A High-Order Compact Gas-Kinetic Scheme in a Rotating Coordinates Frame and on Sliding Mesh

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ABSTRACT

This paper extends the high-order compact gas-kinetic scheme (CGKS) to flow simulations on a rotating coordinate frame. The kinetic equation with the inclusion of centrifugal and Coriolis accelerations is used in the construction of the scheme. With the updates of both cell-averaged flow variables and their gradients in the rotating and stationary domains, a third-order compact reconstruction is developed with sliding interface between them. To capture shock waves and complicated wave interactions, the HWENO-type non-linear reconstruction and gradient compression factors are incorporated in the scheme. For achieving high-order time accuracy, based on the time-accurate flux function the multi-stage multi-derivative time stepping method is implemented in the scheme for the fourth-order time accuracy with two stages. The CGKS is validated in cases from subsonic acoustic wave propagation to high Mach number shock interaction. The compact scheme achieves high-order accuracy and remarkable robustness.

1. Introduction

Flow simulations with rotating parts, including turbo machinery, helicopters, tilt rotors and ship propellers, have significant industrial applications. The computational domain is usually divided into moving and stationary parts with a sliding interface between them. This paper is about the development of a high-order compact gas-kinetic scheme on a rotating coordinate frame and connect its solution with the stationary domain through a sliding interface.

The gas-kinetic scheme (GKS) is a kinetic theory-based numerical method to solve the Euler and Navier–Stokes equations (Xu 2001). Under the initial condition of a generalised Riemann problem, a time-accurate gas distribution function is constructed in GKS to calculate the numerical fluxes and evaluate the time-dependent flow variables at a cell interface. As a result, both cell-averaged flow variables and their gradients can be updated. Therefore, the HWENO-type method and the two-step multi-resolution WENO reconstruction can be used in the scheme for the high-order spatial data reconstruction (Zhu and Qiu 2018; Ji et al. “Two-Step Multi-Resolution” 2021). At the same time, due to the time-accurate flux function, the multi-stage multi-derivative (MSMD) method is adopted to update the solution with high-order temporary accuracy. Specifically, the two-stage fourth-order (S2O4) time stepping method is used in the compact GKS (CGKS) (Li and Du 2016; Li 2019). The CGKS has been constructed on both structured and unstructured meshes in 2D and 3D cases (Ji et al. 2018; Ji et al. “Compact high-order gas-kinetic” 2021; Zhao et al. 2019, 2020, 2022). To further improve the robustness of the scheme in high-speed flow simulations, the following modelling has been further incorporated in the scheme. First, the evolution of possible discontinuous flow variables at different sides of a cell interface is constructed for updating reliable cell averaged gradient of flow variables (Zhao et al. “Direct modeling for computational” 2023; Zhao et al. “High-order compact gas-kinetic” 2023). Second, the nonlinear limiting process is implemented on the high-order time derivative of the flux function under MSMD framework. Equipped with the above remedies, the CGKS
on 3D unstructured mesh is extremely robust in hypersonic flow computation and is able to use a large time step, such as CFL number ≈ 0.8, in the fourth-order compact scheme. Alternatively, a gradient compression factor is designed to improve the robustness and efficiency of CGKS in cases with low-quality mesh (Ji, Shyy, and Xu 2021).

There are two ways to solve the flow problems with a rotating mesh movement. The first one is the arbitrary Lagrangian–Eulerian (ALE)-based moving mesh method (Hirt, Amsden, and Cook 1974; Zhang and Liang 2015; Duan, Jia, and Wang 2020). These methods are similar to the methods under the unified coordinates (Hui and Xu 2012; Jin and Xu 2007), where flow variables and geometric conservation laws have to be solved simultaneously. Another approach fixes the coordinate on a rotating frame. In such non-inertia reference of the frame, the centrifugal and Coriolis forces will appear in the kinetic governing equation (Zhou 2019). In this paper, we are going to develop the GKS by following the second approach. With the inclusion of external forces in the kinetic equation, the CGKS can be constructed with the inclusion of the forcing effect on the particle trajectory (Xu 2002; Luo, Xu, and Liu 2011). The corresponding macroscopic governing equations solved by the CGKS will be derived using the Chapman–Enskog expansion.

The sliding-mesh method has been developed for many years. Johnstone, Chen, and Sandberg (2015) proposed a novel sliding-mesh method based on a characteristic interface condition. Their new sliding gird technique requires only a single layer of halo nodes in the communication process. Ramírez et al. (2015) developed a high-order sliding mesh interface to simulate unsteady viscous flow. Both compressible inviscid flow and incompressible viscous flow were simulated with the moving least squares (MLS) (Cueto-Felgueroso et al. 2007) reconstruction at the sliding interface. All the above sliding mesh methods are for the non-compact schemes. The high-order compact schemes have many advantages in comparison with non-compact ones due to the compact stencils around the sliding interface. Many high-order compact schemes have been developed based on the evolution of cell’s inner degrees of freedom, such as discontinuous Galerkin (DG) (Cockburn and Shu 1998), spectral difference (SD) (Liu, Vinokur, and Wang 2006), flux reconstruction (FR) (Huynh 2007) and correction procedure via reconstruction (CPR) (Haga, Gao, and Wang 2011). Based on FR/CPR scheme, Duan, Jia, and Wang (2020) developed a sliding mesh method by using an auxiliary Cartesian grid to exchange information between the sliding interface. Based on the SD method, Zhang and Liang (2015) used mortar elements to project flow variables and fluxes back and forth. To improve the adaptability of geometry, Zhang, Qiu, and Liang (2018) extended their work to deal with arbitrarily non-uniform mesh for the FR method. Recently, Gao (2022) developed a three-dimensional sliding mesh method based on the mortar approach and applied it to the turbine and noise problem. By adopting Sutherland–Hodgman algorithm (Sutherland and Hodgman 1974), the polygon clipping method can be used to deal with complicated geometry. In the current study, a third-order finite volume CGKS only requires Neumann neighbouring cells in the reconstruction. Therefore, it becomes straightforward to construct the corresponding CGKS with sliding mesh. Here, a ghost cell will be created by merging several cells in the reconstruction around the sliding interface. To ensure the conservative property, the mortar interface is generated for the calculation of fluxes.

This paper is organised as follows. The kinetic BGK equation and GKS in a rotating coordinate frame will be introduced in Section 2. Section 3 is about the two-stage four-order time integrating method for the solution updates with source terms. Section 4 concentrates on the initial reconstruction. The treatment of the sliding interface is presented in Section 5. Many test cases will be used to validate the current method in Section 6. The last section is the conclusion.

