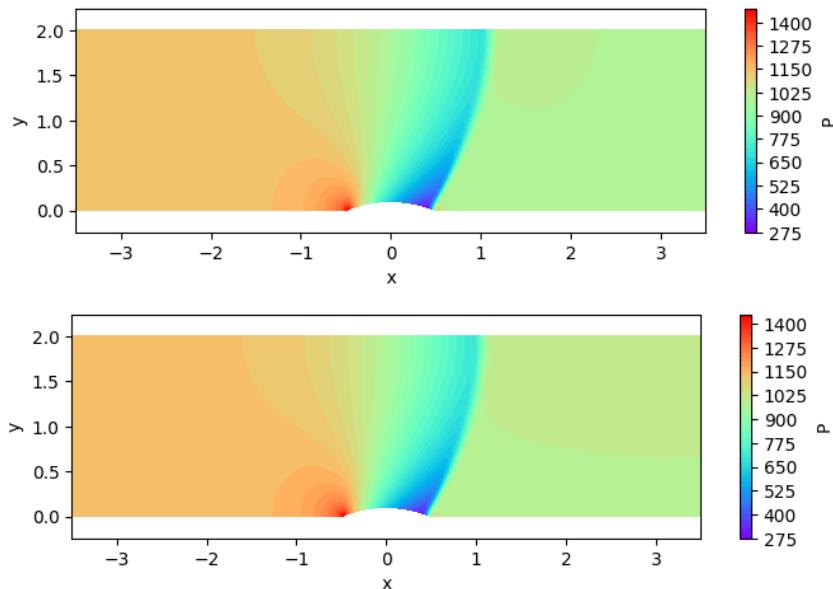


Figure 3: Pressure contours at $M=0.8$ computed using AUSM (top) and Roe (bottom) schemes

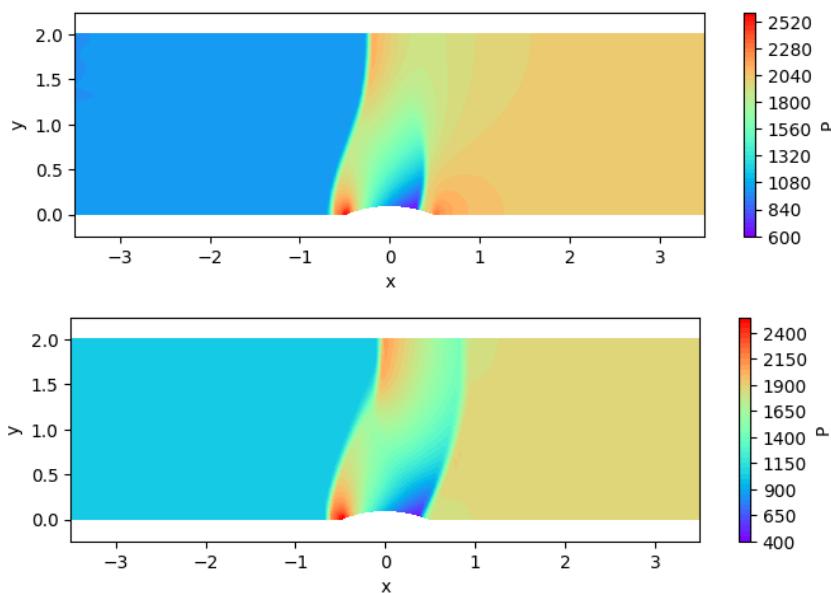


Figure 4: Pressure contours at $M=1.4$ computed using AUSM (top) and Roe (bottom) schemes

The Mach number contours are shown for the transonic and supersonic cases ($M=0.8$) in figures 5 and 6:

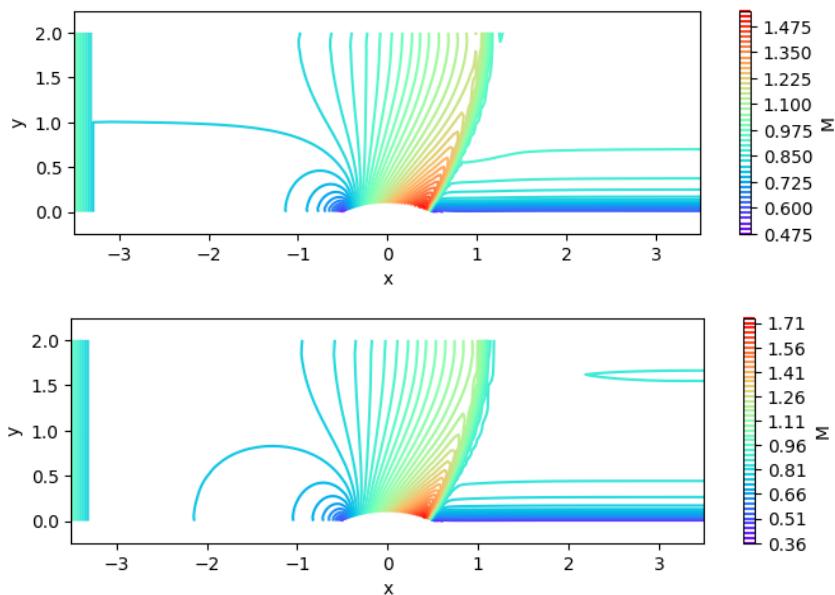


Figure 5: Mach number contours at $M=0.8$ computed using AUSM (top) and Roe (bottom) schemes

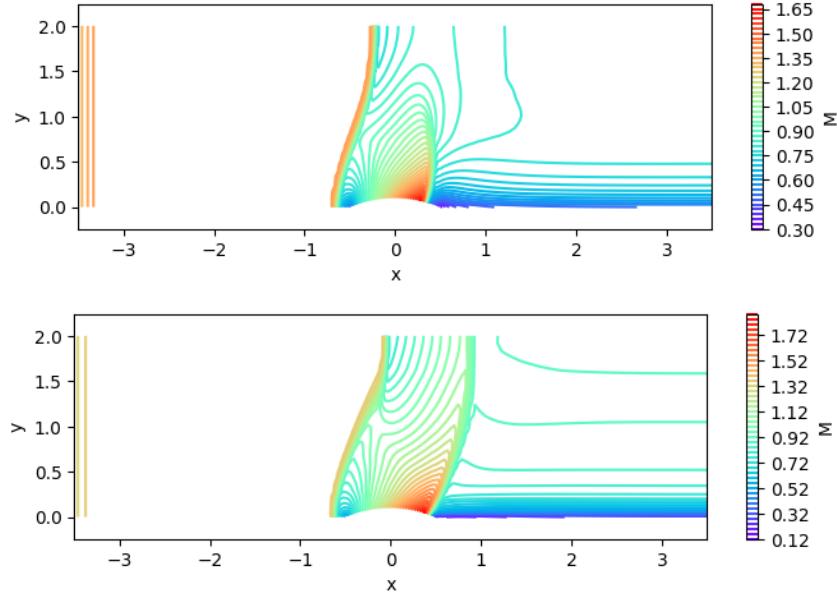


Figure 6: Mach number contours at $M=1.4$ computed using AUSM (top) and Roe (bottom) schemes

The mismatch in the shock shape for $M=1.4$ is clearly visible in the Mach contour as well. The velocity streamlines have been plotted in figures 7 and 8:

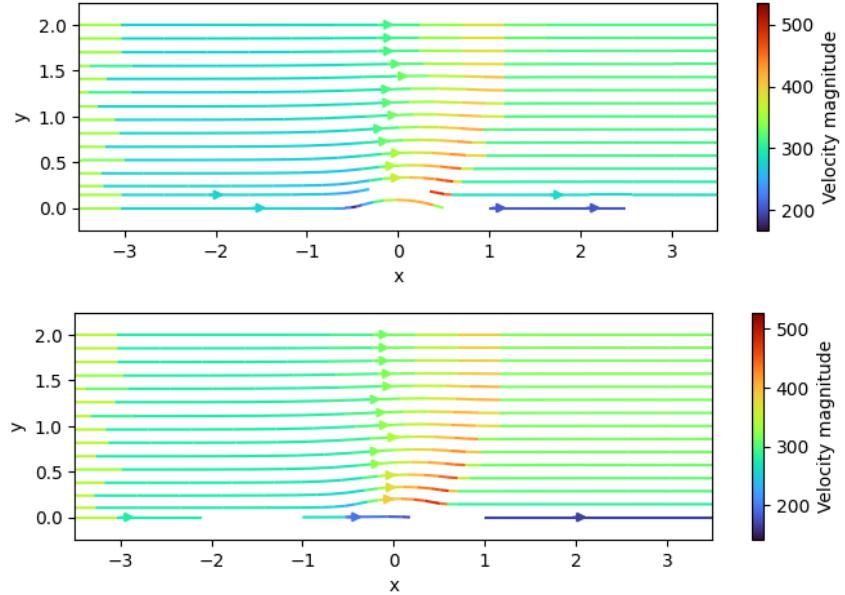


Figure 7: Velocity streamlines at $M=0.8$ computed using AUSM (top) and Roe (bottom) schemes

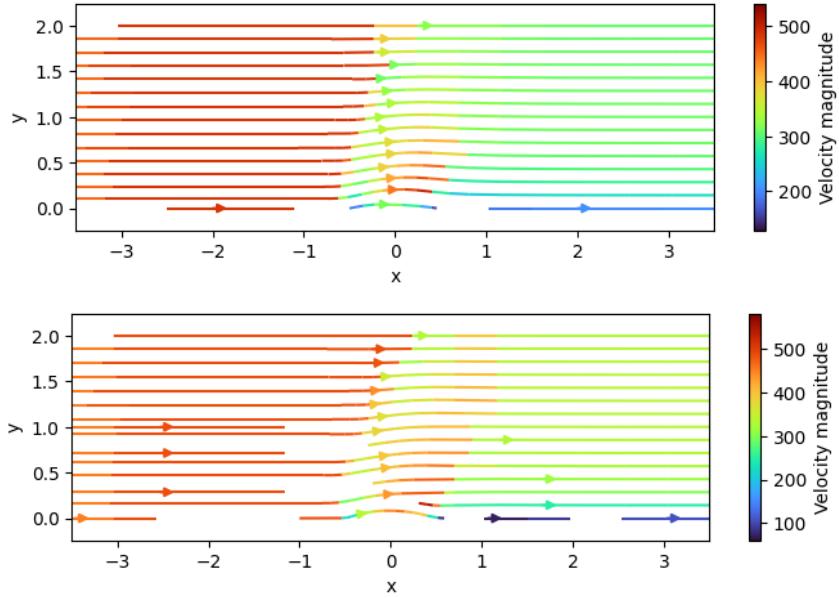


Figure 8: Velocity streamlines at $M=1.4$ computed using AUSM (top) and Roe (bottom) schemes

The stream plots of velocity corroborates with the physical flow features that are expected at the respective Mach numbers.

