A true-direction reconstruction of the quiet direct simulation method for inviscid gas flows


1. Introduction

There are a number of approaches for the simulation of gas flows, depending on the nature and level of rarefaction of the flow. Computational fluid dynamics (CFD) typically uses the finite volume method to solve a set of discretized governing equations, usually the Euler or Navier–Stokes equations for continuum flows. Contemporary finite-volume CFD divides the computational domain into a grid of cells, and fluxes of mass, momentum, and energy are calculated through the interfaces between these cells. This technique may suffer from the major disadvantage that the poor alignment of the grid with the flow field may result in large errors for some important flows (e.g., explosive blast wave), since fluxes can only occur between cells that share an interface, i.e., no reflection of the true-direction nature of the gas flow. Thus, CFD requires a careful grid design to ensure accurate results, convergence, and stability.

Since the development of direct simulation Monte Carlo (DSMC) by Bird [1] for statistically solving the Boltzmann equation, a large number of continuum kinetic theory-based schemes have emerged following a similar spirit. In 1980, Pullin [2] proposed the equilibrium flux method (EFM) as an analytical equivalent to the equilibrium particle simulation method (EPSM), which is a direct simulation method where particles are forced to assume the Maxwell–Boltzmann equilibrium velocity probability distribution function instead of performing collisions. Later, Smith et al. [3] proposed a general form of the EFM method known as the true-direction equilibrium flux method (TDFEM), which captures relatively accurately the transport mechanism employed by EPSM. Fluxes calculated by TDFEM represent the true analytical solution to the molecular free flight problem, under the assumptions of thermal equilibrium and uniformly distributed quantities in each cell.
2.1. Quiet direct simulation (QDS)

Albright et al. [4] developed a numerical scheme for the solution of the Euler equations, known as the quiet direct simulation Monte Carlo (QDSMC) method. In this method, the integrals encountered in the TDEFM formulation are replaced by approximations using Gaussian numerical integration, effectively replacing the continuous velocity distribution function with a series of discrete velocities. The method was later renamed the quiet direct simulation (QDS) method, because of the lack of stochastic processes, and was extended to second-order spatial accuracy [5]. The lack of complex mathematical functions results in a computationally very efficient scheme with a considerably higher performance than EFM while maintaining the advantages of true-directional fluxes like TDEFM.

Because of the assumption of unrestricted motion during free flight, each of the abovementioned kinetic solvers has a large amount of (cell-size-based) numerical diffusion. To combat this dissipation, a common strategy, employed in conventional finite volume methods, is to apply the higher-order reconstruction of properties or fluxes. Macrossan [6] applied EFM using higher-order spatial extensions, while Smith [7] attempted the analytical inclusion of gradients into true-direction volume-to-volume fluxes, only to find that the complete analytical inclusion of gradient terms in the TDEFM flux expressions is impossible. Smith et al. [5] reduced the numerical diffusion by applying “simplified” flux reconstruction at the interface that improves the original QDS to be almost second order in spatial accuracy; this method is called QDS-2N.

In this paper, we extend the second-order QDS algorithm (QDS-2N) [5] to higher-order reconstruction through the true-direction polynomial multi-dimensional reconstruction of conserved properties across each cell width; this method is called QDS-N². The net fluxes are computed through the individual contributions of QDS particles, computed by taking moments over the polynomial reconstruction. The particle properties are updated, considering the average value of the conserved quantity between the region bounds, which are required in translational directions and the application of splitting. The fluxes of conserved properties are calculated by a sum of weighted moments over the polynomial spatial reconstruction of mass, momentum, and energy across the cell width. The verification simulations of four two-dimensional cases are carried out to show the improved accuracy of the proposed QDS-N² scheme for inviscid gas flow simulations.

2. Numerical method

2.1. Quiet direct simulation (QDS)

The normal random variable \( N(0, 1) \) is defined by the probability density:

\[
p(x) = \frac{e^{-x^2/2}}{\sqrt{2\pi}}.
\]  

(1)

By using a Gaussian quadrature approximation, the integral of Eq. (1) over its limits can be approximated by:

\[
\int_{-\infty}^{\infty} e^{-x^2/2} f(x)dx \approx \sum_{j=1}^{N} w_j f(q_j)
\]

(2)

where \( w_j \) and \( q_j \) are the weights and abscissas of the Gaussian quadrature (also known as the Gauss–Hermite parameters), and \( N \) is the number of terms. The abscissas are the roots of the Hermite polynomials, which can be defined by the recurrence equation:

\[
H_{n+1}(q) = 2qH_n(q) - 2nH_{n-1}(q)
\]

(3)

where \( H_{-1} = 0 \), and \( H_0 = 1 \). The weights can be determined from:

\[
w_j = \frac{2^{n-1}n!\sqrt{\pi}}{n^n [H_{n-1}(q_j)]^2}.
\]

(4)