2. Gas-Kinetic Scheme

2.1. BGK Equation in Rotating Framework

The gas-kinetic BGK equation in a rotating frame is

$$\frac{\partial f}{\partial t} + \mathbf{w} \cdot \nabla f + a_w \cdot \nabla_{wf} = g - f, \tau,$$

where \( f = f(\mathbf{x}, t, \mathbf{w}, \xi) \) is the gas distribution function, \( g \) is the corresponding equilibrium state and \( \tau \) is the collision time. \( \mathbf{w} = (w_1, w_2, w_3) \) is the particle velocity in the rotating frame. And the acceleration in the rotating frame is

$$a_w = \frac{\mathbf{d}w}{\mathbf{d}t} = -\Omega \times (\Omega \times r) - 2(\Omega \times \mathbf{w}),$$
where $\Omega$ is the angular velocity of the rotating frame and $r$ is a position vector from the origin of rotation to the position of the particle. $-\Omega \times (\Omega \times r)$ is the centrifugal force and $-2(\Omega \times w)$ is the Coriolis force. Denote $v = (v_1, v_2, v_3)$ as the particle velocity in absolute inertia reference of frame, the relationship among velocities is

$$v = U + w,$$

where $U = \Omega \times r$ is the convection velocity due to the frame rotation. According to $a_v = \frac{dv}{dt} = \frac{dw}{dt} + (\Omega \times w)$, the acceleration term in the rotating frame can be expressed as $-(\Omega \times v) \cdot \nabla v$. Then the BGK equation becomes

$$\frac{df}{dt} + w \cdot \nabla f - (\Omega \times v) \cdot \nabla v = \frac{g - f}{\tau},$$

where $f$ can be defined by absolute velocity $v$, such as $f = f(x, t, v, \xi)$. The collision term in the above equation describes the evolution process from a non-equilibrium state to an equilibrium one with the satisfaction of compatibility condition

$$\int \frac{g - f}{\tau} \Psi d\Xi = 0,$$

where $\Psi = (1, v_1, v_2, v_3, \frac{1}{2}(v_1^2 + v_2^2 + v_3^2 + \xi^2))^T$ and $d\Xi = d\xi_1 \cdot d\xi_K$ ($K$ is the number of internal degree of freedom, i.e. $K = 2$ for three-dimensional diatomic gas). Based on the Chapman–Enskog Expansion (see Appendix I), the Euler and N–S equations in the rotating frame can be obtained. The N–S equations in a rotating frame are

$$\begin{align*}
\frac{\partial \rho}{\partial t} + \nabla \cdot (\rho (V - U)) &= 0, \\
\frac{\partial \rho V}{\partial t} + \nabla \cdot (\rho (V - U) V + pI - \overline{\sigma}) &= -\Omega \times \rho V, \\
\frac{\partial \rho E}{\partial t} + \nabla \cdot (\rho H(V - U) + pV - \kappa \nabla T - \overline{\sigma} \cdot V) &= 0,
\end{align*}$$

where $\rho, V, p, T, E, H$ and $\overline{\sigma}$ are the density, absolute velocity, pressure, temperature, energy, enthalpy and viscosity stress of fluid. With the gradient of temperature $\nabla T$ and viscosity stress $\overline{\sigma}$ equal to zero, the N–S equations become Euler equations.

### 2.2. Finite Volume Method

The whole domain $\Omega$ is discretised into small cells $\Omega_i$

$$\Omega = \bigcup_{i} \Omega_i, \Omega_i \cap \Omega_j = \phi (i \neq j).$$

The boundary can be expressed as

$$\partial \Omega_i = \bigcup_{p=1}^{N_f} \Gamma_{ip}.$$

Taking moments of the BGK equation (1) and integrating over the cell $\Omega_i$, the semi-discretised form of the finite volume scheme can be written as

$$\frac{dW_i}{dt} = -\frac{1}{|\Omega_i|} \sum_{p=1}^{N_f} \int_{\Gamma_{ip}} (F(W) \cdot n_p dS + S(W)),$$

where $W_i$ is the cell average conservative value, $|\Omega_i|$ is the volume of cell $\Omega_i$, $F$ is the flux via cell surface, $n_p = (n_1, n_2, n_3)^T$ is the normal direction of cell surface and $S$ is the source term due to rotation. The integration of flux can be approximated by Gaussian integrating (the index $i$ is omitted)

$$\int_{\Gamma_{ip}} F(W) \cdot n_p dS \approx |S_p| \sum_{k=1}^{M_p} \omega_k F(x_{p,k}, t) \cdot n_{p,k},$$

where $|S_p|$ is the cell surface area, $\omega_k$ is the weight of Gaussian integrating and $x_{p,k}$ is the position of Gaussian points on the cell surface. To calculate the flux through the surface, we can use coordinate transform

$$F(x_{p,k}, t) \cdot n_p = T^{-1} \tilde{F}(TW) = T^{-1} \tilde{F}(\tilde{W}),$$

where $T = \text{diag}(1, T', 1)$ is rotating matrix, and

$$T' = \begin{pmatrix}
    n_1 & n_2 & n_3 \\
    -n_2 & n_1 + \frac{n_2^2}{1+n_1} & -\frac{n_3}{1+n_1} \\
    -n_3 & -\frac{n_2n_3}{1+n_1} & 1 - \frac{n_3^2}{1+n_1}
\end{pmatrix}, \quad n_1 \neq -1,$$

and when $n_1 = -1$, $T'$ becomes diag$(-1, -1, 1)$. And the flux can be evaluated by

$$\tilde{F} = \int f(\tilde{x}_{p,k}, t, \tilde{v}, \xi)\tilde{w}_1d\tilde{\Xi},$$

where the origin point of the local coordinate is $\tilde{x}_{p,k} = (0, 0, 0)$ with $x$-direction in $n_p$, and $\tilde{W} = (1, \tilde{v}_1, \tilde{v}_2, \tilde{v}_3, \frac{1}{2}(\tilde{v}_1^2 + \tilde{v}_2^2 + \tilde{v}_3^2 + \xi^2))^T$. The microscopic velocities in local coordinate are given by $\tilde{v} = T'v$ and $\tilde{w}_1 = n_1w_1 + n_2w_2 + n_3w_3$. 

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2.3. Gas Evolution Model

To construct the numerical fluxes at \( x = (0, 0, 0)^T \), the integral solution of the BGK equation (1) is

\[
f(x, t, v, \xi) = \frac{1}{\tau_n} \int_0^t g(x', t', v', \xi) e^{-(t-t')/\tau_n} \, dt' + e^{(t-t_n)} f_0(x_0, v_0),
\]

where the partial absolute velocity is

\[
v = v' + \int_{t_n}^{t} a_i \, dt
\]

\[
= v' + \left[ v' - \left( v' \cdot \frac{\Omega}{\Omega} \right) \frac{\Omega}{\Omega} \right] (1 - \cos \Omega (t - t')) - \left( \frac{\Omega \times v'}{\Omega} \sin \Omega (t - t') \right)
\approx v' - (\Omega \times v')(t - t').
\]

In the rotating frame, the particle velocity and trajectory become \( w' - (\Omega \times v')(t - t') \) and

\[
x \approx x' + w'(t-t') + \frac{1}{2} (\Omega \times v')(t - t')^2.
\]

In Equation (4), \( f_0 \) is the initial gas distribution function and \( g \) is the corresponding equilibrium state. \((x_0, v_0)\) are the initial position and velocity by tracing back particles \((x, v)\) at time \( t \) back to \( t = 0. \tau_n \) is the numerical collision time (Luo and Xu 2013). For inviscid flow, it set as

\[
\tau_n = C_1 \Delta t + C_2 \frac{|p_l - p_r|}{p_l + p_r} \Delta t,
\]

and for viscous flow, it is

\[
\tau_n = \tau + C_2 \frac{|p_l - p_r|}{p_l + p_r} \Delta t,
\]

where \( p_l \) and \( p_r \) denote the pressure on the left and right sides of the cell interface. In this paper, we have \( C_1 = 0.01, C_2 = 5.0. \)

Before the construction of the initial distribution function \( f_0 \) and equilibrium state \( g \), we first denote

\[
a = (a_1, a_2, a_3) = \nabla_x g/g, A = g_1/g.
\]