3.2 Residual convergence

The residual convergence plots for the 2 schemes have been compared at $M=1.4$ (supersonic) in figure 9:

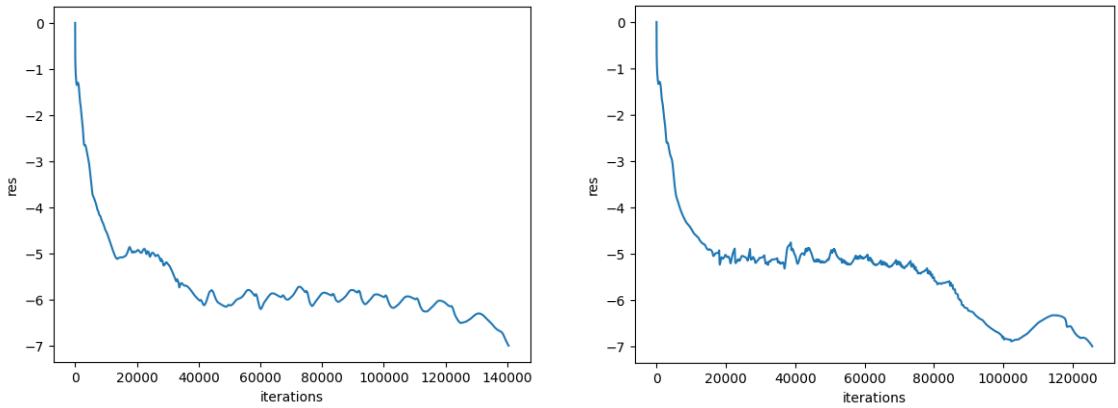


Figure 9: The residual convergence plots for the AUSM scheme (left) and the Roe scheme (right) at $M=1.4$

Although it is known for AUSM to be faster than the Roe scheme, but here we see that Roe is giving a slightly faster convergence although the difference is quite less.

3.3 Higher order reconstructed case

The MUSCL technique was applied to the transonic flow case and the results are shown below in figure 10. Its residual convergence plot (converged till $1e-12$) is shown in figure 11.

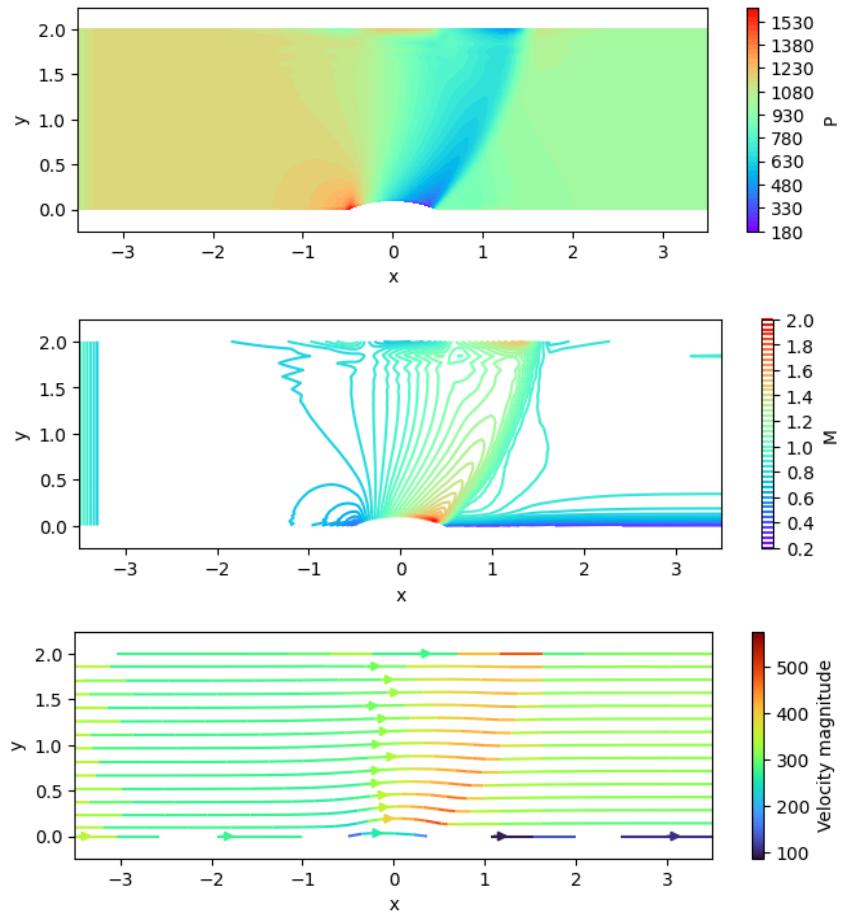


Figure 10: Pressure contours (top), Mach contours (middle) and velocity streamlines (bottom) at $M=0.8$ computed using AUSM with MUSCL reconstruction

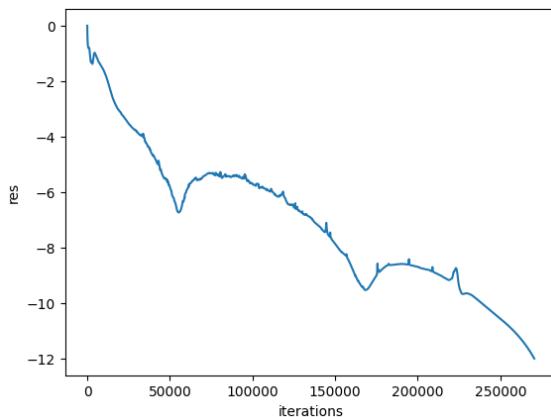


Figure 11: Residual convergence for $M=0.8$ with MUSCL reconstruction

The higher order reconstruction solutions are giving an oscillatory result, in spite of using a limiter.

4 Conclusion

For the linear reconstruction of the states, Roe seems to be giving more accurate results than AUSM, which is quite evident in the M=1.4 (supersonic) case. But due to their highly dissipative nature, both the schemes converge to non-oscillatory stable results. But the flow features like shocks are quite smeared out. In the case of higher-order reconstructed states, as the discontinuities at the cell interfaces are lower, the dissipative property of the scheme is lower as well and therefore the results are oscillatory. But the convergence obtained was much faster as it reached $1e - 7$ at around 150,000 iterations against 270,000 for regular AUSM.

Appendix

The code snippets have been added here for the readers' reference.

Listing 1: Initialization

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3 file1 = 'meshfile.txt'
4
5 T_inf = 300
6 R_air = 287
7 M_inf = 1.4
8 p_stat = 1000                                # Freestream static pressure is
                                                given to be 0.01 atm ~ 1e3 Pa
9 gamma = 1.4
10 rho_inf = p_stat/(R_air*T_inf)
11 a_inf = np.sqrt(gamma*p_stat/rho_inf)
12 u_inf = M_inf*a_inf
13 v_inf = 0
14 cfl = 0.5
15 tol = 1e-7
16 if M_inf < 1:
17     mode = 'sub'
18 else:
19     mode = 'sup'
20 restart = False
21 if restart:
22     last_itr = 130000
23     last_time = 0.6938204408751635
24 high_recon = False

```

Listing 2: Meshing

```

1 with open(file1, 'r') as meshfile:
2     sortlines = []
3     lines = meshfile.readlines()
4     for line in lines:
5         line = line.strip()
6         strnum = line.split()
7         fnum = [float(x) for x in strnum]
8         sortlines.append(fnum)
9 dims = sortlines[0]
10 dim = [int(x) for x in dims]
11 dim = np.array(dim)
12

```

```

13 nodes = sortlines[1:-1]
14
15 cellsdef = []
16 for i in range(dim[0]*(dim[1]-1)):
17     if (i + 1) % 97 == 0:
18         continue
19     cellsdef.append([i+1, i+1+dim[0], i+dim[0], i])
20
21 cells = np.zeros((dim[0]-1,dim[1]-1), dtype=int)
22 cellind = 0
23 for j in range(dim[1]-1):
24     for i in range(dim[0]-1):
25         cells[i,j] = int(cellind)
26         cellind += 1
27
28 facenormals = []
29 for cell in cellsdef:
30     nE = np.array(((nodes[cell[1]][1]-nodes[cell[0]][1]),-(nodes[cell
31         [1]][0]-nodes[cell[0]][0])))
32     nN = np.array(((nodes[cell[2]][1]-nodes[cell[1]][1]),-(nodes[cell
33         [2]][0]-nodes[cell[1]][0])))
34     nW = np.array(((nodes[cell[3]][1]-nodes[cell[2]][1]),-(nodes[cell
35         [3]][0]-nodes[cell[2]][0])))
36     nS = np.array(((nodes[cell[0]][1]-nodes[cell[3]][1]),-(nodes[cell
37         [0]][0]-nodes[cell[3]][0])))
38     facenormals.append([nE, nN, nW, nS])
39
40 faceareas = []
41 for cell in cellsdef:
42     p0 = np.array(nodes[cell[0]])
43     p1 = np.array(nodes[cell[1]])
44     p2 = np.array(nodes[cell[2]])
45     p3 = np.array(nodes[cell[3]])
46     aE = np.linalg.norm(p1-p0)
47     aN = np.linalg.norm(p2-p1)
48     aW = np.linalg.norm(p3-p2)
49     aS = np.linalg.norm(p0-p3)
50     faceareas.append([aE, aN, aW, aS])
51
52 cellvols = []
53 for cell in cellsdef:
54     p0 = np.array(nodes[cell[0]])
55     p1 = np.array(nodes[cell[1]])
56     p2 = np.array(nodes[cell[2]])
57     p3 = np.array(nodes[cell[3]])
58     a1 = 0.5 * np.abs(p0[0]*(p1[1] - p2[1]) + p1[0]*(p2[1] - p0[1]) +
59     p2[0]*(p0[1] - p1[1]))
60     a2 = 0.5 * np.abs(p0[0]*(p2[1] - p3[1]) + p2[0]*(p3[1] - p0[1]) +
61     p3[0]*(p0[1] - p2[1]))
62     v = a1+a2
63     cellvols.append(v)
64
65 xfacesN = np.zeros((dim[0],dim[1]-1, 2))
66 for j in range(dim[1]-1):
67     for i in range(dim[0]):
68         if i == 0:

```

```

65         xfacesN[i,j] = -facenormals[cells[i,j]][2]
66         continue
67     xfacesN[i,j] = facenormals[cells[i-1,j]][0]
68
69 yfacesN = np.zeros((dim[0]-1,dim[1], 2))
70 for i in range(dim[0]-1):
71     for j in range(dim[1]):
72         if j == 0:
73             yfacesN[i,j] = -facenormals[cells[i,j]][3]
74             continue
75         yfacesN[i,j] = facenormals[cells[i,j-1]][1]
76
77 xfacesA = np.zeros((dim[0],dim[1]-1))
78 for j in range(dim[1]-1):
79     for i in range(dim[0]):
80         if i == 0:
81             xfacesA[i,j] = faceareas[cells[i,j]][2]
82             continue
83         xfacesA[i,j] = faceareas[cells[i-1,j]][0]
84
85 yfacesA = np.zeros((dim[0]-1,dim[1]))
86 for i in range(dim[0]-1):
87     for j in range(dim[1]):
88         if j == 0:
89             yfacesA[i,j] = faceareas[cells[i,j]][3]
90             continue
91         yfacesA[i,j] = faceareas[cells[i,j-1]][1]
92
93 vols_2d = np.reshape(cellvols, (96,48), order='C')

```

Listing 3: Boundary conditions