The net fluxes of mass, momentum and energy of a cell are given by the sum of individual flux contributions from all the particles flowing in and out as follows:

\[
F_{\text{MASS}} = \left( \sum_{i=1}^{M} F_{\text{MASS}}^i \right)_{\text{in}} - \left( \sum_{i=1}^{N} F_{\text{MASS}}^i \right)_{\text{out}},
\]

\[
F_{\text{MOM}} = \left( \sum_{i=1}^{M} F_{\text{MOM}}^i \right)_{\text{in}} - \left( \sum_{i=1}^{N} F_{\text{MOM}}^i \right)_{\text{out}},
\]

\[
F_{\text{ENG}} = \left( \sum_{i=1}^{M} F_{\text{ENG}}^i \right)_{\text{in}} - \left( \sum_{i=1}^{N} F_{\text{ENG}}^i \right)_{\text{out}}.
\]

(5)

where \( F_{\text{MASS}}^i, F_{\text{MOM}}^i \) and \( F_{\text{ENG}}^i \) is the individual mass flux, individual momentum flux and individual energy flux from particle \( j \) respectively, \( M \) and \( N \) is the number of inflow and outflow particles respectively into the cell under consideration. Each of the individual contributions (with first order spatial accuracy) can be described by the expressions, e.g., in one-dimensional case:

\[
F_{\text{MASS}} = \frac{v_j \Delta t}{\Delta x} m_j
\]

\[
F_{\text{MOM}} = \frac{v_j \Delta t}{\Delta x} m_j v_j
\]

\[
F_{\text{ENG}} = \frac{1}{2} \left( v_j^2 + v_j + \epsilon_j \right)
\]

(6)

where the particle mass \( m_j \), particle velocity \( v_j \), and particle internal energy \( \epsilon_j \) are expressed as:

\[
m_j = \frac{\rho \Delta x w_j}{\sqrt{\pi}}, \quad v_j = u + \sqrt{2\sigma q_j}
\]

(7)

where \( \rho \) is the density, \( u \) is the bulk (or mean) flow velocity, and \( \sigma = (RT)^{1/2} \) in a given source cell. Note \( R \) is the universal gas constant, and \( T \) is the gas temperature. The total number of degrees of freedom \( \xi \) is defined as \( \xi = 2(\gamma - 1) \) where \( \gamma \) is the specific heat ratio (equals \( C_p/C_v \)), and \( \Omega \) is the number of simulated degrees of freedom (e.g., \( \Omega = 1 \) for one dimensional flow). In the existing QDS-2N [5], the values of \( \rho, u, \) and \( \sigma \) employed in QDS particle initialization are taken from reconstructions based on linear variations between neighbor cells. Despite fluxes being true direction in nature, the reconstructions performed in previous implementations are directionally decoupled—i.e., a flux is computed through the product of (separate) fluxes previously computed (for 2D flow) in the \( x \) and \( y \) directions. For the 2D case, the particle mass and velocities in Eq. (7) become:

\[
m_{jk} = \frac{\rho \Delta x \Delta y w_j w_k}{\pi} \quad v_j = u_x + \sqrt{2\sigma^2 q_j}
\]

(8)

\[
v_k = u_y + \sqrt{2\sigma^2 q_k}
\]

where there are \( K = 1, \ldots, M \) particles in the \( y \)-direction and the definition of other variables are the same as those in 1D case. The internal energy remains identical to the 1-D case, allowing for a corresponding increase in \( \Omega \) to account for the extra simulated dimension. The fluxes from sources cell to any arbitrary destination cell can be calculated by the particle position distributions. The fluxes of mass, momentum and energy, which are based on the proportion of the overlapped area to the area of the original cell, are given by:

\[
F_{\text{MASS}} = \frac{A_x}{A_s} m_{jk} \quad F_{\text{MOM}} = \frac{A_s}{A_s} m_{jk} v_j
\]

(9)

\[
F_{\text{ENG}} = \frac{A_s}{A_s} m_{jk} \left[ \frac{1}{2} (v_j^2 + v_k^2) + \epsilon_{jk} \right]
\]


where $A$ is the overlapped area as $u \cdot v \cdot dt^2$ and $A_i$ is the source cell area as $dx \cdot dy$.

### 2.2. Spatial reconstruction and flux calculation

In the current study, referring to Fig. 1, the general extension to higher order in QDS in one-dimensional case is performed using a spatial reconstruction of the form:

$$
Q_s(x) = Q_c + \frac{dQ}{dx} \bigg|_{x=0.5\Delta x} (x - 0.5\Delta x)
+ \left( \frac{d^2Q}{dx^2} \bigg|_{x=0.5\Delta x} (x - 0.5\Delta x)^2 \right)
+ \cdots + \left( \frac{d^{n-1}Q}{dx^{n-1}} \bigg|_{x=0.5\Delta x} (x - 0.5\Delta x)^{n-1} \right)
+ \frac{1}{n!} \left( \frac{d^nQ}{dx^n} \bigg|_{x=0.5\Delta x} (x - 0.5\Delta x)^n \right)$$

where $Q_c(x)$ is the value of a conserved property (mass, momentum, or energy) at a distance $x$ from the left hand side of the cell, and integer $n$ indicates the order of the reconstruction. Note $Q_c(x)$ is the value of $Q(x)$ at the cell center. This value is calculated from $Q(x)$ integrating over the cell width divided by the cell width equaling to the existing average value of the source cell $Q_s$ presented below:

$$
Q_s = \frac{1}{\Delta x} \int_{x_L}^{x_R} Q(x) dx
= Q_c + \left( \frac{dQ}{dx} \bigg|_{x=0.5\Delta x} (x - 0.5\Delta x)^2 \right)
+ \cdots + \left( \frac{d^{n-1}Q}{dx^{n-1}} \bigg|_{x=0.5\Delta x} (x - 0.5\Delta x)^{n-1} \right)
+ \frac{1}{n!} \left( \frac{d^nQ}{dx^n} \bigg|_{x=0.5\Delta x} (x - 0.5\Delta x)^n \right)
$$

By using our revised reconstruction, the bounds of integration are $X_L = 0$ and $X_R = \Delta x$. Then, Eq. (11) leads to:

$$
Q_s = Q_c + \frac{\Delta x^2}{24} \left( \frac{d^2Q}{dx^2} \right)_{x=0.5\Delta x} + \frac{\Delta x^4}{1920} \left( \frac{d^4Q}{dx^4} \right)_{x=0.5\Delta x}
+ \cdots + \left( \frac{d^{n-1}Q}{dx^{n-1}} \right)_{x=0.5\Delta x}
+ \frac{1}{n!} \left( \frac{d^nQ}{dx^n} \right)_{x=0.5\Delta x}
$$

Alternatively, $Q_c$ can be expressed as follows:

$$
Q_c = \overline{Q}_s - \frac{\Delta x^2}{24} \left( \frac{d^2Q}{dx^2} \right)_{x=0.5\Delta x} - \frac{\Delta x^4}{1920} \left( \frac{d^4Q}{dx^4} \right)_{x=0.5\Delta x}
- \cdots - 2\Delta x^{n-1} \left( \frac{1}{2} \frac{d^nQ}{dx^n} \right)_{x=0.5\Delta x}
$$

where $n$ is assumed to be an odd number. This shows that the correction is only required when the scheme is third order ($n = 3$) accurate or higher, otherwise $Q_c = \overline{Q}_s$ (e.g., $n = 2$). Thus, the complete correct form of the higher order reconstruction of $Q(x)$ using $\overline{Q}_s$ contains additional terms on every even derivative:

$$
Q(x) = \left( \overline{Q}_s - \frac{\Delta x^2}{24} \left( \frac{d^2Q}{dx^2} \right)_{x=0.5\Delta x} - \frac{\Delta x^4}{1920} \left( \frac{d^4Q}{dx^4} \right)_{x=0.5\Delta x} \right)
+ \left( \frac{dQ}{dx} \right)_{x=0.5\Delta x} \left( -0.5\Delta x \right)^2
+ \frac{1}{3!} \left( \frac{d^3Q}{dx^3} \right)_{x=0.5\Delta x} \left( -0.5\Delta x \right)^3
+ \frac{1}{n!} \left( \frac{d^nQ}{dx^n} \right)_{x=0.5\Delta x} \left( -0.5\Delta x \right)^n.
$$

Specifically, as $n = 2$, the above is reduced to the following form because of $Q_c = \overline{Q}_s$, as shown below:

$$
Q(x) = \overline{Q}_s + \frac{dQ}{dx} (x - 0.5\Delta x) + \frac{1}{2} \frac{d^2Q}{dx^2} (x - 0.5\Delta x)^2
$$

The above reduces to exactly the same form as in QDS-2N [5].

Next, referring to Fig. 2, the outgoing flux value of average conserved quantity successfully moving from the source cell into the destination cell (denoted by $Q_{tr}$) is:

$$
Q_{tr} = \frac{1}{(X_R - X_L)} \int_{x_L}^{x_R} Q(x) dx
= \frac{1}{(X_R - X_L)} \left[ \left. Q_c + \frac{1}{2} \frac{dQ}{dx} (x - 0.5\Delta x)^2 \right|_{x=X_L}^{x=X_R} \right]
\times (x - 0.5\Delta x)^2 + \ldots
$$

$$
= Q_c + \frac{1}{(X_R - X_L)} \left[ \left. \frac{1}{2} \frac{dQ}{dx} \right|_{x=X_L}^{x=X_R} \right]
\times (x - 0.5\Delta x)^2 + \ldots
$$

where the bounds of integration are $X_L = \Delta x - u.t$ and $X_R = \Delta x$.