In the following derivation, quadratic terms of time will be ignored directly. With the consideration of possible discontinuity at a cell interface, the initial distribution is constructed as

\[
f_0(x_0, v_0) = f_0^l(x_0, v_0)(1 - \mathbb{H}(x_1)) + f_0^r(x_0, v_0)\mathbb{H}(x_1),
\]

where \( \mathbb{H} \) is the Heaviside function. \( f_0^l \) and \( f_0^r \) are the initial gas distribution functions on the left and right sides of the interface, which are determined by corresponding initial macroscopic variables and their spatial derivatives. With the second-order accuracy, \( f^k_0(x_0, v_0) \) is constructed by Taylor expansion around \((x, v)\)

\[
f_0^k(x_0, v_0) = f_0^k(x, v) - wt \cdot \nabla_x f_0^k + (\Omega \times v) t \cdot \nabla_v f_0^k
\]

for \( k = l, r \). Due to Chapman–Enskog expansion, \( f_0^k \) is given by

\[
f_0^k = g^k[1 - \tau (A^k + a^k \cdot w)] + \tau (\Omega \times v) \cdot \nabla_v g^k,
\]

where \( g^k \) is the equilibrium distribution function defined by the macroscopic variables \( W^k \) at the both sides of a cell interface, \( a^k \) are defined by the spatial derivatives of \( g^k \)

\[
a^k_1 = \left( \frac{\partial g^k}{\partial \rho} \frac{\partial \rho}{\partial x_i} + \frac{\partial g^k}{\partial \nu_1} \frac{\partial \nu_1}{\partial x_i} + \frac{\partial g^k}{\partial \nu_2} \frac{\partial \nu_2}{\partial x_i} \right) f_0^k
\]

\[
+ a^k_2 v_1 + a^k_3 v_2 + a^k_4 v_3
\]

\[
+ a^k_5 \frac{1}{2}(v_1^2 + v_2^2 + v_3^2),
\]

and

\[
A^k = A^k_1 + A^k_2 v_1 + A^k_3 v_2 + A^k_4 v_3
\]

\[
+ A^k_5 \frac{1}{2}(v_1^2 + v_2^2 + v_3^2),
\]

are determined by compatibility condition

\[
\int (f_0^k - g^k) \psi \, dv \, d\xi = 0.
\]

Substituting Equations (6) and (7) into (5), the initial gas distribution has the following form:

\[
f_0 = \begin{cases}
  g^l \left[ 1 - (a^l \cdot w) t - \tau (A^l + a^l \cdot w) \right] \\
  + (t + \tau) (\Omega \times v) \cdot \nabla_v g^l, & x_1 < 0,
  \\
  g^r \left[ 1 - (a^r \cdot w) t - \tau (A^r + a^r \cdot w) \right] \\
  + (t + \tau) (\Omega \times v) \cdot \nabla_v g^r, & x_1 \geq 0.
\end{cases}
\]
Then, the equilibrium distribution is defined by the Taylor expansion

\[ g(x',t',v') = \bar{g}(x,0,v) + \nabla_{\bar{x}} \bar{g} \cdot (x' - x) + \nabla_{\bar{v}} \bar{g} \cdot (v' - v) + \bar{g}_{t} t' \]

where \( \bar{g} \) and \( \bar{a} \) are determined from the reconstruction of macroscopic flow variables presented in Section 4.5, and \( A \) is obtained by compatibility condition (8). By substituting Equations (9) and (10) into Equation (4) and keeping the second-order accuracy, the solution at a cell interface becomes

\[
f(x,t,v,\xi) = (1 - e^{-t/\tau}) \bar{g} + e^{-t/\tau} \left[ \mathcal{H}(w_1) g' + (1 - \mathcal{H}(w_1)) g'' \right] + tA \bar{g} - \tau (1 - e^{-t/\tau}) \left[ (\bar{a} \cdot w) g - (\bar{\Omega} \times v) \cdot \nabla \bar{g} + A \bar{g} \right] - \tau e^{-t/\tau} \mathcal{H}(w_1) \left[ (\bar{a'} \cdot w) g' - (\bar{\Omega} \times v) \cdot \nabla g' + A' g' \right] - \tau e^{-t/\tau} (1 - \mathcal{H}(w_1)) \left[ (\bar{a'} \cdot w) \bar{g} - \mathcal{H}(w_1) (\bar{a'} \cdot w) g \right] - (1 - \mathcal{H}(w_1)) (\bar{a'} \cdot w) g'.
\]  

(11)

The fluxes in Equation (3) can be obtained by taking the moments of the above distribution function. The calculation of moments can be found in Appendix 2.

2.4. Evolution of the Cell-Averaged Spatial Gradients

By taking moments of the above gas distribution function in Equation (11), the time-accurate conservative flow variables at a cell interface can be also obtained as

\[ W_{p,k}(t^{n+1}) = T' \left( \int_{\Gamma_p} \tilde{\Psi} f(\bar{x}_{p,k}, t^{n+1}, \bar{v}, \xi) d\Sigma \right) \].

(12)

According to Divergence theorem, the cell averaged gradients over cell \( \Omega_i \) at time \( t^{n+1} \) are

\[ \nabla W_{i}^{n+1} = \frac{1}{|\Omega_i|} \sum_{p=1}^{N_i} \int_{\Gamma_p} W^{n+1} \nabla n_p dS \].

(13)

where the surface integration can be calculated by Gaussian quadrature (the index \( i \) is omitted)

\[ \int_{\Gamma_p} W^{n+1} n_p dS \approx \sum_{k=1}^{M_p} |s_p| \omega_k W_{p,k}(t^{n+1}) n_{p,k}. \]  

(14)

Besides evaluating the cell averaged gradients, the solution updates of the scheme are presented next.

3. Solution Updates and Temporal Discretisation

According to the semi-discretisation equation (2), the right side contains two parts, the net flux \( \mathcal{L}_F \) and the source term. The two-stage fourth-order (S2O4) temporal discretisation is adopted here for the solution updates (Li and Du 2016),

\[ W_{i}^{n+1} = W_{i}^{n} + \frac{1}{2} \Delta t \mathcal{L}_F(W_{i}^{n}) + \frac{1}{8} \Delta t^2 \frac{\partial}{\partial t} \mathcal{L}_F(W_{i}^{n}), \]

\[ W_{i}^{n+1} = W_{i}^{n} + \int_{t^n}^{t^n + \frac{1}{2} \Delta t} S(W_{i}^{n}) dt, \]

\[ W_{i}^{(n+1)} = W_{i}^{n} + \Delta t \mathcal{L}_F(W_{i}^{n}) + \frac{1}{6} \Delta t^2 \left( \frac{\partial}{\partial t} \mathcal{L}_F(W_{i}^{n}) + 2 \frac{\partial}{\partial t} \mathcal{L}_F(W_{i}^{n}) \right) + \frac{1}{2} \Delta t \mathcal{L}_F(W_{i}^{n+1}) \]

\[ W_{i}^{n+1} = W_{i}^{(n+1)} + \int_{t^n}^{t^n + \Delta t} S(W_{i}^{(n+1)}) dt. \]