```

1 # Boundary Conditions
2
3 def in_bc(dim, r, mode):
4     if mode == 'sub':
5         rho_l = rho_inf
6         u_l = u_inf
7         v_l = v_inf
8         p_l = r[3, 0, :]
9
10    if mode == 'sup':
11        rho_l = rho_inf
12        u_l = u_inf
13        v_l = v_inf
14        p_l = p_stat
15
16    return rho_l, u_l, v_l, p_l
17
18 def out_bc(dim, r, mode):
19    if mode == 'sub':
20        rho_r = r[0, dim[0]-2, :]
21        u_r = r[1, dim[0]-2, :]
22        v_r = r[2, dim[0]-2, :]
23        p_r = p_stat
24
25    if mode == 'sup':
26        rho_r = r[0, dim[0]-2, :]

```

```

27         u_r = r[1, dim[0]-2, :]
28         v_r = r[2, dim[0]-2, :]
29         p_r = r[3, dim[0]-2, :]
30
31     return rho_r, u_r, v_r, p_r
32
33 def slipwall(dim, r, yfacesA, yfacesN):
34     rho_d = r[0, :, 0]
35     nx = yfacesN[:, 0, 0]/yfacesA[:, 0]
36     ny = yfacesN[:, 0, 1]/yfacesA[:, 0]
37     u_d = (r[1, :, 0]*(ny**2-nx**2)-2*r[2, :, 0]*nx*ny)/(nx**2+ny**2)
38     v_d = (r[2, :, 0]*(nx**2-ny**2)-2*r[1, :, 0]*nx*ny)/(nx**2+ny**2)
39     p_d = r[3, :, 0]
40     rho_u = r[0, :, dim[1]-2]
41     p_u = r[3, :, dim[1]-2]
42
43     return rho_d, p_d, rho_u, p_u, u_d, v_d

```

Listing 4: Roe flux

```

1 def fluxroe(dim, r):
2
3
4     # Fluxes initialization
5     xFluxes = np.zeros((dim[0], dim[1]-1, 4))
6     yFluxes = np.zeros((dim[0]-1, dim[1], 4))
7
8     # X-face fluxes
9     for i in range(1, dim[0]-1):
10         area = xfacesA[i, :]
11         nx = xfacesN[i, :, 0]/area
12         ny = xfacesN[i, :, 1]/area
13         # Left states
14         rho_l = r[0, i-1, :]
15         u_l = r[1, i-1, :]
16         v_l = r[2, i-1, :]
17         p_l = r[3, i-1, :]
18         h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l
19             **2)
20         u_nl = u_l*nx + v_l*ny
21         vel_l = np.sqrt(u_l**2 + v_l**2)
22         # Right states
23         rho_r = r[0, i, :]
24         u_r = r[1, i, :]
25         v_r = r[2, i, :]
26         p_r = r[3, i, :]
27         h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r
28             **2)
29         u_nr = u_r*nx + v_r*ny
30         vel_r = np.sqrt(u_r**2 + v_r**2)
31
32         # Roe states
33         r_factor = np.sqrt(rho_r/rho_l)
34         rho_roe = np.sqrt(rho_r*rho_l)
35         u_roe = (u_l + u_r*r_factor)/(1+r_factor)
36         v_roe = (v_l + v_r*r_factor)/(1+r_factor)
37         vel_roe = np.sqrt(u_roe**2 + v_roe**2)
38         h0_roe = (h0_l + h0_r*r_factor)/(1+r_factor)

```

```

37     a_roe = np.sqrt((gamma-1.0)*(h0_roe-0.5*(vel_roe**2)))
38     un_roe = u_roe*nx + v_roe*ny
39
40     # Dissipation using Roe states
41
42     alpha0 = area*np.abs(un_roe)*((rho_r-rho_l)-(p_r-p_l)/a_roe**2)
43     alpha1 = (area/(2*a_roe**2))*np.abs(un_roe+a_roe)*((p_r-p_l)+(
44         rho_roe*a_roe*(u_nr-u_nl)))
45     alpha2 = (area/(2*a_roe**2))*np.abs(un_roe-a_roe)*((p_r-p_l)-(
46         rho_roe*a_roe*(u_nr-u_nl)))
47     alpha3 = alpha0 + alpha1 + alpha2
48     alpha4 = a_roe*(alpha1-alpha2)
49     alpha5 = area*np.abs(un_roe)*((rho_roe*(u_r-u_l))-(nx*rho_roe*(
50         u_nr-u_nl)))
51     alpha6 = area*np.abs(un_roe)*((rho_roe*(v_r-v_l))-(ny*rho_roe*(
52         u_nr-u_nl)))
53
54     diss0 = alpha3
55     diss1 = u_roe*alpha3 + nx*alpha4 + alpha5
56     diss2 = v_roe*alpha3 + ny*alpha4 + alpha6
57     diss3 = h0_roe*(alpha3-alpha0) + un_roe*alpha4 + u_roe*alpha5 +
58             v_roe*alpha6 + 0.5*(u_roe**2+v_roe**2)*alpha0
59
60     fm_pl = 0.5*area*rho_l*u_nl
61     fm_mi = 0.5*area*rho_r*u_nr
62
63     xFluxes[i,:,:0] = (fm_pl + fm_mi) - 0.5*diss0
64     xFluxes[i,:,:1] = (fm_pl*u_l + fm_mi*u_r) + 0.5*(p_l+p_r)*nx*
65         area - 0.5*diss1
66     xFluxes[i,:,:2] = (fm_pl*v_l + fm_mi*v_r) + 0.5*(p_l+p_r)*ny*
67         area - 0.5*diss2
68     xFluxes[i,:,:3] = (fm_pl*h0_l + fm_mi*h0_r) - 0.5*diss3
69
70     # Left boundary flux
71     area = xfacesA[0,:]
72     nx = xfacesN[0,:,:0]/area
73     ny = xfacesN[0,:,:1]/area
74
75     # Left states
76     bc_l = in_bc(dim, r, mode)
77     rho_l = bc_l[0]
78     u_l = bc_l[1]
79     v_l = bc_l[2]
80     p_l = bc_l[3]
81     h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l**2)
82     u_nl = u_l*nx + v_l*ny
83     vel_l = np.sqrt(u_l**2 + v_l**2)
84
85     # Right states
86     rho_r = r[0, 0, :]
87     u_r = r[1, 0, :]
88     v_r = r[2, 0, :]
89     p_r = r[3, 0, :]
90     h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r**2)
91     u_nr = u_r*nx + v_r*ny
92     vel_r = np.sqrt(u_r**2 + v_r**2)
93
94     # Roe states
95     r_factor = np.sqrt(rho_r/rho_l)
96     rho_roe = np.sqrt(rho_r*rho_l)

```

```

88     u_roe = (u_l + u_r*r_factor)/(1+r_factor)
89     v_roe = (v_l + v_r*r_factor)/(1+r_factor)
90     vel_roe = np.sqrt(u_roe**2 + v_roe**2)
91     h0_roe = (h0_l + h0_r*r_factor)/(1+r_factor)
92     a_roe = np.sqrt((gamma-1.0)*(h0_roe-0.5*(vel_roe**2)))
93     un_roe = u_roe*nx + v_roe*ny
94
95     # Dissipation using Roe states
96
97     alpha0 = area*np.abs(un_roe)*((rho_r-rho_l)-(p_r-p_l)/a_roe**2)
98     alpha1 = (area/(2*a_roe**2))*np.abs(un_roe+a_roe)*((p_r-p_l)+(
99         rho_roe*a_roe*(u_nr-u_nl)))
100    alpha2 = (area/(2*a_roe**2))*np.abs(un_roe-a_roe)*((p_r-p_l)-(
101        rho_roe*a_roe*(u_nr-u_nl)))
102    alpha3 = alpha0 + alpha1 + alpha2
103    alpha4 = a_roe*(alpha1-alpha2)
104    alpha5 = area*np.abs(un_roe)*((rho_roe*(u_r-u_l))-(nx*rho_roe*(u_nr-
105        -u_nl)))
106    alpha6 = area*np.abs(un_roe)*((rho_roe*(v_r-v_l))-(ny*rho_roe*(u_nr-
107        -u_nl)))
108
109    diss0 = alpha3
110    diss1 = u_roe*alpha3 + nx*alpha4 + alpha5
111    diss2 = v_roe*alpha3 + ny*alpha4 + alpha6
112    diss3 = h0_roe*(alpha3-alpha0) + un_roe*alpha4 + u_roe*alpha5 +
113        v_roe*alpha6 + 0.5*(u_roe**2+v_roe**2)*alpha0
114
115    fm_pl = 0.5*area*rho_l*u_nl
116    fm_mi = 0.5*area*rho_r*u_nr
117
118    xFluxes[0,:,:0] = (fm_pl + fm_mi) - 0.5*diss0
119    xFluxes[0,:,:1] = (fm_pl*u_l + fm_mi*u_r) + 0.5*(p_l+p_r)*nx*area -
120        0.5*diss1
121    xFluxes[0,:,:2] = (fm_pl*v_l + fm_mi*v_r) + 0.5*(p_l+p_r)*ny*area -
122        0.5*diss2
123    xFluxes[0,:,:3] = (fm_pl*h0_l + fm_mi*h0_r) - 0.5*diss3
124
125    # Right boundary flux
126    area = xfacesA[dim[0]-1,:]
127    nx = xfacesN[dim[0]-1,:,:0]/area
128    ny = xfacesN[dim[0]-1,:,:1]/area
129
130    # Left states
131    rho_l = r[0, dim[0]-2, :]
132    u_l = r[1, dim[0]-2, :]
133    v_l = r[2, dim[0]-2, :]
134    p_l = r[3, dim[0]-2, :]
135    h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l**2)
136    u_nl = u_l*nx + v_l*ny
137    vel_l = np.sqrt(u_l**2 + v_l**2)
138
139    # Right states
140    bc_r = out_bc(dim, r, mode)
141    rho_r = bc_r[0]
142    u_r = bc_r[1]
143    v_r = bc_r[2]
144    p_r = bc_r[3]
145    h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r**2)
146    u_nr = u_r*nx + v_r*ny
147    vel_r = np.sqrt(u_r**2 + v_r**2)

```

