Article
High-Effectiveness and -Accuracy Difference Scheme Based on Nonuniform Grids for Solving Convection–Diffusion Equations with Boundary Layers

Fang Tian 1,*, Mingjing Wang 1 and Yongbin Ge 2

1 School of Mathematics and Statistics, Ningsxia University, Yinchuan 750021, China; 12021130531@stu.nxu.edu.cn
2 School of Science, Dalian Minzu University, Dalian 116600, China; gybnxu@yeah.net
* Correspondence: tianf20191226@nxu.edu.cn

Abstract: In this paper, some rational high-accuracy compact finite difference schemes on nonuniform grids (NRHOC) are introduced for solving convection–diffusion equations. The derived NRHOC schemes not only can suppress the oscillatory property of numerical solutions but can also obtain a high-accuracy approximate solution, and they can effectively solve the convection–diffusion problem with boundary layers by flexibly adjusting the discrete grid, which can be obtained with the singularity in the computational region. Three numerical experiments with boundary layers are conducted to verify the accuracy of the proposed NRHOC schemes. We compare the computed results with the analytical solutions, the results of the rational high-accuracy compact finite difference schemes on uniform grids (RHOC), and the other schemes in the literature. For all test problems, good computed results are obtained with the presented NRHOC schemes. It is shown that the presented NRHOC schemes have a better resolution for the solution of convection-dominated problems.

Keywords: nonuniform grid; compact finite difference scheme; convection–diffusion equation; boundary layer

MSC: 65N06; 65N12; 65N22

1. Introduction
The Navier–Stokes equations are very difficult to solve due to their highly nonlinear nature, and in particular, approximate solutions are required to have high accuracy. The convection–diffusion equation is a linearized model of the Navier–Stokes equations, and the study of its high-precision values can help develop algorithms for solving Navier–Stokes equations with high accuracy. At the same time, the convection–diffusion equation itself has a wide range of applications in certain fields of physics, chemistry, biology, environment, finance, biology, etc. [1–3]. Therefore, the study of high-precision numerical solutions to convection–diffusion equations is very important and has some academic research value. Researchers have developed various higher-accuracy, compact finite difference methods to solve kinds of convection–diffusion problems [4–24].

For the 1D (one-dimensional) convection–diffusion equation

$$-\varepsilon u_{xx} + c(x)u_x = f(x) \quad (1)$$

the fourth-order high-accuracy compact finite difference scheme (referred to as FOC scheme) obtained by Spotz [7] using the original equation and truncation error correction method is

$$-\varepsilon \left( 1 + \frac{c^2 h^2}{12 \varepsilon} \right) \Delta^2_x u_i + c \Delta_x u_i = \left( 1 - \frac{ch^2}{12\varepsilon} \frac{h^2}{12} \right) f_i \quad (2)$$
This scheme gives theoretical fourth-order accuracy when solving problems with smooth solutions in the computational region, but when the computational region contains boundary layers or singularities, the computational error of this scheme on coarse grids is large and does not achieve theoretical fourth-order accuracy. If the boundary-layer problem is to be solved with high accuracy, the mesh must be divided finely enough. Too fine grids result in huge computational cost, are memory-consuming, and greatly reduce computational efficiency. From the derivation process, it is not difficult to find that the FOC scheme is a strongly differential scheme. As the mesh Reynolds number, $Pe = \frac{ch}{\varepsilon}$, increases, the dissipation error increases rapidly, thus greatly reducing the resolution of the difference scheme. As we know, in numerical simulation, the resolution of the difference scheme is a key factor affecting the numerical accuracy and reliability of the numerical results. To reduce dissipation errors and improve computational accuracy, we have modified the FOC scheme to obtain a rational fourth-order compact difference scheme on a uniform grid (see Appendix A).

\[
-a\delta^2_{x} u_i + c \delta_{x} u_i = f_i + c_1 f_{xi} + c_2 f_{xxi}
\]

where the coefficients are

\[
a = \frac{1 + \frac{c^2 h^2}{6\varepsilon}}{1 + \frac{c^2 h^2}{12\varepsilon}}, \quad C_1 = -\frac{h^2}{12} \frac{ch}{\varepsilon}, \quad C_2 = \frac{h^2}{12} \frac{1 + \frac{c^2 h^2}{6\varepsilon}}{1 + \frac{c^2 h^2}{12\varepsilon}}
\]