The source term only appears in moment equations \( \frac{d\rho V}{dt} = -\bar{\Omega} \times (\rho \bar{V}) \), which is integrated as

\[ \int_{t^n}^{t^n + \Delta t} -\bar{\Omega} \times (\rho \bar{V}) dt = -\left( \rho_n V_n \times \frac{\bar{\Omega}}{\Omega_2} \right) \sin (\Omega \Delta t) - \left[ \rho_n V_n \bar{\Omega} - \left( \rho_n V_n \cdot \frac{\bar{\Omega}}{\Omega_2} \right) \bar{\Omega} \right] (1 - \cos (\Omega \Delta t)). \]

The time-dependent gas distribution function at Gauss points on the interfaces is updated by

\[ f^* = f^n + \frac{1}{2} \Delta t f^n, \]

\[ f^{n+1} = f^n + \Delta t f^n. \]

where the time-dependent conservative values at each Gauss point can be obtained by Equation (12). Then by Equations (13) and (14), the cell-averaged slopes can be updated.
4. HWENO Reconstruction

The third-order compact reconstruction (Ji et al. 2020) is adopted here with cell-averaged values and cell-averaged first-order spatial derivative. To capture shock, WENO weights (Zhu and Shu 2020) and gradient compression factor (CF) (Ji, Shyy, and Xu 2021) are used. In this work, we further improve the WENO procedures and CF with the consideration of simplicity and robustness. Only one large stencil and one sub stencil are involved in the new WENO procedure.

4.1. Third-Order Compact Reconstruction for Large Stencil

Firstly, a linear reconstruction is presented. To achieve a third-order accuracy in space, a quadratic polynomial \( p^2 \) is constructed as follows:

\[
p^2 = a_0 + \frac{1}{h} [a_1 (x - x_0) + a_2 (y - y_0) + a_3 (z - z_0)] + \frac{1}{2h^2} [a_4 (x - x_0)^2 + a_5 (y - y_0)^2 + a_6 (z - z_0)^2] + \frac{1}{h^2} [a_7 (x - x_0)(y - y_0) + a_8 (y - y_0)(z - z_0) + a_9 (x - x_0)(z - z_0)],
\]

where \( h = \frac{v}{\max S_j} \) (V is the volume of cell and \( S_j \) is the area of cell’s surface) is the cell size, and \((x_0, y_0, z_0)\) is the coordinate of cell centre.

The \( p^2 \) on \( \Omega_0 \) is constructed on the compact stencil \( S \) including \( \Omega_0 \) and its all von Neumann neighbours \( \Omega_m(m = 1, \ldots, N_f) \), where \( N_f = 6 \) for hexahedron cell or \( N_f = 5 \) for triangular prism). The cell averages \( \overline{Q} \) over \( \Omega_0 \) and \( \Omega_m \) and cell averages of space partial derivatives \( \overline{Q}_x, \overline{Q}_y, \overline{Q}_z \) over \( \Omega_m \) are used to obtain \( p^2 \).

The polynomial \( p^2 \) is required to exactly satisfy cell averages over both \( \Omega_0 \) and \( \Omega_m \) \((m = 1, \ldots, N_f)\)

\[
\iiint_{\Omega_0} p^2 dV = \overline{Q}_0|\Omega_0|, \quad \iiint_{\Omega_m} p^2 dV = \overline{Q}_m|\Omega_m|,
\]

with the following condition satisfied in a least-square sense:

\[
\iiint_{\Omega_m} \frac{\partial}{\partial x} p^2 dV = (\overline{Q}_x)_m|\Omega_m|,
\]

\[
\iiint_{\Omega_m} \frac{\partial}{\partial y} p^2 dV = (\overline{Q}_y)_m|\Omega_m|,
\]

\[
\iiint_{\Omega_m} \frac{\partial}{\partial z} p^2 dV = (\overline{Q}_z)_m|\Omega_m|.
\]

To solve the above system, the constrained least-square method is used.

4.2. Green–Gauss Reconstruction for the Sub Stencil

The classical Green–Gauss reconstruction with only cell-averaged values is adopted to provide the linear polynomial \( p^1 \) for the sub stencil.

\[
p^1 = \overline{Q} + x \cdot \sum_{m=1}^{N_f} \frac{\overline{Q}_m + \overline{Q}_0}{2} S_m n_m,
\]

where \( S_m \) is the area of the cell’s surface and \( n_m \) is the surface’s normal vector.

4.3. Gradient Compression Factor

The CF was first proposed in Ji, Shyy, and Xu (2021). Here several improvements have been made: there is no \( \epsilon \) in the improved expression of CF; the difference of Mach number is added for improving the robustness under strong rarefaction waves. Denote \( \alpha_i \in [0, 1] \) as gradient compression factor at targeted cell \( \Omega_i \)

\[
\alpha_i = \prod_{p=1}^{m} \prod_{k=0}^{M_p} \alpha_{p,k},
\]

where \( \alpha_{p,k} \) is the CF obtained by the \( k \)th Gaussian point at the interface \( p \) around cell \( \Omega_i \), which can be calculated by

\[
\alpha_{p,k} = \frac{1}{1 + A^2},
\]

\[
A = \frac{|p^l - p^r|}{p^l} + \frac{|p^l - p^r|}{p^r} + (M_{a,l} - M_{a,r})^2 + (M_{a,l} - M_{a,r})^2,
\]

where \( p \) is the pressure, \( M_{a,l} \) and \( M_{a,r} \) are the Mach numbers defined by normal and tangential velocity, and superscript \( l, r \) denote the left and right values of the Gaussian points.

Then, the updated slope is modified by

\[
\overline{\nabla W}^{n+1}_i = \alpha_i \nabla W^{n+1}_i,
\]

and the Green–Gauss reconstruction is modified as

\[
p^1 = \overline{Q} + \alpha x \cdot \sum_{m=1}^{N_f} \frac{\overline{Q}_m + \overline{Q}_0}{2} S_m n_m.
\]
4.4. Non-Linear WENO Weights

To deal with discontinuity, the idea of multi-resolution WENO reconstruction is adopted (Ji, Shyy, and Xu 2021; Zhu and Shu 2020). Here only two polynomials are chosen

\[ P_2 = \frac{1}{\gamma_2} p_2^2 - \frac{\gamma_1}{\gamma_2} p_1^1, \quad P_1 = p_1^1. \]

Here, we choose \( \gamma_1 = \gamma_2 = 0.5 \). So the quadratic polynomial \( p_2^2 \) can be written as

\[ p_2^2 = \gamma_1 P_1 + \gamma_2 P_2. \] (15)

Then, we can define the smoothness indicators

\[ \beta_j = \sum_{|\alpha|=1} \Omega_1^{\frac{1}{2}|\alpha|-1} \iiint_{\Omega} \left( D^I p_j(x) \right)^2 \mathrm{d}V, \]

where \( \alpha \) is a multi-index and \( D \) is the derivative operator, \( r_1 = 1, r_2 = 2 \). Special care is given for \( \beta_1 \) for better robustness

\[ \beta_1 = \min(\beta_{1,\text{Green-Gauss}}, \beta_{1,\text{least-square}}), \]

where \( \beta_{1,\text{Green-Gauss}} \) is the smoothness indicator defined by Green-Gauss reconstruction and \( \beta_{1,\text{least-square}} \) is the smoothness indicator defined by second-order least-square reconstruction. Then the smoothness indicators \( \beta_i \) are non-dimensionalised by

\[ \tilde{\beta}_i = \frac{\beta_i}{Q_0^2 + \beta_1 + 10^{-40}}. \]