```

139
140     # Roe states
141     r_factor = np.sqrt(rho_r/rho_l)
142     rho_roe = np.sqrt(rho_r*rho_l)
143     u_roe = (u_l + u_r*r_factor)/(1+r_factor)
144     v_roe = (v_l + v_r*r_factor)/(1+r_factor)
145     vel_roe = np.sqrt(u_roe**2 + v_roe**2)
146     h0_roe = (h0_l + h0_r*r_factor)/(1+r_factor)
147     a_roe = np.sqrt((gamma-1.0)*(h0_roe-0.5*(vel_roe**2)))
148     un_roe = u_roe*nx + v_roe*ny
149
150     # Dissipation using Roe states
151
152     alpha0 = area*np.abs(un_roe)*((rho_r-rho_l)-(p_r-p_l)/a_roe**2)
153     alpha1 = (area/(2*a_roe**2))*np.abs(un_roe+a_roe)*((p_r-p_l)+(
154         rho_roe*a_roe*(u_nr-u_nl)))
155     alpha2 = (area/(2*a_roe**2))*np.abs(un_roe-a_roe)*((p_r-p_l)-(
156         rho_roe*a_roe*(u_nr-u_nl)))
157     alpha3 = alpha0 + alpha1 + alpha2
158     alpha4 = a_roe*(alpha1-alpha2)
159     alpha5 = area*np.abs(un_roe)*((rho_roe*(u_r-u_l))-(nx*rho_roe*(u_nr
160         -u_nl)))
161     alpha6 = area*np.abs(un_roe)*((rho_roe*(v_r-v_l))-(ny*rho_roe*(u_nr
162         -u_nl)))
163
163     diss0 = alpha3
164     diss1 = u_roe*alpha3 + nx*alpha4 + alpha5
165     diss2 = v_roe*alpha3 + ny*alpha4 + alpha6
166     diss3 = h0_roe*(alpha3-alpha0) + un_roe*alpha4 + u_roe*alpha5 +
167         v_roe*alpha6 + 0.5*(u_roe**2+v_roe**2)*alpha0
168
168     fm_pl = 0.5*area*rho_l*u_nl
169     fm_mi = 0.5*area*rho_r*u_nr
170
171     xFluxes[dim[0]-1,:,:0] = (fm_pl + fm_mi) - 0.5*diss0
172     xFluxes[dim[0]-1,:,:1] = (fm_pl*u_l + fm_mi*u_r) + 0.5*(p_l+p_r)*nx*
173         area - 0.5*diss1
174     xFluxes[dim[0]-1,:,:2] = (fm_pl*v_l + fm_mi*v_r) + 0.5*(p_l+p_r)*ny*
175         area - 0.5*diss2
176     xFluxes[dim[0]-1,:,:3] = (fm_pl*h0_l + fm_mi*h0_r) - 0.5*diss3
177
177     # Y-face fluxes
178     for j in range(1, dim[1]-1):
179         area = yfacesA[:,j]
180         nx = yfacesN[:,j,0]/area
181         ny = yfacesN[:,j,1]/area
182
183         # Lower states
184         rho_d = r[0, :, j-1]
185         u_d = r[1, :, j-1]
186         v_d = r[2, :, j-1]
187         p_d = r[3, :, j-1]
188         h0_d = (gamma/(gamma - 1))*(p_d/rho_d) + 0.5 * (u_d**2 + v_d
189             **2)
190         u_nd = u_d*nx + v_d*ny
191         vel_d = np.sqrt(u_d**2 + v_d**2)
192
193         # Upper states
194         rho_u = r[0, :, j]
195         u_u = r[1, :, j]

```

```

189     v_u = r[2, :, j]
190     p_u = r[3, :, j]
191     h0_u = (gamma/(gamma - 1))*(p_u/rho_u) + 0.5 * (u_u**2 + v_u
192         **2)
193     u_nu = u_u*nx + v_u*ny
194     vel_u = np.sqrt(u_u**2 + v_u**2)

195     # Roe states
196     r_factor = np.sqrt(rho_u/rho_d)
197     rho_roe = np.sqrt(rho_u*rho_d)
198     u_roe = (u_d + u_u*r_factor)/(1+r_factor)
199     v_roe = (v_d + v_u*r_factor)/(1+r_factor)
200     vel_roe = np.sqrt(u_roe**2 + v_roe**2)
201     h0_roe = (h0_d + h0_u*r_factor)/(1+r_factor)
202     a_roe = np.sqrt((gamma-1.0)*(h0_roe-0.5*(vel_roe**2)))
203     un_roe = u_roe*nx + v_roe*ny

204     # Dissipation using Roe states
205
206     alpha0 = area*np.abs(un_roe)*((rho_u-rho_d)-(p_u-p_d)/a_roe**2)
207     alpha1 = (area/(2*a_roe**2))*np.abs(un_roe+a_roe)*((p_u-p_d)+(
208         rho_roe*a_roe*(u_nu-u_nd)))
209     alpha2 = (area/(2*a_roe**2))*np.abs(un_roe-a_roe)*((p_u-p_d)-(
210         rho_roe*a_roe*(u_nu-u_nd)))
211     alpha3 = alpha0 + alpha1 + alpha2
212     alpha4 = a_roe*(alpha1-alpha2)
213     alpha5 = area*np.abs(un_roe)*((rho_roe*(u_u-u_d))-(nx*rho_roe*(
214         u_nu-u_nd)))
215     alpha6 = area*np.abs(un_roe)*((rho_roe*(v_u-v_d))-(ny*rho_roe*(
216         u_nu-u_nd)))

217     diss0 = alpha3
218     diss1 = u_roe*alpha3 + nx*alpha4 + alpha5
219     diss2 = v_roe*alpha3 + ny*alpha4 + alpha6
220     diss3 = h0_roe*(alpha3-alpha0) + un_roe*alpha4 + u_roe*alpha5 +
221         v_roe*alpha6 + 0.5*(u_roe**2+v_roe**2)*alpha0

222     fm_pl = 0.5*area*rho_d*u_nd
223     fm_mi = 0.5*area*rho_u*u_nu

224     yFluxes[:,j,0] = (fm_pl + fm_mi) - 0.5*diss0
225     yFluxes[:,j,1] = (fm_pl*u_d + fm_mi*u_u) + 0.5*(p_d+p_u)*nx*
226         area - 0.5*diss1
227     yFluxes[:,j,2] = (fm_pl*v_d + fm_mi*v_u) + 0.5*(p_d+p_u)*ny*
228         area - 0.5*diss2
229     yFluxes[:,j,3] = (fm_pl*h0_d + fm_mi*h0_u) - 0.5*diss3

230     bc_w = slipwall(dim,r,yfacesA,yfacesN)

231     # Lower boundary flux
232     area = yfacesA[:,0]
233     nx = yfacesN[:,0,0]/area
234     ny = yfacesN[:,0,1]/area
235     # Lower states
236     rho_d = bc_w[0]
237     u_d = r[1, :, 0]
238     v_d = -r[2, :, 0]
239     p_d = bc_w[1]

```

```

239     h0_d = (gamma/(gamma - 1))*(p_d/rho_d) + 0.5 * (u_d**2 + v_d**2)
240     u_nd = u_d*nx + v_d*ny
241     vel_d = np.sqrt(u_d**2 + v_d**2)
242     # Upper states
243     rho_u = r[0, :, 0]
244     u_u = r[1, :, 0]
245     v_u = r[2, :, 0]
246     p_u = r[3, :, 0]
247     h0_u = (gamma/(gamma - 1))*(p_u/rho_u) + 0.5 * (u_u**2 + v_u**2)
248     u_nu = u_u*nx + v_u*ny
249     vel_u = np.sqrt(u_u**2 + v_u**2)
250
251     # Roe states not needed at wall
252
253     fm_pl = 0
254     fm_mi = 0
255
256     yFluxes[:, 0, 0] = (fm_pl + fm_mi)
257     yFluxes[:, 0, 1] = (fm_pl*u_d + fm_mi*u_u) + 0.5*(p_d+p_u)*nx*area
258     yFluxes[:, 0, 2] = (fm_pl*v_d + fm_mi*v_u) + 0.5*(p_d+p_u)*ny*area
259     yFluxes[:, 0, 3] = (fm_pl*h0_d + fm_mi*h0_u)
260
261     # Upper boundary flux
262     area = yfacesA[:, dim[1]-1]
263     nx = yfacesN[:, dim[1]-1, 0]/area
264     ny = yfacesN[:, dim[1]-1, 1]/area
265     # Lower states
266     rho_d = r[0, :, dim[1]-2]
267     u_d = r[1, :, dim[1]-2]
268     v_d = r[2, :, dim[1]-2]
269     p_d = r[3, :, dim[1]-2]
270     h0_d = (gamma/(gamma - 1))*(p_d/rho_d) + 0.5 * (u_d**2 + v_d**2)
271     u_nd = u_d*nx + v_d*ny
272     vel_d = np.sqrt(u_d**2 + v_d**2)
273     # Upper states
274     rho_u = bc_w[2]
275     u_u = r[1, :, dim[1]-2]
276     v_u = -r[2, :, dim[1]-2]
277     p_u = bc_w[3]
278     h0_u = (gamma/(gamma - 1))*(p_u/rho_u) + 0.5 * (u_u**2 + v_u**2)
279     u_nu = u_u*nx + v_u*ny
280     vel_u = np.sqrt(u_u**2 + v_u**2)
281
282     # Roe states not needed at wall
283
284     fm_pl = 0
285     fm_mi = 0
286
287     yFluxes[:, dim[1]-1, 0] = (fm_pl + fm_mi)
288     yFluxes[:, dim[1]-1, 1] = (fm_pl*u_d + fm_mi*u_u) + 0.5*(p_d+p_u)*nx*
289         area
290     yFluxes[:, dim[1]-1, 2] = (fm_pl*v_d + fm_mi*v_u) + 0.5*(p_d+p_u)*ny*
291         area
292     yFluxes[:, dim[1]-1, 3] = (fm_pl*h0_d + fm_mi*h0_u)
293
294     res = ((xFluxes[1:, :, :] - xFluxes[:-1, :, :]) + (yFluxes[:, 1:, :]
295         :] - yFluxes[:, :-1, :]))

```

```
294     return res
```

Listing 5: AUSM flux

```

1 def fluxls(dim, r):
2
3     # Fluxes initialization
4     xFluxes = np.zeros((dim[0], dim[1]-1, 4))
5     yFluxes = np.zeros((dim[0]-1, dim[1], 4))
6
7     # X-face fluxes
8     for i in range(1, dim[0]-1):
9         area = xfacesA[i,:]
10        nx = xfacesN[i,:,0]/area
11        ny = xfacesN[i,:,1]/area
12        # Left states
13        rho_l = r[0, i-1, :]
14        u_l = r[1, i-1, :]
15        v_l = r[2, i-1, :]
16        p_l = r[3, i-1, :]
17        h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l
18            **2)
19        a_l = np.sqrt(gamma*(p_l/rho_l))
20        u_nl = u_l*nx + v_l*ny
21        # Right states
22        rho_r = r[0, i, :]
23        u_r = r[1, i, :]
24        v_r = r[2, i, :]
25        p_r = r[3, i, :]
26        h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r
27            **2)
28        a_r = np.sqrt(gamma*(p_r/rho_r))
29        u_nr = u_r*nx + v_r*ny
30
31        a_avg = 0.5*(a_l + a_r)
32        xma_l = u_nl/a_avg
33        xma_r = u_nr/a_avg
34
35        al_l = 0.5*(1.0 + np.sign((xma_l)))
36        al_r = 0.5*(1.0 - np.sign((xma_r)))
37
38        be_l = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_l)))
39        be_r = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_r)))
40
41        # Defining the van Leer polynomials
42        d_pl = 0.25*((xma_l+1)**2)*(2-xma_l)
43        d_mi = 0.25*((xma_r-1)**2)*(2+xma_r)
44
45        d_l = al_l*(1+be_l) - be_l*d_pl
46        d_r = al_r*(1+be_r) - be_r*d_mi
47
48        cvl_pl = al_l * (1+be_l) * xma_l - be_l * 0.25 * (1+xma_l)**2
49        cvl_mi = al_r * (1+be_r) * xma_r + be_r * 0.25 * (1-xma_r)**2
50
51        # Using the van Leer polynomials to define the Liou-Stefan
52        # polynomials
53        cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
54        cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
```