Now, the transition of mean values $\overline{Q}_{tr}$ can be used to calculate particle properties. Assigning the flux out of average conserved properties $\overline{Q}_{1tr}$, $\overline{Q}_{2tr}$, and $\overline{Q}_{3tr}$, as the mass, momentum, and energy, respectively, the resulting QDS particle properties for particle $f$ are:

$$
m_f = \frac{\overline{Q}_{1tr} W_f}{\sqrt{\pi}}
$$

$$
v_f = \frac{\overline{Q}_{2tr}}{Q_{1tr}} + \left[ \frac{2r}{(C_v)^2} \left( \frac{\overline{Q}_{2tr}}{\overline{Q}_{1tr}} - \frac{1}{2} \left( \frac{\overline{Q}_{2tr}}{\overline{Q}_{1tr}} \right)^2 \right) \right]^{1/2}
$$

$$
\epsilon_f = \frac{(\xi - \Omega)}{2} \left( \frac{\overline{Q}_{2tr}}{\overline{Q}_{1tr}} - \frac{1}{2} \left( \frac{\overline{Q}_{2tr}}{\overline{Q}_{1tr}} \right)^2 \right).
$$

$$
$$
To calculate the average value of conserved property for higher order reconstruction, it is important how the flux limiting is coupled. According to the value of conserved property $Q(x)$ (see Eq. (10)), the gradient of $Q_c$ is defined in flux limiting during the reconstruction process. In each cell, we employ the monotonized central difference (MC) limiter to the effective gradients of conserved properties, as described below:

$$\left( \frac{dQ}{dx} \right)_F = \left( \frac{dQ}{dx} \right)_F \phi(\theta)$$  \hspace{1cm} (18)

$$\phi(\theta) = \max \left[ 0, \min \left( 2, \frac{\theta + 1}{2}, 2\theta \right) \right]$$  \hspace{1cm} (19)

where $\phi$ is the equivalent flux limiter and $F$ is the gradient calculated using forward differences. The $\theta$ is the ratio of the first order gradient calculated using forward and backward differences:

$$\theta = \left( \frac{dQ}{dx} \right)_B \left( \frac{dQ}{dx} \right)_F^{-1}.$$

Therefore, an alternate representation of the variation of $Q(x)$ over space can be:

$$Q(x) = Q_c + \left[ \left( \frac{dQ}{dx} \right)_F \phi(\theta) \right]_{x=0.5\Delta x} (x - 0.5\Delta x)$$

$$+ \frac{1}{2} \left[ \left( \frac{d^2Q}{dx^2} \right)_F \phi(\theta) \right]_{x=0.5\Delta x} (x - 0.5\Delta x)^2$$

$$+ \cdots + \frac{1}{n!} \left[ \left( \frac{d^nQ}{dx^n} \right)_F \phi(\theta) \right]_{x=0.5\Delta x} (x - 0.5\Delta x)^n.$$

(21)

Fig. 3. Flowchart describing QDS particle computation with gradient inclusion.

Fig. 4. Two-dimensional motion of a single QDS particle showing “sub-particle” contributions.

2.3. 1D flux calculation and implementation

In QDS simulations, we require flux from a volume to another volume. Since fluxes are split, the quantities of the fluxes depend entirely on the region from which they originated. The flux calculation is described as a flowchart in Fig. 3, and summarized briefly as follows:

1. The gradients of conserved properties $Q$ are first calculated using standard finite difference approximations in each cell $i$. For example, for a 5th order accurate reconstruction, one might use the...
stencils like:

\[
\begin{align*}
\frac{dQ}{dx}_i &= \left( \frac{Q_{i+1} - Q_{i-1}}{\Delta x} \right) \\
\frac{d^2Q}{dx^2}_i &= \frac{1}{\Delta x^2} \left( \left( \frac{dQ}{dx} \right)_{i+1} + \left( \frac{dQ}{dx} \right)_{i-1} - 2 \left( \frac{dQ}{dx} \right)_{i} \right) \\
\frac{d^3Q}{dx^3}_i &= \frac{1}{\Delta x^2} \left( \left( \frac{d^2Q}{dx^2} \right)_{i+1} + \left( \frac{d^2Q}{dx^2} \right)_{i-1} - 2 \left( \frac{d^2Q}{dx^2} \right)_{i} \right).
\end{align*}
\] (22)

2. For each QDS particle:

a. Calculate the approximate particle velocity based on the current cell \( \overline{Q}_i \), which should give the same particle velocity as 1st order QDS.

\[
v_j = \frac{\overline{Q}_{i+2}}{\overline{Q}_{i+1}} + \left[ \frac{2R}{C_p^2} \left( \frac{\overline{Q}_{i+3}}{\overline{Q}_{i+1}} - \frac{1}{2} \left( \frac{\overline{Q}_{i+2}}{\overline{Q}_{i+1}} \right)^2 \right) \right]^{\frac{1}{2}}.
\] (23)
Fig. 7. Zoom of Schlieren image of shock–bubble problem at time 0.2; (a) QDS N2 scheme with 300 × 100 cells; (b) QDS 2N scheme with 300 × 100 cells; (c) QDS 2N scheme with 450 × 150 cells; (d) QDS 2N scheme with 600 × 200 cells.

b. Calculate the integral bounds $X_L$ and $X_R$:

\[
\begin{array}{l}
\text{If } V > 0, \quad \left\{ \begin{array}{l}
X_L = \Delta x - u \Delta t \\
X_R = \Delta x
\end{array} \right. \\
\text{otherwise } \left\{ \begin{array}{l}
X_L = 0 \\
X_R = u \Delta t
\end{array} \right.
\end{array}
\tag{24}
\]

(0, 0)

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<td>$\rho = 1.5$, $u_x = 0$</td>
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<tr>
<td>$p = 0.3$, $u_y = 0$</td>
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</tr>
<tr>
<td>$p = 0.029$, $u_y = 1.206$</td>
<td>$p = 0.3$</td>
</tr>
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Fig. 8. The initial conditions for the first problem of Euler-four-shock interaction.

c. Calculate the flux out values of average conserved properties $\overline{Q}_{ir}$ of particles to successfully move into the destination region.

d. Calculate the particle properties based on the average values $\overline{Q}_{ir}$.

e. Calculate the fluxes of conserved properties to neighboring cells following the standard QDS algorithm [5].