The nondimensionalised global smoothness indicator \( \tilde{\sigma} \) can be defined as

\[ \tilde{\sigma} = \left| \tilde{\beta}_1 - \tilde{\beta}_0 \right|^4. \]

Therefore, the corresponding non-linear weights are given by

\[ \tilde{\omega}_m = \gamma_m \left( 1 + \frac{\tilde{\sigma}}{\epsilon + \tilde{\beta}_m} \right)^2, \quad \epsilon = 10^{-5}, \]

\[ \tilde{\omega}_m = \frac{\tilde{\omega}_m}{\sum_m \tilde{\omega}_m}, \quad m = 1, 2. \]

Replacing \( \gamma_m \) in Equation (15) by \( \tilde{\omega}_m \), the final non-linear reconstruction can be obtained

\[ R(x) = \tilde{\omega}_2 P_2 + \tilde{\omega}_1 P_1. \]

The desired non-equilibrium states at Gaussian points become

\[ Q_{p,k}^{\text{I}} = R(x, p_k), \quad \left( Q_{x_i}^{\text{I}} \right)_{p,k} = \frac{\partial R}{\partial x_i}(x, p_k). \]

4.5. Reconstruction of Equilibrium States

After reconstructing the non-equilibrium state, a kinetic weighted average method can be used to get equilibrium states and tangential derivatives (Ji et al. 2020),

\[
\int \bar{g} \psi \mathrm{d}v \mathrm{d}x = \int \psi_0 \mathrm{d}v \mathrm{d}x + \int \psi_1 \mathrm{d}v \mathrm{d}x,
\]

\[
\int \tilde{a} \bar{g} \psi \mathrm{d}v \mathrm{d}x = \int \frac{\partial \psi}{\partial x_i} \mathrm{d}v \mathrm{d}x + \int \tilde{a}_i g \psi \mathrm{d}v \mathrm{d}x, \quad i = 2, 3.
\]

For the normal derivatives, the above solution is further modified according to the idea in linear diffusive generalised Riemann problem (dGRP) (Gassner, Lörcher, and Munz 2007)

\[
\int \tilde{a}_i \bar{g} \psi \mathrm{d}v \mathrm{d}x = \int \frac{\partial \psi}{\partial x_i} \mathrm{d}v \mathrm{d}x + \int \tilde{a}_i g \psi \mathrm{d}v \mathrm{d}x + \frac{W_r - W_l}{(x_r - x_l) \cdot n}
\]

where \( x_r \) and \( x_l \) are the coordinates of left and right cell centroids, and \( n \) is the normal vector of interface. By adding a penalty term in Equation (16), the whole scheme is essentially free from the odd-even decoupling phenomenon (Blazek 2015).

5. Sliding Mesh Method

To simulate the problem with the sliding interface, the computational domain is divided into rotating and stationary parts. The whole computational algorithm is shown as Algorithm 1, where the bold text is special treatments relating to the sliding interface. The detailed algorithm will be discussed in the following sections.

5.1. Establish Mortar by Polygon Clipping

To communicate the information between the rotating part and the stationary part, the mortar elements need to be established. As shown in Figure 1(a), the rotating and stationary parts overlap on the same circle but do not have common interfaces. So, new mortar interfaces, as shown by black dotted lines in Figure 1(b), need to be defined.
Algorithm 1 CGKS with sliding interface

1: Label the interface between the rotor and stator, and the neighbour cells of the interface
2: Initial rotation angle $\alpha = 0$
3: while the computation uncompleted do
4: calculate the time step $\Delta t$ according to CFL number
5: rotate the interface $\alpha$, establish mortars
6: for $i=1,2$ (for S2O4) do
7: define boundary condition for ghost cell
8: reconstruct the cell distribution except the adjoint cells of the sliding interface
9: reconstruct the adjoint cells of sliding interface
10: define boundary condition at boundary Gaussian point
11: evolution for interfaces
12: evolution for mortars
13: update cell average conservative values
14: add source term for the rotating domain
15: update cell average first-order spatial derivatives of conservative values except the adjoint cells of the sliding interface
16: update cell average first-order spatial derivatives of conservative values for the adjoint cells of the sliding interface
17: $\alpha = \alpha + \omega * \Delta t$
18: end for
19: end while

The in-house 3-D code based on prism mesh is used for the current simulation. The interfaces between the rotating part and the stationary part are surface meshes. The Sutherland–Hodgman algorithm (Sutherland and Hodgman 1974) is used for clipping the intersecting polygon of the two connected elements on the interface. This method is an effective and accurate algorithm for convex polygon clipping, which could deal with triangles, quadrangles, and so on. We triangulate the clipped polygon to make the algorithm easily adapt to different intersecting polygons.

As shown in Figure 1(a), the sliding interface is a cylindrical surface, but straight-edge meshes are used in the computation. We transform this interface into the cylindrical coordinate and consider that all the nodes have the same radius, so the polygon clipping process is done in $\theta - z$ plane. As shown in Figure 1(c), the black dotted lines show generated mortars by two interfaces, where the red line shows the inner part and the blue line the outer part.

5.2. Reconstruction for Sliding Mesh

To reconstruct the adjoint cells of the sliding interface, the same stencils are used for these cells. However, one of the interfaces is the sliding interface, so there is no directly joining neighbour cell at this interface. As shown in Figure 2, the right face of cell 0 is a sliding interface, where two cells (cell 4 and cell 5) joint with cell 0. Under this condition, a ghost cell is created by merging cell 4 and cell 5. For third-order reconstruction, the new constraints become

$$\int\int\int_{\Omega_4+\Omega_5} p^2 dV = \overline{Q}_4|\Omega_4| + \overline{Q}_5|\Omega_5|,$$
$$\int\int\int_{\Omega_4+\Omega_5} \frac{\partial}{\partial x} p^2 dV = (\overline{Q}_x)_4 |\Omega_4| + (\overline{Q}_x)_5 |\Omega_5|,$$
$$\int\int\int_{\Omega_4+\Omega_5} \frac{\partial}{\partial y} p^2 dV = (\overline{Q}_y)_4 |\Omega_4| + (\overline{Q}_y)_5 |\Omega_5|,$$
$$\int\int\int_{\Omega_4+\Omega_5} \frac{\partial}{\partial z} p^2 dV = (\overline{Q}_z)_4 |\Omega_4| + (\overline{Q}_z)_5 |\Omega_5|.$$
moved to the position of $0', 1', 2', 3'$. However, the governing equations in the related framework are used to simulate the rotation effect. So the cells are still in the position of 0, 1, 2, 3. To reconstruct cell 0, we need to rotate cells 4 and 5 to the position $4'$ and $5'$ with a transformation of both geometry information and cell-average values, including conservation values and their spatial derivatives.

5.3. Flux Evaluation via Mortars and Cell-Average Slope Update

After the mortars are created, the origin interface will be replaced by mortars. So we need to set Gaussian points (as shown by black dots in Figure 1c) in the mortars, and the left and right values of Gaussian points can be obtained by reconstruction. After reconstruction, the fluxes via mortar and point-wise conservative variables of Gaussian points on mortars can be updated.

6. Numerical Experiments

In the following cases, the three-dimensional solver is used to solve two-dimensional problems. Two layers and periodic boundary conditions are used in the $z$ direction. The time step is given by $\Delta t = \min \Delta t_i$, where $\Delta t_i$ is the time step defined in each cell

$$\Delta t_i = C_{\text{CFL}} \frac{h_i}{|V - U|_i + c_i + 2v_i/h_i},$$

where $C_{\text{CFL}}$ is the CFL number, $|V - U|_i$, $c_i$, and $v_i = (\mu/\rho)_i$ are the magnitude of related velocities, sound speed and kinematic viscosity coefficient of cell $i$. Here, we set the CFL number as 0.5 if no specified.