```

52
53     # Defining the mass-flux using the LS formulation
54     fm_pl = area * a_avg * cls_pl * rho_l
55     fm_mi = area * a_avg * cls_mi * rho_r
56     xFluxes[i,:, 0] = fm_pl + fm_mi
57     xFluxes[i,:, 1] = fm_pl * u_l + fm_mi * u_r + (d_l*p_l + d_r*
58         p_r) * nx*area
59     xFluxes[i,:, 2] = fm_pl * v_l + fm_mi * v_r + (d_l*p_l + d_r*
60         p_r) * ny*area
61     xFluxes[i,:, 3] = fm_pl * h0_l + fm_mi * h0_r
62
63     # Left boundary flux
64     area = xfacesA[0,:]
65     nx = xfacesN[0,:,0]/area
66     ny = xfacesN[0,:,1]/area
67     # Left states
68     bc_l = in_bc(dim, r, mode)
69     rho_l = bc_l[0]
70     u_l = bc_l[1]
71     v_l = bc_l[2]
72     p_l = bc_l[3]
73     h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l**2)
74     a_l = np.sqrt(gamma*(p_l/rho_l))
75     u_nl = u_l*nx + v_l*ny
76     # Right states
77     rho_r = r[0, 0, :]
78     u_r = r[1, 0, :]
79     v_r = r[2, 0, :]
80     p_r = r[3, 0, :]
81     h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r**2)
82     a_r = np.sqrt(gamma*(p_r/rho_r))
83     u_nr = u_r*nx + v_r*ny
84
85     a_avg = 0.5*(a_l + a_r)
86     xma_l = u_nl/a_avg
87     xma_r = u_nr/a_avg
88
89     al_l = 0.5*(1.0 + np.sign((xma_l)))
90     al_r = 0.5*(1.0 - np.sign((xma_r)))
91
92     # Defining the van Leer polynomials
93     d_pl = 0.25*((xma_l+1)**2)*(2-xma_l)
94     d_mi = 0.25*((xma_r-1)**2)*(2+xma_r)
95
96     d_l = al_l*(1+be_l) - be_l*d_pl
97     d_r = al_r*(1+be_r) - be_r*d_mi
98
99     cvl_pl = al_l * (1+be_l) * xma_l - be_l * 0.25 * (1+xma_l)**2
100    cvl_mi = al_r * (1+be_r) * xma_r + be_r * 0.25 * (1-xma_r)**2
101
102    # Using the van Leer polynomials to define the Liou-Stefan
103    # polynomials
104    cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
105    cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
106

```

```

107 # Defining the mass-flux using the LS formulation
108 fm_pl = area * a_avg * cls_pl * rho_l
109 fm_mi = area * a_avg * cls_mi * rho_r
110 xFluxes[0,:, 0] = fm_pl + fm_mi
111 xFluxes[0,:, 1] = fm_pl * u_l + fm_mi * u_r + (d_l*p_l + d_r*p_r) *
    nx*area
112 xFluxes[0,:, 2] = fm_pl * v_l + fm_mi * v_r + (d_l*p_l + d_r*p_r) *
    ny*area
113 xFluxes[0,:, 3] = fm_pl * h0_l + fm_mi * h0_r
114
115 # Right boundary flux
116 area = xfacesA[dim[0]-1,:]
117 nx = xfacesN[dim[0]-1,:,:0]/area
118 ny = xfacesN[dim[0]-1,:,:1]/area
119 # Left states
120 rho_l = r[0, dim[0]-2, :]
121 u_l = r[1, dim[0]-2, :]
122 v_l = r[2, dim[0]-2, :]
123 p_l = r[3, dim[0]-2, :]
124 h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l**2)
125 a_l = np.sqrt(gamma*(p_l/rho_l))
126 u_nl = u_l*nx + v_l*ny
127 # Right states
128 bc_r = out_bc(dim, r, mode)
129 rho_r = bc_r[0]
130 u_r = bc_r[1]
131 v_r = bc_r[2]
132 p_r = bc_r[3]
133 h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r**2)
134 a_r = np.sqrt(gamma*(p_r/rho_r))
135 u_nr = u_r*nx + v_r*ny
136
137 a_avg = 0.5*(a_l + a_r)
138 xma_l = u_nl/a_avg
139 xma_r = u_nr/a_avg
140
141 al_l = 0.5*(1.0 + np.sign((xma_l)))
142 al_r = 0.5*(1.0 - np.sign((xma_r)))
143
144 be_l = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_l)))
145 be_r = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_r)))
146
147 # Defining the van Leer polynomials
148 d_pl = 0.25*((xma_l+1)**2)*(2-xma_l)
149 d_mi = 0.25*((xma_r-1)**2)*(2+xma_r)
150
151 d_l = al_l*(1+be_l) - be_l*d_pl
152 d_r = al_r*(1+be_r) - be_r*d_mi
153
154 cvl_pl = al_l * (1+be_l) * xma_l - be_l * 0.25 * (1+xma_l)**2
155 cvl_mi = al_r * (1+be_r) * xma_r + be_r * 0.25 * (1-xma_r)**2
156
157 # Using the van Leer polynomials to define the Liou-Stefan
158 # polynomials
159 cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
160 cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
161
162 # Defining the mass-flux using the LS formulation

```

```

162     fm_pl = area * a_avg * cls_pl * rho_l
163     fm_mi = area * a_avg * cls_mi * rho_r
164     xFluxes[dim[0]-1,:, 0] = fm_pl + fm_mi
165     xFluxes[dim[0]-1,:, 1] = fm_pl * u_l + fm_mi * u_r + (d_l*p_l + d_r
166             *p_r) * nx*area
166     xFluxes[dim[0]-1,:, 2] = fm_pl * v_l + fm_mi * v_r + (d_l*p_l + d_r
167             *p_r) * ny*area
167     xFluxes[dim[0]-1,:, 3] = fm_pl * h0_l + fm_mi * h0_r
168
169     # Y-face fluxes
170     for j in range(1, dim[1]-1):
171         area = yfacesA[:,j]
172         nx = yfacesN[:,j,0]/area
173         ny = yfacesN[:,j,1]/area
174         # Lower states
175         rho_d = r[0, :, j-1]
176         u_d = r[1, :, j-1]
177         v_d = r[2, :, j-1]
178         p_d = r[3, :, j-1]
179         h0_d = (gamma/(gamma - 1))*(p_d/rho_d) + 0.5 * (u_d**2 + v_d
180             **2)
180         a_d = np.sqrt(gamma*(p_d/rho_d))
181         u_nd = u_d*nx + v_d*ny
182         # Upper states
183         rho_u = r[0, :, j]
184         u_u = r[1, :, j]
185         v_u = r[2, :, j]
186         p_u = r[3, :, j]
187         h0_u = (gamma/(gamma - 1))*(p_u/rho_u) + 0.5 * (u_u**2 + v_u
188             **2)
188         a_u = np.sqrt(gamma*(p_u/rho_u))
189         u_nu = u_u*nx + v_u*ny
190
191         a_avg = 0.5*(a_d + a_u)
192         xma_d = u_nd/a_avg
193         xma_u = u_nu/a_avg
194
195         al_d = 0.5*(1.0 + np.sign((xma_d)))
196         al_u = 0.5*(1.0 - np.sign((xma_u)))
197
198         be_d = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_d)))
199         be_u = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_u)))
200
201         # Defining the van Leer polynomials
202         d_pl = 0.25*((xma_d+1)**2)*(2-xma_d)
203         d_mi = 0.25*((xma_u-1)**2)*(2+xma_u)
204
205         d_d = al_d*(1+be_d) - be_d*d_pl
206         d_u = al_u*(1+be_u) - be_u*d_mi
207
208         cvl_pl = al_d * (1+be_d) * xma_d - be_d * 0.25 * (1+xma_d)**2
209         cvl_mi = al_u * (1+be_u) * xma_u + be_u * 0.25 * (1-xma_u)**2
210
211         # Using the van Leer polynomials to define the Liou-Stefan
212         # polynomials
213         cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
213         cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
214

```

```

215      # Defining the mass-flux using the LS formulation
216      fm_pl = area * a_avg * cls_pl * rho_d
217      fm_mi = area * a_avg * cls_mi * rho_u
218      yFluxes[:, j, 0] = fm_pl + fm_mi
219      yFluxes[:, j, 1] = fm_pl * u_d + fm_mi * u_u + (d_d*p_d + d_u*
220          p_u) * nx*area
221      yFluxes[:, j, 2] = fm_pl * v_d + fm_mi * v_u + (d_d*p_d + d_u*
222          p_u) * ny*area
223      yFluxes[:, j, 3] = fm_pl * h0_d + fm_mi * h0_u
224
225      bc_w = slipwall(dim,r,yfacesA,yfacesN)
226      # Lower boundary flux
227      area = yfacesA[:,0]
228      nx = yfacesN[:,0,0]/area
229      ny = yfacesN[:,0,1]/area
230      # Upper states
231      rho_u = r[0, :, 0]
232      u_u = r[1, :, 0]
233      v_u = r[2, :, 0]
234      p_u = r[3, :, 0]
235      h0_u = (gamma/(gamma - 1))*(p_u/rho_u) + 0.5 * (u_u**2 + v_u**2)
236      a_u = np.sqrt(gamma*(p_u/rho_u))
237      u_nu = u_u*nx + v_u*ny
238      # Lower state
239      rho_d = bc_w[0]
240      p_d = bc_w[1]
241      a_d = np.sqrt(gamma*(p_d/rho_d))
242      u_nd = -u_nu
243
244      a_avg = 0.5*(a_d + a_u)
245      xma_d = u_nd/a_avg
246      xma_u = u_nu/a_avg
247
248      al_d = 0.5*(1.0 + np.sign((xma_d)))
249      al_u = 0.5*(1.0 - np.sign((xma_u)))
250
251      # Defining the van Leer polynomials
252      d_pl = 0.25*((xma_d+1)**2)*(2-xma_d)
253      d_mi = 0.25*((xma_u-1)**2)*(2+xma_u)
254
255      d_d = al_d*(1+be_d) - be_d*d_pl
256      d_u = al_u*(1+be_u) - be_u*d_mi
257
258      cvl_pl = al_d * (1+be_d) * xma_d - be_d * 0.25 * (1+xma_d)**2
259      cvl_mi = al_u * (1+be_u) * xma_u + be_u * 0.25 * (1-xma_u)**2
260
261      # Using the van Leer polynomials to define the Liou-Stefan
262      # polynomials
263      cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
264      cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
265
266      # Defining the mass-flux using the LS formulation
267      fm_pl = 0
268      fm_mi = 0
269      yFluxes[:, 0, 0] = fm_pl + fm_mi

```