2.4. 2D flux calculation and implementation

Multi-dimensional extension is performed using the same principle applied for a one-dimensional reconstruction. The variation of conserved quantity $Q(x, y)$ over two dimensional space is assumed to be:

\[
Q(x, y) = Q_c + \left( \frac{dQ}{dx} \right)_{x=0.5\Delta x} (x - 0.5\Delta x) + \ldots
\]

\[
+ \left( \frac{d^2Q}{dx^2} \right)_{x=0.5\Delta x} \frac{(x - 0.5\Delta x)^2}{2} + \ldots
\]

\[
+ \left( \frac{d^{n-1}Q}{dx^{n-1}} \right)_{x=0.5\Delta x} \frac{(x - 0.5\Delta x)^{n-1}}{(n-1)!} + \ldots
\]

\[
+ \left( \frac{dQ}{dy} \right)_{y=0.5\Delta y} (y - 0.5\Delta y)
\]

\[
+ \left( \frac{d^2Q}{dy^2} \right)_{y=0.5\Delta y} \frac{(y - 0.5\Delta y)^2}{2} + \ldots
\]

\[
+ \left( \frac{d^{n-1}Q}{dy^{n-1}} \right)_{y=0.5\Delta y} \frac{(y - 0.5\Delta y)^{n-1}}{(n-1)!} .
\tag{25}
\]
The subsequent cell centered value of $Q_c$ is:

$$Q_c = Q_s - \frac{\Delta x^2}{24} \left( \frac{d^6 Q}{dx^2} \right) - \frac{\Delta x^4}{1920} \left( \frac{d^8 Q}{dx^4} \right) \cdots$$

$$- \frac{2 \Delta x^{n-1}}{n!} \left( \frac{\partial^{n-1} Q}{\partial x^{n-1}} \right) \cdots$$

$$+ \frac{\Delta y^2}{24} \left( \frac{d^6 Q}{dy^2} \right) - \frac{\Delta y^4}{1920} \left( \frac{d^8 Q}{dy^4} \right) \cdots$$

$$- \frac{2 \Delta y^{n-1}}{n!} \left( \frac{\partial^{n-1} Q}{\partial y^{n-1}} \right) \cdot (26)$$

Following this, the average value of conserved quantity in the region bounded by $[X_L, X_R] - [Y_B, Y_T]$ is formulated as:

$$\overline{Q}_{tr} = \frac{1}{(X_R - X_L)(Y_T - Y_B)} \int_{Y_B}^{Y_T} \int_{X_L}^{X_R} Q(x, y) \, dx \, dy$$

$$= Q_c + \frac{1}{(X_R - X_L)} \left\{ \left[ \left( \frac{dQ}{dx} \right) \frac{(X_L - 0.5 \Delta x)^2}{2} \cdots \right. \right.$$  

$$+ \left. \frac{\partial^{n-1} Q}{\partial x^{n-1}} \frac{(X_L - 0.5 \Delta x)^n}{n!} \right\}$$

$$- \left\{ \left( \frac{dQ}{dx} \right) \frac{(X_R - 0.5 \Delta x)^2}{2} \cdots \right. \right.$$  

$$+ \left. \frac{\partial^{n-1} Q}{\partial x^{n-1}} \frac{(X_R - 0.5 \Delta x)^n}{n!} \right\}$$

$$+ \frac{1}{(Y_T - Y_B)} \left\{ \left[ \left( \frac{dQ}{dy} \right) \frac{(Y_T - 0.5 \Delta y)^2}{2} \cdots \right. \right.$$  

$$+ \left. \frac{\partial^{n-1} Q}{\partial y^{n-1}} \frac{(Y_T - 0.5 \Delta y)^n}{n!} \right\}$$

$$- \left\{ \left( \frac{dQ}{dy} \right) \frac{(Y_B - 0.5 \Delta y)^2}{2} \cdots \right. \right.$$  

$$+ \left. \frac{\partial^{n-1} Q}{\partial y^{n-1}} \frac{(Y_B - 0.5 \Delta y)^n}{n!} \right\} \right\} \cdot (27)$$

where $Y_T$ and $Y_B$ are the bounds of integration in $y$ direction.