6.1. Isentropic Vortex Propagation

The isentropic vortex propagation problem is selected to test the solver for inviscid flow. The computation
domain is \([0, 1] \times [0, 1]\). The flow at time \(t\) is

\[
(U, V) = (U_0, V_0) + \frac{\kappa}{2\pi} e^{0.5(1-r^2)} (-\bar{y}, \bar{x}),
\]

\[
T = 1 - \frac{(\gamma - 1)\kappa^2}{8\gamma \pi^2} e^{1-r^2}
\]

\[
S = 1,
\]

\[
T = \frac{P}{\rho}, \quad S = \frac{P}{\rho\gamma},
\]

where the non-dimensional coordinate is \((\bar{x}, \bar{y}) = (\frac{x-x_r}{r_0}, \frac{y-y_r}{r_0})\), \(r_0 = 0.05\), the radius \(r = \sqrt{\bar{x}^2 + \bar{y}^2}\), and the vortex strength \(\kappa = 5\). The \((x_r, y_r)\) depends on time

\[
x_r = x_0 - U_0 t, \quad y_r = y_0 - V_0 t.
\]

In our simulation, we choose \((x_0, y_0) = (0.5, 0.5)\) and the background velocity \((U_0, V_0) = (1, 1)\). Periodic boundary conditions are applied in both the \(x\) and \(y\) directions.

The computation domain is divided into two parts: the rotating inner part with a radius of 0.2 and the stationary outer part. The angular speed of the rotating part is set as \(\omega = 2\pi\). Both the rotating and stationary cases are calculated to test the method. Four meshes with cell number 1754 \(\times\) 2, 7122 \(\times\) 2, 29324 \(\times\) 2, 120252 \(\times\) 2 are used, the time step is set as \(8e^{-4}, 4e^{-4}, 2e^{-4}, 1e^{-4}\) correspondingly. The coarsest mesh is shown in Figure 4.

To validate the accuracy of the scheme, the density error is defined at \(t = 1\), when the vortex has travelled for one period. The errors and numerical orders of the rotating and the stationary cases are shown in
Figure 4. The coarsest mesh used for isentropic vortex propagation (the inner black circle indicates sliding interface).

Table 1. ω = 2π rad/s.

| Mesh       | ErrorL₁ | O₁    | ErrorL₂ | O₂    | ErrorL∞ | O∞   |
|------------|---------|-------|---------|-------|---------|------|
| 1754 × 2  | 5.80E−04| 7.89E−03| 3.97E−01|       |         |      |
| 7122 × 2  | 2.23E−04| 2.87E−03| 1.46E−01| 1.47E−01| 1.44E−01|      |
| 29324 × 2 | 4.35E−05| 5.23E−04| 2.46E−02| 2.20E−02| 2.74E−02|      |
| 120252 × 2| 7.15E−06| 6.99E−05| 2.90E−03| 3.94E−03| 2.48E−03|      |

Tables 1 and 2. For both the stationary and rotating cases, the numerical orders are close to the theoretical third-order accuracy; and the error of the rotating case is slightly larger than that of the stationary case.

The density and x-velocity contours of the rotating case by the finest mesh at $t = 0.1 \sqrt{2}$ are shown in Figure 5, when the centre of the vortex is located on the sliding interface. No distortion can be observed in the contour, which implies that the vortex can propagate through the sliding interface without reflection and deformation.

6.2. Flow Over a Rotating Ellipse Cylinder

This case is selected to verify the method for subsonic viscous flow. The income flow is set as $\rho_\infty = 1.0, p_\infty = 1/\gamma, U_\infty = 0.05$, which has a Mach number of 0.05. The ellipse, with a major axis length of $A = 1.0$ and a minor axis length of $B = 0.5$, rotates counterclockwise at an angular speed of $\omega = \pi / 40$. The Reynolds number based on the length of the ellipse major axis and the incoming velocity is $Re = 200$. The sliding interface is located at $r = 1.5$. The computation mesh is plotted in Figure 6 with total 31516 × 2 elements. 160 nodes are used to discretise the ellipse, and the height of the first layer near the ellipse is $5 \times 10^{-3}$. The adiabatic non-slip wall is set on the ellipse surface, and the far-field boundary condition is set at the outer boundary.

The lift and drag coefficients in one period are shown in Figure 7. The present numerical result is plotted by line, and reference data by Zhang and Liang (2015) is shown by symbols. The lift and drag coefficients agree well with the reference data. Also, the
vorticity contours and streamlines at different times in one period are plotted in Figure 8. It can be observed that a clockwise vortex and a counterclockwise vortex generate around the ends of the ellipse. From the time $t = 0$ to $t = 3/8T$, the clockwise vortex sheds off from the leading edge and hits the trailing edge. While from time $t = 1/2T$ to time $t = 7/8T$, a counterclockwise vortex slowly emerges and then goes downstream without reattaching to the ellipse. This makes the flow not fully symmetric in a periodic cycle. And the whole process repeats as the ellipse rotates.

### 6.3. Stirred Tank

This case is a 2-D laminar case as proposed in Zhang and Liang (2015). The computational domain is composed of several parts: a $r = 0.5$ cylinder located on the original point, an outer wall with a radius of 5, six uniformly distributed agitating blades with a thickness of 0.1, each extending from $r = 1$ to $r = 2$, four baffles with same thickness and height 1 installed on the outer wall. The computational domain is split into two parts, an inner rotating part, and an outer fixed part. The sliding interface is located at $r = 3$.

The initial condition is set as $\rho_0 = 1.0, p_0 = 1.0/\gamma, u = v = 0$. The inner part with the cylinder and six blades rotates at angular speed $\omega = 0.2$, so the Mach number defined by cylinder surface velocity is $M_i = 0.05$. And the Reynolds number defined by the diameter of the inner cylinder and the angular speed is $\text{Re} = \rho \omega d^2/\mu = 100$. Nonslip wall boundary conditions are applied to all boundaries. The adiabatic wall boundary condition is adopted on the six blades, and the isothermal wall condition is used on other walls. The mesh used in the computation is plotted in Figure 9, with $1024^2 \times 2$ elements.

The density contours at different times are plotted in Figure 10. At the time $\omega t = 0.25$, the fluid is pushed and squeezed by blades, so large fluctuations can be observed. Soon the fluid becomes very chaotic due to the baffles and the outer wall. At the time $\omega t = 1.5$, vortical structures are generated by flow passing the...
Figure 8. The vorticity contours and streamlines at different in a periodic: (a) $t = 0$, (b) $t = \frac{1}{8}T$, (c) $t = \frac{1}{4}T$, (d) $t = \frac{3}{8}T$, (e) $t = \frac{1}{2}T$, (f) $t = \frac{5}{8}T$, (g) $t = \frac{7}{8}T$, (h) $t = \frac{3}{4}T$.

Figure 9. Mesh used in stirred tank case (red dotted lines indicate sliding interfaces).

baffles and associated with bouncing pressure waves, blades-induced vortices, unsteady boundary layers, etc. As the blade continues to rotate for a longer time, the chaotic flow structure slowly dissipates, and the flow structure becomes organised. Finally, the flow field reaches a quasi-steady state in the rotating reference framework and changes little with time. The density variation becomes smaller and smaller, and the contour closes to uniform in the circumferential direction. The radial gradient of density is caused by centrifugal force.