For both the FOC scheme and the modified rational scheme in (3), Fourier error analysis reveals that the dissipation error is significantly reduced for the modified scheme for larger grid Reynolds numbers, but the error is still larger at medium and high wavelengths (see Appendix B). By monitoring the calculation process of specific examples, it is easy to see that if the computational grid is not sufficiently subdivided, there are no computational nodes or not enough computational nodes are distributed near and inside the boundary layer, which results in large computational errors in places with large gradient changes. Therefore, a more feasible and reasonable approach to improving the calculation accuracy of boundary-layer attachments is to distribute more calculation nodes in the area with large gradients or boundary layers, while fewer calculation nodes are distributed in the area with a small gradient or relatively gentle changes in physical quantities. This not only can consider the stability of the algorithm and the accuracy of the calculation results but can also reduce the calculation amount, which requires nonuniform mesh generation in the calculation area and that the calculation be carried out on a nonuniform grid. The main methods currently available to solve this type of problem are local grid refinement methods [25,26], mesh adaptive algorithms [27], coordinate transformation techniques [28–30], transform-free, nonuniform, grid-based direct discretization methods [31–34], and other new methods [35–38].

A fourth-order compact difference scheme for solving the two-dimensional convection–diffusion equation was first proposed by Gupta [4] using the original equation and the truncation error correction method. However, the difference scheme is a strongly dissipative scheme that performs poorly in solving large-gradient or boundary-layer problems. Kalita et al. [31] extended this fourth-order compact difference scheme to a nonuniform grid and developed a polynomial compact difference scheme on nonuniform grids, which improved the solution accuracy of some boundary-layer problems but did not completely improve the strong dissipation of the difference scheme. In this paper, we obtain fourth-order rational compact difference schemes based on nonuniform grids for 1D and 2D convection–diffusion equations using the ideas of the construction of modified differential Equation (3).

The remainder of this study is arranged into four sections as follows.

Section 2 develops NRHOC difference schemes for 1D convection–diffusion equations.
Section 3 constructs NRHOC difference schemes for 2D convection–diffusion equations.
Section 4 carries out numerical experiments to verify the feasibility of the presented NRHOC finite difference schemes. Section 5 concludes the work of the paper.

2. RHOC Scheme on Nonuniform Grids (NRHOC Scheme) for 1D Convection–Diffusion Equations

Let us consider the following 1D convection–diffusion equation:

\[
\begin{aligned}
-\varepsilon u_{xx} + c(x)u_x &= f, & a < x < b \\
u(a) &= \phi(x), \quad u(b) = \psi(x)
\end{aligned}
\] (5)

where \(\varepsilon\) is constant, with \(\varepsilon > 0\), and is the diffusion coefficient; \(c(x)\) is the convection coefficient; \(f(x)\) is called the source term; \(u(x)\) is the unknown quantity to be computed. Suppose that \(c(x), f(x),\) and \(u(x)\) are all sufficiently smooth functions of \(x\).

2.1. NRHOC Scheme for 1D Constant-Coefficient Convection–Diffusion Equations

If \(c(x)\) is a constant, Equation (5) is

\[-\varepsilon u_{xx} + cu_x = f(x)\]

(6)

First, the solution interval \([a, b]\) is subdivided arbitrarily into \(N\) sub-intervals, and the inserted computational nodes are \(a = x_0 < x_1 < x_2 < \cdots < x_N = b\) in order. We define the step lengths of the left and right sides of node \(x_i\) as \(h_L = x_i - x_{i-1}\) and \(h_R = x_{i+1} - x_i\), \(i = 0, 1, 2, \cdots, N\), respectively.

Assuming that solution \(u\) is smooth enough in the computational region, the approximate formula for calculating the first- and second-order derivatives at node \(x_i\) is derived using Taylor’s formula as

\[
\left(\frac{\partial u}{\partial x}\right)_i = \delta x u_i - \chi_1 \left(\frac{\partial^3 u}{\partial x^3}\right)_i - \chi_2 \left(\frac{\partial^4 u}{\partial x^4}\right)_i + O(\eta)
\]

(7)

\[
\left(\frac{\partial^2 u}{\partial x^2}\right)_i = \delta^2 x u_i - \chi_3 \left(\frac{\partial^3 u}{\partial x^3}\right)_i - \chi_4 \left(\frac{\partial^4 u}{\partial x^4}\right)_i + O(\omega)
\]

(8)

where

\[
\chi_1 = \frac{h_L h_R}{6}, \chi_2 = \frac{h_L h_R (h_R - h_L)}{24}, \chi_3 = \frac{h_R - h_L}{3}, \chi_4 = \frac{h_R^3 - h_L^3 - h_R h_L^2 + h_L^3}{12}
\]

(9)

\[
\eta = h_L h_R (h_R^2 - h_L h_R + h_L^2), \omega = \frac{(h_R - h_L)(h_R^3 + h_L^3)}{60}
\]

(10)

\[
\delta x u_i = \frac{h_R^2 u_{i+1} + (h_R^2 - h_L^2) u_i - h_L^2 u_{i-1}}{h_L h_R (h_L + h_R)}
\]