Since the average requires bounding regions in both translational directions, application of splitting (as applied to TDEFM to improve computational efficiency) is impossible, and the full QDS-$N^2$ number of particles (i.e. 9 when 3 particles are used per
direction, 16 for 4, etc.) are required for a complete flux computation. Previous extensions required only QDS-2N particles. Unlike the one-dimensional reconstruction, each particle carries three separate fluxes (for three different destination cells) and so any single QDS particle possesses three “sub-particles” based on different integral bounds. This concept is demonstrated in Fig. 4, showing each unique sub-region (A–C). The area of the sub-region A is $u \cdot v \cdot dt^2$ as described earlier in Section 2.1 for the QDS-2N method.

3. Results and discussion

In this section, we present four test cases for comparing the QDS-N$^2$ and the QDS-2N schemes that are hereafter referred to as N$^2$ and 2N, respectively. These four test cases include shock–bubble interaction, Euler-four-shock interaction, Euler-four-contact interaction and advection of vortical disturbances. The major discussions focus on the time cost and accuracy between the two schemes in two-dimensional problems.

3.1. Shock–bubble interaction

The strength of correct multi-dimension reconstruction is demonstrated in a two-dimensional shock/bubble interaction problem [8]. The initial conditions for this problem are shown in Fig. 5. The simulation calculates a shock wave, moving from left to right with a velocity of Mach number 2.85 in an ideal, inviscid gas and interacting with a bubble at $x = 0.3$. The results are presented with the snapshot at $t = 0.2$. The results of the numerical schlieren (gradients of density) are presented in Fig. 6 for various QDS schemes and the TVD scheme on a grid of $1700 \times 500$ cells. The application of correct multi-dimensional reconstructions results in a relatively high resolution of the circulation and reflected shock located at $x = 0.6$.

Fig. 7 displays two QDS schemes with different numbers of cells. We compare the N$^2$ scheme with $300 \times 100$ cells (Fig. 7a) against the 2N scheme (Fig. 7b–d) by using $300 \times 100$, $450 \times 150$, and $600 \times 200$ cells. For the sake of comparison, the limiter for each simulation is the monotonized central (MC) limiter. In Fig. 7a and b, the difference in resolution is clear despite the fact that both these
schemes employ the same number of cells. As the number of cells employed by the 2N scheme increases (shown up to 600 × 200 cells here), the results gradually approach those of the N² scheme with relatively few cells (1/4). Obviously, the multi-dimensional computation (N² scheme) achieves higher accuracy than the 2N scheme.

Further, we consider the computation time required by each scheme in this case. The N² scheme is true-directional, in that each possible combination of discrete velocity must be considered (nine instances with three discrete velocities per direction), while the 2N method employs approximate dimensional extension and only requires six discrete velocity computations in a two-dimensional simulation. Moreover, for each particle, three space-averaging computations are required for each fraction falling into separate destination cells. Therefore, the N² scheme requires more computation time for the same number of cells. The computation time of the two solvers are summarized in Table 1. According to these data, the N² extension of QDS requires approximately three times the computation time as compared to that of the original 2N scheme for the same number of cells. However, for any given degree of accuracy, we find that the N² scheme provides an increase in computational efficiency of almost three times (300 × 100 vs. 600 × 200 for 2N vs. N²). Thus, the application of the N² scheme is justified over that of the 2N scheme for high-resolution solutions.

### 3.2. Euler-four-shock interaction

This test case was introduced in Salichs [9], which computed the numerical solution employing the piecewise hyperbolic method-Marquina’s flux formula (PHM-MFF) and power PHM-MFF schemes. The test problem is initially divided into four quadrants sharing a common corner at 0.75 and 0.75 in the domain [0, 1] × [0, 1], as illustrated in Fig. 8. These quadrants initially have the following different but uniform conditions:

\[
(\rho, u, v, p) = \begin{cases} 
(1.5, 0, 0, 1.5), & [0.75, 1] \times [0.75, 1] (A) \\
(0.5323, 1.206, 0, 0.3), & [0, 0.75] \times [0, 0.75] (B) \\
(0.138, 1.206, 1.206, 0.029), & [0, 0.75] \times [0.75, 1] (C) \\
(0.5323, 0, 1.206, 0.3), & [0.75, 1] \times [0, 0.75] (D).
\end{cases}
\]

Fig. 9 shows four results of a comparison between the 2N and N² solvers at the time of 0.4. The Courant–Friedrichs–Lewy (CFL) factor is set as 0.5. We compare the results of the two QDS solvers using 100 × 100 to 300 × 300 cells. As can be seen, the result of the
Fig. 14. Density contour obtained from QDS-2N solver with 5 particles (a) and 9 particles in each direction (b); QDS-\( N^2 \) solver with 5 particles (c) and 9 particles in each direction (d) by using 1000 \( \times \) 1000 cells, 2nd order method with MINMOD limiter at time of 0.8. The CFL number is 0.5. Level from 0 to 2.1 at 0.05 interval of line.

Table 1
Comparison of computational expenses for QDS schemes using 2N and \( N^2 \) dimensional reconstruction.