6.4. $Ma = 3$ Cylinder

A steady supersonic flow is used to show the influence of sliding mesh in discontinuous flow. There is
a cylinder with radius $r = 0.5$ located at centre, and the computational domain is a cylindrical domain with radius $R = 10$. The whole domain is divided into three parts: from $R_1 = 0.5$ to $R_2 = 1$, from $R = 1$ to $R = 1.5$ and from $R = 1.5$ to $R = 10$. In the simulation, the second part rotates at angular speed $\omega = 1$, so the sliding interfaces are located at $R = 1$ and $R = 1.5$. And zero angular speed $\omega = 0$ case is also calculated for comparison. The inviscid slip wall boundary condition is used on the cylinder surface, and the far-field boundary condition with income Mach number 3 is used on the outer boundary. As shown in Figure 11, total $6320 \times 2$ elements are used in the simulation, and red dotted lines indicate the sliding interfaces. As shown in Figure 12, the Mach number and static pressure contours are plotted. No oscillation can be observed near the shock. Overall, no rotating and rotating mesh cases agree well with each other. In the Mach number contours, slight asymmetry can be observed in the wake of the cylinder, where the density and pressure are very low, and the non-linear weights become sensitive to the local geometry.

### 6.5. Three Cylinders Rotating at Supersonic Speed

This is an unsteady case with complicated shock interactions, which was used to illustrate the applicability of the diffuse interface model (Kemm et al. 2020). The computational domain is $[-2,2] \times [-2,2]$. Three cylinders with radius $r_i = 0.2$ are located at position...
Figure 11. Mesh used in $Ma=3$ cylinder (red dotted lines indicate sliding interfaces).

Figure 12. The Mach number and pressure contours of $Ma=3$ cylinder case from the calculations with mesh rotating mesh (upper part) and non-rotating mesh (lower part).

Figure 13. Mesh for three cylinders rotating in a compressible gas at supersonic speed.

$(x, y) = (R_0 \cos \psi_i, R_0 \sin \psi_i)$, where $R_0 = 1.0$. These cylinders rotate clockwise at the same angle speed $\omega = 3$, which yields a Mach 3 speed at the location of the cylinders. The sliding interface is at the position of $r = 1.5$. The periodic boundary condition is applied in the $x$ and $y$ directions. The mesh is plotted in Figure 13 and total $90036 \times 2$ elements are used.

The density contours are shown in Figure 14. The shocks emerge in the front of cylinders and then interact with the trailing wake. No unphysical oscillation can be observed on the sliding interface, which indicates that the proposed method can deal with moving shocks well.

In Table 3, we show the total computational time and interface communication time for 100-step computation. The sliding interface of each side has $600 \times 2$
Figure 14. The density contour of three cylinders rotating in a compressible gas at supersonic rotating speed (the red dotted cycle indicates sliding interface): (a) $t = 0.35$, (b) $t = 0.5$, (c) $t = 0.75$ and (d) $t = 1.75$.

Table 3. Computation time and interface communication time.

| Total time | Communication time | Percentage |
|------------|--------------------|------------|
| 912        | 44                 | 4.82%      |

cell interfaces, about 1.33% of the total cell, and the communication time accounts for 4.82% of total time.

7. Conclusion

In this paper, a third-order CGKS is developed in a rotating coordinate frame with the combination of the sliding mesh method for simulating flow problems with rotating parts. Due to the kinetic nature of the gas evolution model, the dynamic effect from centrifugal and Coriolis forces in the rotating frame can be easily incorporated into the time-accurate flux function and flow variable updates at a cell interface. As a result, both cell-averaged flow variables and their gradients can be updated and used in the high-order compact reconstruction. The high-order and compactness of the scheme have advantages for flow simulation with rotating parts in capturing the unsteady flow passing through the sliding interface. The current CGKS can use a large CFL number, such as CFL number 0.5, in the determination of time step in the flow simulation with highly compressible shock waves. Many test cases, covering viscous and inviscid, subsonic and supersonic ones, are used to validate the scheme. The numerical performance of the scheme in the density wave propagation, vortex flow, shock passing through sliding interfaces, and rotating cylinders at supersonic speed, shows the accuracy and robustness of the high-order method. The current scheme can be extended straightforwardly to the three-dimensional case. In the coming work, large-scale three-dimensional flow
computations, such as propeller noise and wake–shock interactions in the transonic compressor, will be presented. At the same time, the parallel technology will be further developed to improve computational efficiency in 3D applications. The high efficiency of the scheme can be easily realised because of the compactness of the stencils.

**Disclosure statement**

No potential conflict of interest was reported by the author(s).

**Funding**

This work was supported by National Key R&D Program of China (Grant Nos. 2022YFA1004500), National Natural Science Foundation of China (12172316), and Hong Kong Research Grant Council (16208021, 16301222).

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Defining \( L_\alpha = \int \psi_\alpha \left( \frac{\partial}{\partial t} + w_1 \frac{\partial}{\partial x_1} - \epsilon_{ij} \Omega_i v_j \frac{\partial}{\partial x_j} \right) g d\mathcal{V} \) and considering

\[
\epsilon_{ij} \Omega_i \int \psi_\alpha \frac{\partial g}{\partial v_1} d\mathcal{V} = \epsilon_{ij} \Omega_i \int \frac{\partial (v_1 \psi_\alpha)}{\partial v_1} d\mathcal{V} - g \frac{\partial \psi_\alpha}{\partial v_1} d\mathcal{V} = \epsilon_{ij} \Omega_i \int -g (\delta_{ij} \psi_\alpha + v_j \frac{\partial \psi_\alpha}{\partial v_1}) d\mathcal{V} = -\epsilon_{ij} \Omega_i \int g v_j \frac{\partial \psi_\alpha}{\partial v_1} d\mathcal{V} = -\epsilon_{ij} \Omega_i \int g v_j \frac{\partial \psi_\alpha}{\partial v_1} d\mathcal{V} = -\epsilon_{ij} \Omega_i \int g v_j \frac{\partial \psi_\alpha}{\partial v_1} d\mathcal{V},
\]

the \( L_\alpha \) becomes

\[ L_\alpha = < \psi_\alpha > + \epsilon_{ij} \Omega_i < v_j \frac{\partial \psi_\alpha}{\partial v_1} > = o(\epsilon) \]

We can get

\[ L_1 = \rho_1 + (\rho W_i)_l, \]

for \( a = 2, 3, 4 \)

\[ L_a = (\rho V_a)_j + (\rho V_a W_i)_j + p_{a} + \rho \epsilon_{ab} \Omega_i V_j, \]

and

\[ L_5 = \left( \frac{1}{2} \rho V_a^2 + \frac{K + 3}{2} p \right) \frac{\partial}{\partial t} + \left( \frac{1}{2} \rho V_a W_k + \frac{K + 5}{2} \rho W_k \right) \frac{\partial}{\partial x_k} + U_k p_{k} \frac{\partial}{\partial x_k} \]

Then, we have

\[ \rho V_a + \rho W_a V_j + p_{a} + \rho \epsilon_{ab} \Omega_i V_j = o(\epsilon), \]

\[ K + 3 \frac{p_{a}}{2} + \rho W_k \frac{\partial}{\partial x_k} + \frac{K + 5}{2} \rho W_k \frac{\partial}{\partial x_k} = o(\epsilon). \]