```

270     yFluxes[:,0, 1] = fm_pl * u_d + fm_mi * u_u + (d_d*p_d + d_u*p_u) *
271         nx*area
272     yFluxes[:,0, 2] = fm_pl * v_d + fm_mi * v_u + (d_d*p_d + d_u*p_u) *
273         ny*area
274     yFluxes[:,0, 3] = fm_pl * h0_d + fm_mi * h0_u
275
276     # Upper boundary flux
277     area = yfacesA[:,dim[1]-1]
278     nx = yfacesN[:,dim[1]-1,0]/area
279     ny = yfacesN[:,dim[1]-1,1]/area
280
281     # Lower states
282     rho_d = r[0, :, dim[1]-2]
283     u_d = r[1, :, dim[1]-2]
284     v_d = r[2, :, dim[1]-2]
285     p_d = r[3, :, dim[1]-2]
286     h0_d = (gamma/(gamma - 1))*(p_d/rho_d) + 0.5 * (u_d**2 + v_d**2)
287     a_d = np.sqrt(gamma*(p_d/rho_d))
288     u_nd = u_d*nx + v_d*ny
289
290     # Upper states
291     rho_u = bc_w[2]
292     p_u = bc_w[3]
293     a_u = np.sqrt(gamma*(p_u/rho_u))
294     u_nu = -u_nd
295
296     a_avg = 0.5*(a_d + a_u)
297     xma_d = u_nd/a_avg
298     xma_u = u_nu/a_avg
299
300     al_d = 0.5*(1.0 + np.sign((xma_d)))
301     al_u = 0.5*(1.0 - np.sign((xma_u)))
302
303     be_d = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_d)))
304     be_u = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_u)))
305
306     d_d = al_d*(1+be_d) - be_d*d_pl
307     d_u = al_u*(1+be_u) - be_u*d_mi
308
309     cvl_pl = al_d * (1+be_d) * xma_d - be_d * 0.25 * (1+xma_d)**2
310     cvl_mi = al_u * (1+be_u) * xma_u + be_u * 0.25 * (1-xma_u)**2
311
312     # Using the van Leer polynomials to define the Liou-Stefan
313     # polynomials
314     cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
315     cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
316
317     # Defining the mass-flux using the LS formulation
318     fm_pl = 0
319     fm_mi = 0
320     yFluxes[:,dim[1]-1, 0] = fm_pl + fm_mi
321     yFluxes[:,dim[1]-1, 1] = fm_pl * u_d + fm_mi * u_u + (d_d*p_d + d_u
322         *p_u) * nx*area
323     yFluxes[:,dim[1]-1, 2] = fm_pl * v_d + fm_mi * v_u + (d_d*p_d + d_u
324         *p_u) * ny*area
325     yFluxes[:,dim[1]-1, 3] = fm_pl * h0_d + fm_mi * h0_u

```

```

323
324     res = ((xFluxes[1:, :, :] - xFluxes[:-1, :, :]) + (yFluxes[:, 1:, :]
325         :]
326         - yFluxes[:, :-1, :]))

```

Listing 6: Minmod limiter

```

1 def minmod(a,b):
2     s = np.sign(a) + np.sign(b)
3     return 0.5 * s * np.minimum(np.abs(a), np.abs(b))

```

Listing 7: MUSCL reconstruction

```

1 # State reconstruction at faces
2 def muscl_reco(dim, r):
3
4     # Face state arrays
5     rl = np.zeros_like(r)                      # Left face states
6     rr = np.zeros_like(r)                      # Right face states
7     rd = np.zeros_like(r)                      # Lower face states
8     ru = np.zeros_like(r)                      # Upper face states
9
10    for i in range(1, dim[0]-2):
11        drpl0 = r[0,i+1,:] - r[0,i,:]
12        drmi0 = r[0,i,:] - r[0,i-1,:]
13        drpl1 = r[1,i+1,:] - r[1,i,:]
14        drmi1 = r[1,i,:] - r[1,i-1,:]
15        drpl2 = r[2,i+1,:] - r[2,i,:]
16        drmi2 = r[2,i,:] - r[2,i-1,:]
17        drpl3 = r[3,i+1,:] - r[3,i,:]
18        drmi3 = r[3,i,:] - r[3,i-1,:]
19
20        rl[0,i,:] = r[0,i,:] + 0.5*minmod(drpl0,drmi0)
21        rr[0,i,:] = r[0,i,:] - 0.5*minmod(drpl0,drmi0)
22        rl[1,i,:] = r[1,i,:] + 0.5*minmod(drpl1,drmi1)
23        rr[1,i,:] = r[1,i,:] - 0.5*minmod(drpl1,drmi1)
24        rl[2,i,:] = r[2,i,:] + 0.5*minmod(drpl2,drmi2)
25        rr[2,i,:] = r[2,i,:] - 0.5*minmod(drpl2,drmi2)
26        rl[3,i,:] = r[3,i,:] + 0.5*minmod(drpl3,drmi3)
27        rr[3,i,:] = r[3,i,:] - 0.5*minmod(drpl3,drmi3)
28
29    # Left cells
30    drpl0 = r[0,1,:] - r[0,0,:]
31    drmi0 = r[0,0,:] - in_bc(dim,r,mode)[0]
32    drpl1 = r[1,1,:] - r[1,0,:]
33    drmi1 = r[1,0,:] - in_bc(dim,r,mode)[1]
34    drpl2 = r[2,1,:] - r[2,0,:]
35    drmi2 = r[2,0,:] - in_bc(dim,r,mode)[2]
36    drpl3 = r[3,1,:] - r[3,0,:]
37    drmi3 = r[3,0,:] - in_bc(dim,r,mode)[3]
38
39    rl[0,0,:] = r[0,0,:] + 0.5*minmod(drpl0,drmi0)
40    rr[0,0,:] = r[0,0,:] - 0.5*minmod(drpl0,drmi0)
41    rl[1,0,:] = r[1,0,:] + 0.5*minmod(drpl1,drmi1)
42    rr[1,0,:] = r[1,0,:] - 0.5*minmod(drpl1,drmi1)
43    rl[2,0,:] = r[2,0,:] + 0.5*minmod(drpl2,drmi2)
44    rr[2,0,:] = r[2,0,:] - 0.5*minmod(drpl2,drmi2)

```

```

45     rl[3,0,:] = r[3,0,:]
46     rr[3,0,:] = r[3,0,:]
47
48     # Right cells
49     drpl0 = out_bc(dim,r,mode)[0] - r[0,dim[0]-2,:]
50     drmi0 = r[0,dim[0]-2,:]
51     drpl1 = out_bc(dim,r,mode)[1] - r[1,dim[0]-2,:]
52     drmi1 = r[1,dim[0]-2,:]
53     drpl2 = out_bc(dim,r,mode)[2] - r[2,dim[0]-2,:]
54     drmi2 = r[2,dim[0]-2,:]
55     drpl3 = out_bc(dim,r,mode)[3] - r[3,dim[0]-2,:]
56     drmi3 = r[3,dim[0]-2,:]
57
58     rl[0,dim[0]-2,:] = r[0,dim[0]-2,:]
59     rr[0,dim[0]-2,:] = r[0,dim[0]-2,:]
60     rl[1,dim[0]-2,:] = r[1,dim[0]-2,:]
61     rr[1,dim[0]-2,:] = r[1,dim[0]-2,:]
62     rl[2,dim[0]-2,:] = r[2,dim[0]-2,:]
63     rr[2,dim[0]-2,:] = r[2,dim[0]-2,:]
64     rl[3,dim[0]-2,:] = r[3,dim[0]-2,:]
65     rr[3,dim[0]-2,:] = r[3,dim[0]-2,:]
66
67     for j in range(1, dim[1]-2):
68         drpl0 = r[0,:,:j+1] - r[0,:,:j]
69         drmi0 = r[0,:,:j]
70         drpl1 = r[1,:,:j+1] - r[1,:,:j]
71         drmi1 = r[1,:,:j]
72         drpl2 = r[2,:,:j+1] - r[2,:,:j]
73         drmi2 = r[2,:,:j]
74         drpl3 = r[3,:,:j+1] - r[3,:,:j]
75         drmi3 = r[3,:,:j]
76
77         rd[0,:,:j] = r[0,:,:j]
78         ru[0,:,:j] = r[0,:,:j]
79         rd[1,:,:j] = r[1,:,:j]
80         ru[1,:,:j] = r[1,:,:j]
81         rd[2,:,:j] = r[2,:,:j]
82         ru[2,:,:j] = r[2,:,:j]
83         rd[3,:,:j] = r[3,:,:j]
84         ru[3,:,:j] = r[3,:,:j]
85
86     # Lower cells
87     drpl0 = r[0,:,1] - r[0,:,0]
88     drmi0 = r[0,:,0]
89     drpl1 = r[1,:,1] - r[1,:,0]
90     drmi1 = r[1,:,0]
91     drpl2 = r[2,:,1] - r[2,:,0]
92     drmi2 = r[2,:,0]
93     drpl3 = r[3,:,1] - r[3,:,0]
94     drmi3 = r[3,:,0]
95
96     rd[0,:,0] = r[0,:,0]
97     ru[0,:,0] = r[0,:,0]
98     rd[1,:,0] = r[1,:,0]
99     ru[1,:,0] = r[1,:,0]
100    rd[2,:,0] = r[2,:,0]
101    ru[2,:,0] = r[2,:,0]
102    rd[3,:,0] = r[3,:,0]
103    ru[3,:,0] = r[3,:,0]

```

```

103
104     # Upper cells
105     drpl0 = slipwall(dim,r,yfacesA,yfacesN)[2] - r[0,:,:dim[1]-2]
106     drmi0 = r[0,:,:dim[1]-2] - r[0,:,:dim[1]-3]
107     drpl1 = 0
108     drmi1 = r[1,:,:dim[1]-2] - r[1,:,:dim[1]-3]
109     drpl2 = -2*(r[2,:,:dim[1]-2])
110     drmi2 = r[2,:,:dim[1]-2] - r[2,:,:dim[1]-3]
111     drpl3 = slipwall(dim,r,yfacesA,yfacesN)[3] - r[3,:,:dim[1]-2]
112     drmi3 = r[3,:,:dim[1]-2] - r[3,:,:dim[1]-3]
113
114     rd[0,:,:dim[1]-2] = r[0,:,:dim[1]-2] + 0.5*minmod(drpl0,drmi0)
115     ru[0,:,:dim[1]-2] = r[0,:,:dim[1]-2] - 0.5*minmod(drpl0,drmi0)
116     rd[1,:,:dim[1]-2] = r[1,:,:dim[1]-2] + 0.5*minmod(drpl1,drmi1)
117     ru[1,:,:dim[1]-2] = r[1,:,:dim[1]-2] - 0.5*minmod(drpl1,drmi1)
118     rd[2,:,:dim[1]-2] = r[2,:,:dim[1]-2] + 0.5*minmod(drpl2,drmi2)
119     ru[2,:,:dim[1]-2] = r[2,:,:dim[1]-2] - 0.5*minmod(drpl2,drmi2)
120     rd[3,:,:dim[1]-2] = r[3,:,:dim[1]-2] + 0.5*minmod(drpl3,drmi3)
121     ru[3,:,:dim[1]-2] = r[3,:,:dim[1]-2] - 0.5*minmod(drpl3,drmi3)
122
123     return rl, rr, rd, ru

```

Listing 8: AUSM (using MUSCL reconstruction)