<table>
<thead>
<tr>
<th>Number of cells</th>
<th>QDS scheme</th>
<th>2N (min)</th>
<th>( N^2 ) (min)</th>
</tr>
</thead>
<tbody>
<tr>
<td>300 ( \times ) 100</td>
<td>8.41</td>
<td>23.15</td>
<td></td>
</tr>
<tr>
<td>450 ( \times ) 150</td>
<td>28.9</td>
<td>78.3</td>
<td></td>
</tr>
<tr>
<td>600 ( \times ) 200</td>
<td>68.56</td>
<td>183.6</td>
<td></td>
</tr>
<tr>
<td>1000 ( \times ) 500</td>
<td>478.4</td>
<td>1282.6</td>
<td></td>
</tr>
</tbody>
</table>

Table 2
QDS scheme time cost in Euler-4-shocks interaction case.

<table>
<thead>
<tr>
<th>Number of cells</th>
<th>QDS solvers</th>
<th>2N</th>
<th>( N^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1000 ( \times ) 1000</td>
<td>13.29 h</td>
<td>55.6 h</td>
<td></td>
</tr>
<tr>
<td>100 ( \times ) 100</td>
<td>45 (s)</td>
<td>189 (s)</td>
<td></td>
</tr>
<tr>
<td>200 ( \times ) 200</td>
<td>375 (s)</td>
<td>1520 (s)</td>
<td></td>
</tr>
<tr>
<td>300 ( \times ) 300</td>
<td>1307 (s)</td>
<td>5199 (s)</td>
<td></td>
</tr>
</tbody>
</table>

\( N^2 \) solver obtained using a coarse grid (100 \( \times \) 100 cells) is only approached by the 2N solver when employing considerably fine grids (300 \( \times \) 300 cells). Furthermore, the result that we obtained using the \( N^2 \) solver on a computational grid of 1000 \( \times \) 1000 cells is similar to that obtained using the total variation diminishing–monotone upstream centered schemes for conservation laws (TVD–MUSCL) scheme [8] (see Fig. 10a).

An investigation of the computational expense of each scheme showed that the \( N^2 \) solver takes approximately four times longer to complete the simulation than the 2N solver. This comparison of computational expense is summarized in Table 2. The increase in computational time with the refinement of the computational grid is due to the constant “kinetic” CFL condition that we employ, which is defined as follows:

\[
CFL = \text{Max} \left( \left( \frac{\sqrt{u^2 + 3} \sqrt{RT}}{dx} \right) \frac{dt}{dx} \right). 
\]

This basically ensures that particles in free flight do not travel further than the adjacent cells. Although the result takes more time to compute using the \( N^2 \) solver, the accuracy is considerably better than that of the 2N solver; in fact, it is not clear that the 2N...
Fig. 12 shows the numerical result of the second-order TVD–MUSCL method for a density contour profile on a 1000 × 1000 uniform grid, taken from [8]. For the QDS scheme, the result obtained at the time of 0.8 on a 1000 × 1000 uniform grid can be seen in Fig. 13. Two results are shown for the second-order method with the N⁰ and the 2N solvers. Both enforce a constant CFL number of 0.25. The contours of density are presented with levels of 0–2.4. In this case, a shock wave is generated and spirals from the contact point in an unsteady fashion. By comparing the two figures, we find that both the N⁰ and the 2N solver results are symmetrical and that the result obtained using the N⁰ solver is closer to the TVD–MUSCL result presented in Fig. 12. As in the previous test cases, in the current test case, the accuracy of the N⁰ method is superior to that of the 2N method. In this instance, however, the essentially non-oscillatory weighted (WENO) results [11] are still superior to the N⁰ results; this can be attributed to the small stencil employed for the estimation of the higher-order gradients or to the flux splitting employed and the inevitably associated numerical dissipation; this deserves further investigation.

Further, we have compared the timings and the accuracy for this test problem with different N for both QDS-2N and QDS-N⁰ with 1000 × 1000 cells since both schemes scale differently with N. The results are essentially the same as those obtained for N = 3 when N increases to 5 or 9 for both the above mentioned methods, as shown in Fig. 14. This is reasonable since the integration of a Gaussian function with a polynomial, having two or fewer degrees, becomes exact, if the number of Gaussian–Hermite integration points is 3 or more. Expectedly, the computation time increases roughly 3 times from N = 3 to N = 9 for both the methods. Further, Fig. 15 shows the density contours when the grid resolution increases from 1000 × 1000 cells to 2000 × 2000 and 3000 × 3000 cells. In brief summary, for both the QDS-2N and the QDS-N⁰ methods for solving the Euler equation, accuracy effectively increases with increasing grid resolution, while it is essentially the same when N ≥ 3.

3.4. Advection of vortical disturbance

The final test case consists of an inviscid unsteady flow in which a vortex is located at the center of a uniform domain (x₀, y₀). The mean flow for this case uses Mach number M∞ = 0.1. This case tests the capabilities of the QDS scheme as compared to the exact solution taken from Visbal and Gaitonde [12] in order to accurately advect vortical disturbances. This problem also appears in Tutkun and Edis [13]. The initial conditions are as follows:

\begin{align}
    u &= U∞ - \frac{C(y - y_1)}{Rc} \exp \frac{-r^2}{2}, \\
    v &= \frac{C(x - x_1)}{Rc} \exp \frac{-r^2}{2}
\end{align}

(31)

where u, v, and Rc determine the Cartesian velocity components and the vortex core radius, respectively. C is the vortex strength parameter, defined as follows:

\begin{align}
    \frac{C}{(U∞ Rc)} = 0.02.
\end{align}

(32)

The density is assumed to be constant, and the vortex radius Rc is taken to be 1.0 in this case.