With \( \hat{r} = \hat{r}(t, x_i) \), and \( R_a \) and \( S_a \) defined as

\[ R_a = \int \psi_\alpha \left( \frac{\partial}{\partial t} + w_1 \frac{\partial}{\partial x_1} - \epsilon_{ij} \Omega_i \frac{\partial}{\partial v_1} \right) \frac{\partial g}{\partial v_1} d\mathcal{V}, \]

\[ S_a = -\epsilon_{ij} \Omega_i \int \psi_\alpha \frac{\partial \psi_\alpha}{\partial v_1} \frac{\partial g}{\partial v_1} - \epsilon_{kmn} \frac{\partial g}{\partial v_k} \partial v_n \Omega_m \frac{\partial}{\partial v_k} \frac{\partial g}{\partial v_1} d\mathcal{V}, \]

\[ \hat{r}_a = \int \psi_\alpha \left( \frac{\partial}{\partial t} + w_1 \frac{\partial}{\partial x_1} - \epsilon_{ij} \Omega_i \frac{\partial}{\partial v_1} \right) \frac{\partial g}{\partial v_1} d\mathcal{V} = \epsilon_{ij} \Omega_i \int \psi_\alpha \frac{\partial \psi_\alpha}{\partial v_1} \frac{\partial g}{\partial v_1} - \epsilon_{kmn} \frac{\partial g}{\partial v_k} \partial v_n \Omega_m \frac{\partial}{\partial v_k} \frac{\partial g}{\partial v_1} d\mathcal{V}, \]

we have \( S_1 = S_5 = 0 \) for \( \alpha = 1 \) or \( \alpha = 5 \), and \( S_\alpha = -\epsilon_{a} \Omega_i L_j = o(\epsilon) \) for \( \alpha = 2, 3, 4 \). So, the \( R_\alpha \) becomes

\[ R_\alpha = (\hat{r}_L)_\alpha + S_\alpha + \frac{\partial}{\partial x_i} \int \hat{r} \psi_\alpha \left( \frac{\partial}{\partial t} + w_1 \frac{\partial}{\partial x_1} - \epsilon_{kmn} \Omega_m \frac{\partial}{\partial v_k} \right) g d\mathcal{V} \]

Firstly, for \( \alpha = 1 \)

\[ R_1 = (\hat{r}_L)_j + o(\epsilon) = o(\epsilon). \]

The equation becomes

\[ \rho_1 + (\rho W_i)_l = o(\epsilon^2). \]

For \( a = 2, 3, 4 \)

\[ R_a = \left\{ \hat{r} \left[ < v_a \psi_\alpha > + \epsilon_{kmn} < v_n \psi_\alpha > \Omega_m \right] \right\}_j + o(\epsilon), \]

with the consideration of Equation (A3), we can get

\[ R_a = \left\{ \hat{r} \left[ p(V_a + V_\alpha) - \frac{2}{3} V_{k, k} \delta_{al} + \frac{2}{3} K \left( K + 3 \right) p V_{k, k} \delta_{al} \right] \right\}_j + o(\epsilon). \]

And the method can be used for \( \alpha = 5 \)

\[ R_5 = \left\{ \hat{r} \left[ < v_5 \psi_\alpha > + \epsilon_{kmn} < v_n \psi_\alpha > \Omega_m \right] \right\}_j + o(\epsilon) \]

\[ = \left\{ \hat{r} \left[ \frac{K + 5}{2} p \left( \frac{p_{a}}{p} \right) \frac{\partial}{\partial t} + p \left[ \frac{2}{K + 3} V_{k, k} \delta_{al} + V_{k, k} V_{k, k} \right] \right] \right\}_j + o(\epsilon). \]

In conclusion, by dropping \( o(\epsilon^2) \) terms in Equation (A2), the Navier–Stokes equations can be derived as follows:

\[ \rho_\tau + (\rho W_i)_k = 0, \]

\[ (\rho V_j)_l + (\rho V_j W_k + p \delta_{jk} - \sigma_{jk})_k = -\epsilon_{kmn} \Omega_i V_m, \]

\[ (\rho E)_l + (\rho W_k + p U_k - \kappa T_k - V_i \sigma_{ij})_k = 0, \]

\[ \rho E = \frac{1}{2} \rho V_a^2 + \frac{K + 3}{2} p, \rho H = \rho E + p, T = \frac{m}{k} \rho, \]

where \( E \) is the total energy, \( H \) is the enthalpy, \( T \) is the temperature, \( k \) is the Boltzmann constant, \( m \) is the mass of a molecule, and \( \sigma_{jk} \) is the stress tensor, which is defined by

\[ \sigma_{jk} = \mu \left( V_j \delta_k + V_k \delta_j - \frac{2}{3} V_{l, l} \delta_{jk} \right) + \beta V_{l, l} \delta_{jk}, \]

where \( \mu = \tau \rho \) is the dynamic viscosity coefficient and

\[ \beta = \frac{2}{3} K + 3 \frac{\mu}{m}. \]
is the bulk viscosity coefficient. The thermal conductivity coefficient $\kappa$ is given by

$$\kappa = \frac{K + 5}{2} \frac{k}{m} \tau p.$$  

In addition, the equations can be written in terms of $\gamma$ instead of $K$ by using $K = (5 - 3\gamma)/(\gamma - 1)$ for three-dimensional gas flow. The thermal conductivity becomes

$$\kappa = \frac{\gamma}{\gamma - 1} \frac{k}{m} \tau p = C_p \mu,$$

where $C_p$ is the specific heat capacity at constant pressure and the Prandtl number is 1.

**Appendix 2. Moments of the Maxwellian Distribution Function**

In GKS, the moments of the Maxwellian distribution function with bounded and unbounded integration limits need to be evaluated, and the unbounded integration can refer to Xu (2001). However, when dealing the moving interface, the integration boundary change from 0 to $U$, which means we need to calculate $\int_{-\infty}^{0} g^v n d\Sigma$ and $\int_{0}^{\infty} g^v n d\Sigma$, denoting as $<v^0 >_{v < U}$ and $<v^n >_{v > U}$. Through the integration by part, the moments are

$$<v^0 >_{v > U} = \frac{1}{2} \text{erfc}(-\sqrt{\lambda} W),$$

$$<v^1 >_{v > U} = <v^0 >_{v > U} + \frac{1}{2\sqrt{\pi \lambda}} e^{-\lambda W^2},$$

$$<v^n + 2 >_{v > U} = V <v^n + 1 >_{v > U} + \frac{n + 1}{2\lambda} <v^n >_{v > U} + \frac{1}{2\sqrt{\pi \lambda}} U^{n+1} e^{-\lambda W^2},$$

and

$$<v^0 >_{v < U} = \frac{1}{2} \text{erfc}(-\sqrt{\lambda} W),$$

$$<v^1 >_{v < U} = <v^0 >_{v > U} - \frac{1}{2\sqrt{\pi \lambda}} e^{-\lambda W^2},$$

$$<v^n + 2 >_{v > U} = V <v^n + 1 >_{v > U} + \frac{n + 1}{2\lambda} <v^n >_{v > U} - \frac{1}{2\sqrt{\pi \lambda}} U^{n+1} e^{-\lambda W^2},$$

where $W = V - U$ and $V$ is the macroscopic velocity of fluid element.