```

1 def fluxls_muscl(dim, r, rl, rr, rd, ru):
2
3     # Fluxes initialization
4     xFluxes = np.zeros((dim[0], dim[1]-1, 4))
5     yFluxes = np.zeros((dim[0]-1, dim[1], 4))
6
7     # X-face fluxes
8     for i in range(1, dim[0]-1):
9         area = xfacesA[i,:]
10        nx = xfacesN[i,:,0]/area
11        ny = xfacesN[i,:,1]/area
12        # Left states
13        rho_l = rr[0, i-1, :]
14        u_l = rr[1, i-1, :]
15        v_l = rr[2, i-1, :]
16        p_l = rr[3, i-1, :]
17        h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l
18                    **2)
19        a_l = np.sqrt(gamma*(p_l/rho_l))
20        u_nl = u_l*nx + v_l*ny
21        # Right states
22        rho_r = rl[0, i, :]
23        u_r = rl[1, i, :]
24        v_r = rl[2, i, :]
25        p_r = rl[3, i, :]
26        h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r
27                    **2)
28        a_r = np.sqrt(gamma*(p_r/rho_r))
29        u_nr = u_r*nx + v_r*ny
30
31        a_avg = 0.5*(a_l + a_r)
32        xma_l = u_nl/a_avg
33        xma_r = u_nr/a_avg

```

```

33     al_l = 0.5*(1.0 + np.sign((xma_l)))
34     al_r = 0.5*(1.0 - np.sign((xma_r)))
35
36     be_l = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_l)))
37     be_r = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_r)))
38
39     # Defining the van Leer polynomials
40     d_pl = 0.25*((xma_l+1)**2)*(2-xma_l)
41     d_mi = 0.25*((xma_r-1)**2)*(2+xma_r)
42
43     d_l = al_l*(1+be_l) - be_l*d_pl
44     d_r = al_r*(1+be_r) - be_r*d_mi
45
46     cvl_pl = al_l * (1+be_l) * xma_l - be_l * 0.25 * (1+xma_l)**2
47     cvl_mi = al_r * (1+be_r) * xma_r + be_r * 0.25 * (1-xma_r)**2
48
49     # Using the van Leer polynomials to define the Liou-Stefan
50     # polynomials
51     cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
52     cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
53
54     # Defining the mass-flux using the LS formulation
55     fm_pl = area * a_avg * cls_pl * rho_l
56     fm_mi = area * a_avg * cls_mi * rho_r
57     xFluxes[:, :, 0] = fm_pl + fm_mi
58     xFluxes[:, :, 1] = fm_pl * u_l + fm_mi * u_r + (d_l*p_l + d_r*
59     p_r) * nx*area
60     xFluxes[:, :, 2] = fm_pl * v_l + fm_mi * v_r + (d_l*p_l + d_r*
61     p_r) * ny*area
62     xFluxes[:, :, 3] = fm_pl * h0_l + fm_mi * h0_r
63
64     # Left boundary flux
65     area = xfacesA[0,:]
66     nx = xfacesN[0,:,:0]/area
67     ny = xfacesN[0,:,:1]/area
68     # Left states
69     bc_l = in_bc(dim, r, mode)
70     rho_l = bc_l[0]
71     u_l = bc_l[1]
72     v_l = bc_l[2]
73     p_l = bc_l[3]
74     h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l**2)
75     a_l = np.sqrt(gamma*(p_l/rho_l))
76     u_nl = u_l*nx + v_l*ny
77     # Right states
78     rho_r = rl[0, 0, :]
79     u_r = rl[1, 0, :]
80     v_r = rl[2, 0, :]
81     p_r = rl[3, 0, :]
82     h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r**2)
83     a_r = np.sqrt(gamma*(p_r/rho_r))
84     u_nr = u_r*nx + v_r*ny
85
86     a_avg = 0.5*(a_l + a_r)
87     xma_l = u_nl/a_avg
88     xma_r = u_nr/a_avg
89
90     al_l = 0.5*(1.0 + np.sign((xma_l)))

```

```

88     al_r = 0.5*(1.0 - np.sign((xma_r)))
89
90     be_l = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_l)))
91     be_r = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_r)))
92
93     # Defining the van Leer polynomials
94     d_pl = 0.25*((xma_l+1)**2)*(2-xma_l))
95     d_mi = 0.25*((xma_r-1)**2)*(2+xma_r))
96
97     d_l = al_l*(1+be_l) - be_l*d_pl
98     d_r = al_r*(1+be_r) - be_r*d_mi
99
100    cvl_pl = al_l * (1+be_l) * xma_l - be_l * 0.25 * (1+xma_l)**2
101    cvl_mi = al_r * (1+be_r) * xma_r + be_r * 0.25 * (1-xma_r)**2
102
103    # Using the van Leer polynomials to define the Liou-Stefan
104    # polynomials
105    cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
106    cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
107
108    # Defining the mass-flux using the LS formulation
109    fm_pl = area * a_avg * cls_pl * rho_l
110    fm_mi = area * a_avg * cls_mi * rho_r
111    xFluxes[0,:, 0] = fm_pl + fm_mi
112    xFluxes[0,:, 1] = fm_pl * u_l + fm_mi * u_r + (d_l*p_l + d_r*p_r) *
113        nx*area
114    xFluxes[0,:, 2] = fm_pl * v_l + fm_mi * v_r + (d_l*p_l + d_r*p_r) *
115        ny*area
116    xFluxes[0,:, 3] = fm_pl * h0_l + fm_mi * h0_r
117
118    # Right boundary flux
119    area = xfacesA[dim[0]-1,:]
120    nx = xfacesN[dim[0]-1,:,:0]/area
121    ny = xfacesN[dim[0]-1,:,:1]/area
122    # Left states
123    rho_l = rr[0, dim[0]-2, :]
124    u_l = rr[1, dim[0]-2, :]
125    v_l = rr[2, dim[0]-2, :]
126    p_l = rr[3, dim[0]-2, :]
127    h0_l = (gamma/(gamma - 1))*(p_l/rho_l) + 0.5 * (u_l**2 + v_l**2)
128    a_l = np.sqrt(gamma*(p_l/rho_l))
129    u_nl = u_l*nx + v_l*ny
130    # Right states
131    bc_r = out_bc(dim, r, mode)
132    rho_r = bc_r[0]
133    u_r = bc_r[1]
134    v_r = bc_r[2]
135    p_r = bc_r[3]
136    h0_r = (gamma/(gamma - 1))*(p_r/rho_r) + 0.5 * (u_r**2 + v_r**2)
137    a_r = np.sqrt(gamma*(p_r/rho_r))
138    u_nr = u_r*nx + v_r*ny
139
140    a_avg = 0.5*(a_l + a_r)
141    xma_l = u_nl/a_avg
142    xma_r = u_nr/a_avg
143
144    al_l = 0.5*(1.0 + np.sign((xma_l)))

```

```

143 al_r = 0.5*(1.0 - np.sign((xma_r)))
144
145 be_l = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_l)))
146 be_r = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_r)))
147
148 # Defining the van Leer polynomials
149 d_pl = 0.25*((xma_l+1)**2)*(2-xma_l))
150 d_mi = 0.25*((xma_r-1)**2)*(2+xma_r))
151
152 d_l = al_l*(1+be_l) - be_l*d_pl
153 d_r = al_r*(1+be_r) - be_r*d_mi
154
155 cvl_pl = al_l * (1+be_l) * xma_l - be_l * 0.25 * (1+xma_l)**2
156 cvl_mi = al_r * (1+be_r) * xma_r + be_r * 0.25 * (1-xma_r)**2
157
158 # Using the van Leer polynomials to define the Liou-Stefan
159 # polynomials
160 cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
161 cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
162
163 # Defining the mass-flux using the LS formulation
164 fm_pl = area * a_avg * cls_pl * rho_l
165 fm_mi = area * a_avg * cls_mi * rho_r
166 xFluxes[dim[0]-1,:, 0] = fm_pl + fm_mi
167 xFluxes[dim[0]-1,:, 1] = fm_pl * u_l + fm_mi * u_r + (d_l*p_l + d_r
168 *p_r) * nx*area
169 xFluxes[dim[0]-1,:, 2] = fm_pl * v_l + fm_mi * v_r + (d_l*p_l + d_r
170 *p_r) * ny*area
171 xFluxes[dim[0]-1,:, 3] = fm_pl * h0_l + fm_mi * h0_r
172
173 # Y-face fluxes
174 for j in range(1, dim[1]-1):
175     area = yfacesA[:,j]
176     nx = yfacesN[:,j,0]/area
177     ny = yfacesN[:,j,1]/area
178     # Lower states
179     rho_d = ru[0, :, j-1]
180     u_d = ru[1, :, j-1]
181     v_d = ru[2, :, j-1]
182     p_d = ru[3, :, j-1]
183     h0_d = (gamma/(gamma - 1))*(p_d/rho_d) + 0.5 * (u_d**2 + v_d
184     **2)
185     a_d = np.sqrt(gamma*(p_d/rho_d))
186     u_nd = u_d*nx + v_d*ny
187     # Upper states
188     rho_u = rd[0, :, j]
189     u_u = rd[1, :, j]
190     v_u = rd[2, :, j]
191     p_u = rd[3, :, j]
192     h0_u = (gamma/(gamma - 1))*(p_u/rho_u) + 0.5 * (u_u**2 + v_u
193     **2)
194     a_u = np.sqrt(gamma*(p_u/rho_u))
195     u_nu = u_u*nx + v_u*ny
196
197     a_avg = 0.5*(a_d + a_u)
198     xma_d = u_nd/a_avg
199     xma_u = u_nu/a_avg

```