Fig. 16 shows the vorticity contours of the N⁰² and the 2N solvers with 800 × 800 cells using the second-order method. The limiter in this case is the monotonized central (MC) method. A constant CFL number (0.1) is enforced such that the non-dimensional time step size is dt/∞/Rc = 4.0 × 10⁻³. The result of the N⁰² solver is essentially the same as the exact solution and shows a perfect circular shape of the vorticity distribution while that of the 2N solver does
not. The result of the 2N solver shows more significant dissipation and anisotropy errors as compared to that of the N^2 solver. Fig. 17 shows the vorticity distributions of various simulations along a horizontal line (at \( y = 8.0 \)) passing through the vortex center in Fig. 16. We have compared the results obtained by using the two solvers (2N and N^2) on a uniform grid containing three different levels of resolution (160 \( \times \) 160, 800 \( \times \) 800, and 1600 \( \times \) 1600 cells). The result obtained using the N^2 solver in the case of 800 \( \times \) 800 cells is in excellent agreement with the exact solution and radial symmetry, while the results obtained using the 2N solver are far from the correct solution even in the case of 1600 \( \times \) 1600 cells. Thus, the influence of multi-dimensional reconstruction is significant for the QDS, particularly on the numerical accuracy of the solution for a gas flow field as in the current problem.

The investigation of the computational expense again reveals a trade-off between computational time and accuracy. The computational time of the N^2 solver in terms of calculation time is approximately 3~4 times less than that of the 2N solver for the same computational grid, although the accuracy of the former is considerably better than that of the latter. This leads to a question of whether the use of the N^2 method is worthwhile or not. Thus, we compare the results obtained using the N^2 method using 160 \( \times \) 160 cells with those obtained using the 2N method using 1600 \( \times \) 1600 cells, as shown in Fig. 17. The results show that they are essentially the same for the same level of accuracy; thus, the proposed N^2 solver is approximately 25 times faster than the 2N solver in this case. Once again, we are unsure whether the 2N result will ever converge to the analytical solution, thereby justifying the application of the N^2 solver and its proposed multi-dimensional reconstruction of QDS particles.

4. Conclusion

In this paper, a true-direction multi-dimensional higher-order extension of the QDS method, referred to as the N^2 solver, was introduced and verified using various test cases. The results showed that the N^2 solver was considerably more accurate than the 2N solver in general. It appeared that the N^2 solver (i) improved the solution in the flows unaligned with the computational grid and (ii) significantly reduced the amount of numerical dissipation within the solution. Despite the additional computational expense associated with the N^2 solver for the same computational grid, for any given degree of accuracy, the proposed solver was found to be several times (up to 25 times in the case of the advection of vortical disturbances) faster than the original 2N method. Of particular interest was the test case of the advection of vortical disturbances, where the N^2 method improved the radial symmetry of the result approaching the analytical solution, while the 2N method failed to converge to the analytical solution even when a very fine grid
Fig. 17. The vorticity profiles along the central line passing through the vortex. The comparison contained the exact solution (blue squiggle-symbolline), the QDS $N_2$ solver using $160 \times 160$ cells (red line), $800 \times 800$ cells (black dash-dot line), and $2N$ solver using $800 \times 800$ cells (purple long-dash line), $1600 \times 1600$ cells (green dotted line). Two solvers are computed in the MC limiter and CFL $= 0.1$ at time 8.0. (For interpretation of the references to colour in this figure legend, the reader is referred to the web version of this article.)

was used. In the cases of both the QDS-2N and QDS-N$^2$ methods for solving the Euler equation, accuracy effectively increases with increasing grid resolution, while it was essentially the same when $N \geq 3$ because the integration of the Gauss function with a polynomial (degree $\leq 2$) using the Gauss–Hermite integration technique became exact.

A significant improvement on the errors associated with the flow misalignment with the computational grid was demonstrated through numerical experiments with the proposed multidimensional reconstruction employed by the $N^2$ method. The method demonstrated a promising improvement over the $2N$ extension for future applications where resolution was critical. However, an analytical analysis of the difference between the two schemes is necessary to further reveal the details; such an analysis is currently in progress and will be reported elsewhere in the near future. Moreover, further research is focused on the extension of the method to efficient parallel computation for large-scale problems using both conventional (MPI) parallelization and acceleration using graphics processing units (GPUs) by taking advantage of the highly local nature of the QDS scheme.

Acknowledgment

The authors are thankful for financial support from National Science Council of Taiwan through the Grants NSC-99-2922-I-009-107, and NSC100-2627-E-009-001, NSC101-2627-E-009-001.

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