```

196
197     al_d = 0.5*(1.0 + np.sign((xma_d)))
198     al_u = 0.5*(1.0 - np.sign((xma_u)))
199
200     be_d = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_d)))
201     be_u = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_u)))
202
203     # Defining the van Leer polynomials
204     d_pl = 0.25*((xma_d+1)**2)*(2-xma_d))
205     d_mi = 0.25*((xma_u-1)**2)*(2+xma_u))
206
207     d_d = al_d*(1+be_d) - be_d*d_pl
208     d_u = al_u*(1+be_u) - be_u*d_mi
209
210     cvl_pl = al_d * (1+be_d) * xma_d - be_d * 0.25 * (1+xma_d)**2
211     cvl_mi = al_u * (1+be_u) * xma_u + be_u * 0.25 * (1-xma_u)**2
212
213     # Using the van Leer polynomials to define the Liou-Stefan
214     # polynomials
215     cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
216     cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
217
218     # Defining the mass-flux using the LS formulation
219     fm_pl = area * a_avg * cls_pl * rho_d
220     fm_mi = area * a_avg * cls_mi * rho_u
221     yFluxes[:,j, 0] = fm_pl + fm_mi
222     yFluxes[:,j, 1] = fm_pl * u_d + fm_mi * u_u + (d_d*p_d + d_u*
223         p_u) * nx*area
224     yFluxes[:,j, 2] = fm_pl * v_d + fm_mi * v_u + (d_d*p_d + d_u*
225         p_u) * ny*area
226     yFluxes[:,j, 3] = fm_pl * h0_d + fm_mi * h0_u
227
228     bc_w = slipwall(dim,r,yfacesA,yfacesN)
229
230     # Lower boundary flux
231     area = yfacesA[:,0]
232     nx = yfacesN[:,0,0]/area
233     ny = yfacesN[:,0,1]/area
234
235     # Upper states
236     rho_u = rd[0, :, 0]
237     u_u = rd[1, :, 0]
238     v_u = rd[2, :, 0]
239     p_u = rd[3, :, 0]
240     h0_u = (gamma/(gamma - 1))*(p_u/rho_u) + 0.5 * (u_u**2 + v_u**2)
241     a_u = np.sqrt(gamma*(p_u/rho_u))
242     u_nu = u_u*nx + v_u*ny
243
244     # Lower states
245     u_nd = -u_nu
246
247
248     a_avg = 0.5*(a_d + a_u)
249     xma_d = u_nd/a_avg
250     xma_u = u_nu/a_avg

```

```

251     be_d = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_d)))
252     be_u = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_u)))
253
254     # Defining the van Leer polynomials
255     d_pl = 0.25*((xma_d+1)**2)*(2-xma_d)
256     d_mi = 0.25*((xma_u-1)**2)*(2+xma_u)
257
258     d_d = al_d*(1+be_d) - be_d*d_pl
259     d_u = al_u*(1+be_u) - be_u*d_mi
260
261     cvl_pl = al_d * (1+be_d) * xma_d - be_d * 0.25 * (1+xma_d)**2
262     cvl_mi = al_u * (1+be_u) * xma_u + be_u * 0.25 * (1-xma_u)**2
263
264     # Using the van Leer polynomials to define the Liou-Stefan
265     # polynomials
266     cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
267     cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
268
269     # Defining the mass-flux using the LS formulation
270     fm_pl = 0
271     fm_mi = 0
272     yFluxes[:,0, 0] = fm_pl + fm_mi
273     yFluxes[:,0, 1] = fm_pl * u_d + fm_mi * u_u + (d_d*p_d + d_u*p_u) *
274         nx*area
275     yFluxes[:,0, 2] = fm_pl * v_d + fm_mi * v_u + (d_d*p_d + d_u*p_u) *
276         ny*area
277     yFluxes[:,0, 3] = fm_pl * h0_d + fm_mi * h0_u
278
279     # Upper boundary flux
280     area = yfacesA[:,dim[1]-1]
281     nx = yfacesN[:,dim[1]-1,0]/area
282     ny = yfacesN[:,dim[1]-1,1]/area
283     # Lower states
284     rho_d = ru[0, :, dim[1]-2]
285     u_d = ru[1, :, dim[1]-2]
286     v_d = ru[2, :, dim[1]-2]
287     p_d = ru[3, :, dim[1]-2]
288     h0_d = (gamma/(gamma - 1))*(p_d/rho_d) + 0.5 * (u_d**2 + v_d**2)
289     a_d = np.sqrt(gamma*(p_d/rho_d))
290     u_nd = u_d*nx + v_d*ny
291     # Upper states
292     u_nu = -u_nd
293
294     a_avg = 0.5*(a_d + a_u)
295     xma_d = u_nd/a_avg
296     xma_u = u_nu/a_avg
297
298     al_d = 0.5*(1.0 + np.sign((xma_d)))
299     al_u = 0.5*(1.0 - np.sign((xma_u)))
300
301     be_d = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_d)))
302     be_u = -np.maximum(0.0, 1.0-np.floor(np.abs(xma_u)))
303
304     # Defining the van Leer polynomials
305     d_pl = 0.25*((xma_d+1)**2)*(2-xma_d)
306     d_mi = 0.25*((xma_u-1)**2)*(2+xma_u)

```

```

306     d_d = al_d*(1+be_d) - be_d*d_pl
307     d_u = al_u*(1+be_u) - be_u*d_mi
308
309     cvl_pl = al_d * (1+be_d) * xma_d - be_d * 0.25 * (1+xma_d)**2
310     cvl_mi = al_u * (1+be_u) * xma_u + be_u * 0.25 * (1-xma_u)**2
311
312     # Using the van Leer polynomials to define the Liou-Stefan
313     # polynomials
314     cls_pl = np.maximum(0.0, cvl_pl + cvl_mi)
315     cls_mi = np.minimum(0.0, cvl_pl + cvl_mi)
316
317     # Defining the mass-flux using the LS formulation
318     fm_pl = 0
319     fm_mi = 0
320     yFluxes[:,dim[1]-1, 0] = fm_pl + fm_mi
321     yFluxes[:,dim[1]-1, 1] = fm_pl * u_d + fm_mi * u_u + (d_d*p_d + d_u
322         *p_u) * nx*area
323     yFluxes[:,dim[1]-1, 2] = fm_pl * v_d + fm_mi * v_u + (d_d*p_d + d_u
324         *p_u) * ny*area
325     yFluxes[:,dim[1]-1, 3] = fm_pl * h0_d + fm_mi * h0_u
326
327     res = ((xFluxes[1:, :, :] - xFluxes[:-1, :, :]) + (yFluxes[:, 1:, :]
328         :] - yFluxes[:, :-1, :]))
329
330     return res

```

Listing 9: Initializing the solution variables and time step calculations

```

1  # Initializing the primitive variables' (r) and conservative variables'
2  # (q) arrays
3
4  if not restart:
5      r = np.ones((4,dim[0]-1,dim[1]-1))
6      q = np.ones((4,dim[0]-1,dim[1]-1))
7
8      r[0] = r[0] * rho_inf
9      r[1] = r[1] * u_inf
10     r[2] = r[2] * v_inf
11     r[3] = r[3] * p_stat
12
13     q[0] = q[0] * r[0]
14     q[1] = q[1] * r[0] * r[1]
15     q[2] = q[2] * r[0] * r[2]
16     q[3] = q[3] * (r[3]/(gamma-1) + (0.5*r[0]*(r[2]**2+r[3]**2)))
17
18 else:
19     res_data = np.load('solfiles/solution_iter_130000.npz')
20     r = res_data['r']
21     q = res_data['q']
22     res_arr = res_data['res']
23     last_res = res_arr[-1]
24
25     # Time-step calculation
26     def tcalc(dim, r, facenormals, faceareas, cellvols, cells):
27         delta_t = np.zeros((dim[0]-1,dim[1]-1))
28         for j in range(dim[1]-1):
29             for i in range(dim[0]-1):
30                 lambda_a = np.zeros(4)

```

```

30         vel_p = np.array([r[1, i, j], r[2, i, j]])
31         p_k = r[3, i, j]
32         rho_k = r[0, i, j]
33         a_k = np.sqrt(gamma*p_k/rho_k)
34
35     for k in range(4):
36         area = faceareas[cells[i,j]][k]
37         n = facenormals[cells[i,j]][k]
38         vn = np.dot(vel_p,n)
39         lambda_a[k] = np.abs(vn) + (a_k * area)
40
41
42     t_by_v = cfl/np.sum(np.abs(lambda_a))
43     delta_t[i,j] = t_by_v*cellvols[cells[i,j]]
44
45 return delta_t

```

Listing 10: Main loop

```

1  # Main Loop
2  if not restart:
3      resnorm = []
4  else:
5      resnorm = res_data['res'].tolist()
6
7  if not restart:
8      itr = 0
9  else:
10     itr = len(resnorm)
11
12 if not restart:
13     c_res = 1
14 else:
15     c_res = last_res
16
17 if not restart:
18     time_itr = 0.0
19 else:
20     time_itr = last_time
21
22 if high_recon:
23     rl,rr,rd,ru = muscl_reco(dim,r)
24
25 if high_recon:
26     res = fluxls_muscl(dim, r, rl, rr, rd, ru)          #
27             Initialization of residual vector
28 else:
29     res = fluxls(dim, r)
30
31 # Scales for convergence residuals
32 scale1 = rho_inf * u_inf
33 scale2 = rho_inf * u_inf * u_inf
34 scale3 = rho_inf * u_inf * u_inf
35 scale4 = rho_inf * u_inf * (((gamma*p_stat)/(p_stat*rho_inf))+(0.5*
36                               u_inf*u_inf))
37
38 while np.abs(c_res) > tol and itr <= 500000:

```

```

38
39     delta_t = tcalc(dim, r, facenormals, faceareas, cellvols, cells)
40     del_t = np.min(delta_t)
41     q_new = np.zeros_like(q)
42
43     # Updating the conservative variables
44     for k in range(4):
45         q_new[k] = q[k] - (del_t / vols_2d) * res[:, :, k]
46
47     # Updating the primitive variables
48     r[0] = q_new[0]
49     r[1] = q_new[1]/q_new[0]
50     r[2] = q_new[2]/q_new[0]
51     r[3] = (gamma-1) * (q_new[3] - 0.5*((q_new[1]**2 + q_new[2]**2)/
52                           q_new[0]))
53
54     if high_recon:
55         rl,rr,rd,ru = muscl_reco(dim,r)
56
57         # Updating the face
58         # reconstructed states
59
60         q = np.copy(q_new)
61
62         if high_recon:
63             res = fluxls_muscl(dim, r, rl, rr, rd, ru)
64
65             # Recalculating residuals
66
67         else:
68             res = fluxls(dim,r)
69
70         resnorm.append(0.0)
71
72         resnorm[itr] = np.sum((res[:, :, 0]/scale1)**2+(res[:, :, 1]/scale2)**2+
73                               (res[:, :, 2]/scale3)**2+(res[:, :, 3]/scale4)**2)
74
75
76         if itr == 0:
77             resnorm0 = resnorm[itr]
78             c_res = 1
79
80         else:
81             resnorm0 = resnorm[0]
82             stab = 1e-12
83             c_res = (resnorm[itr]) / (resnorm0 + stab)
84
85         time_itr += del_t
86
87         # if itr%50 == 0 or np.abs(c_res) <= tol:                      #
88             # Printing residuals every 50 iterations and after convergence
89             # print(f"Residual is {np.abs(c_res)} at time {time_itr}s
90             # corresponding to iteration {itr}.".")                         # SM: Only for
91             # debugging. Remove later.
92             print(f"Residual is {np.abs(c_res)} at time {time_itr}s
93                 corresponding to iteration {itr}.".")                     # SM: Only for
94                 debugging. Remove later.
95         itr += 1
96
97         if itr%5000 == 0 or np.abs(c_res) <= tol:                      #
98             # Saving every 5000 iterations and after convergence
99             resnorm_arr = np.array(resnorm)
100            filename = f"solfiles/solution_iter_{itr}.npz"

```

```
85     np.savez(filename, q=q, r=r, res=resnorm_arr)
86     print(f"Saved solution to {filename}")
```