(11)

\[
\delta^2 x u_i = \frac{2[h_L u_{i+1} - (h_R + h_L) u_i + h_R u_{i-1}]}{h_L h_R (h_L + h_R)}
\]

(12)

where \(\delta x u_i\) and \(\delta^2 x u_i\) are the second-order central difference operators on nonuniform grids, respectively.
By substituting (7) and (8) in (6), we yield the following equation:

$$-\epsilon \delta^2 u_i + c_i \delta_x u_i + (\epsilon \chi_3 - c \chi_1) \left( \frac{\partial^3 u_i}{\partial x^3} \right) + (\epsilon \chi_4 - c \chi_2) \left( \frac{\partial^3 u_i}{\partial x^3} \right) + o(\eta) + o(\omega) = f_i$$  \hspace{1cm} (13)

The derivative function is calculated directly from Equation (6), and using the original equation gives

$$u_{xxx} = \frac{c}{\epsilon} u_{xx} - \frac{1}{\epsilon} f_x$$

$$u_{xxxx} = \frac{c^3}{\epsilon^3} u_x - \frac{c^2}{\epsilon^2} f - \frac{c}{\epsilon} f_x = \frac{1}{\epsilon} f_{xx}$$

And by substituting this in (13), we obtain

$$-\epsilon \delta^2 u_i + c_i \delta_x u_i + \left( -\frac{\epsilon}{\tau} \chi_1 + c \chi_3 \right) u_{xxi} + \left( \frac{\epsilon}{\tau} \chi_4 - \frac{\epsilon^3}{\tau^3} \chi_2 \right) u_{xxi} + o(\eta) + o(\omega)$$

$$= \left( 1 + \frac{\epsilon}{\tau} \chi_2 + \frac{\epsilon^3}{\tau^3} \chi_4 \right) f_i + \left( \chi_3 - \frac{\epsilon}{\tau} \chi_1 + \frac{\epsilon}{\tau} \chi_4 - \frac{\epsilon^3}{\tau^3} \chi_2 \right) f_{xi} + (\chi_4 - \frac{\epsilon}{\tau} \chi_2) f_{xxi}$$ \hspace{1cm} (14)

By substituting (7) and (8) in (14), we obtain

$$-\epsilon \left( 1 + \frac{\epsilon}{\tau} \chi_1 - \frac{\epsilon}{\tau} \chi_3 \right) \delta^2 u_i + c_i \left( 1 - \frac{\epsilon}{\tau} \chi_2 + \frac{\epsilon^3}{\tau^3} \chi_4 \right) \delta_x u_i + \kappa_1 u_{xxxi} + o(\eta) + o(\omega)$$

$$= \left( 1 - \frac{\epsilon}{\tau} \chi_2 + \frac{\epsilon^3}{\tau^3} \chi_4 \right) f_i + \left( \chi_3 - \frac{\epsilon}{\tau} \chi_1 + \frac{\epsilon}{\tau} \chi_4 - \frac{\epsilon^3}{\tau^3} \chi_2 \right) f_{xi} + (\chi_4 - \frac{\epsilon}{\tau} \chi_2) f_{xxi}$$ \hspace{1cm} (15)

where

$$\eta_1 = \left( \frac{\epsilon^3}{\tau^3} \chi_4 - \frac{\epsilon^4}{\tau^4} \chi_2 \right) \eta, \quad \omega_1 = \left( -\frac{\epsilon}{\tau} \chi_1 + c \chi_3 \right) \omega$$

$$\kappa_1 = -\chi_3 \left( -\frac{\epsilon}{\tau} \chi_1 + c \chi_3 \right) - \chi_1 \left( \frac{\epsilon^3}{\tau^3} \chi_4 - \frac{\epsilon^4}{\tau^4} \chi_2 \right)$$ \hspace{1cm} (16)

By replacing $u_{xxx}$ in (15) with $u_{xxx} = \frac{\epsilon}{\tau} u_{xx} - \frac{1}{\epsilon} f_x$ and omitting the high-order terms, we obtain

$$-\epsilon \left( 1 + \frac{\epsilon}{\tau} \chi_1 - \frac{\epsilon}{\tau} \chi_3 \right) \delta^2 u_i + c_i \left( 1 + \frac{\epsilon}{\tau} \chi_2 + \frac{\epsilon^3}{\tau^3} \chi_4 \right) \delta_x u_i + (\kappa_1 + \kappa_2) u_{xxxi} + o(\eta_1 + \omega_1)$$

$$= \left( 1 - \frac{\epsilon}{\tau} \chi_2 + \frac{\epsilon^3}{\tau^3} \chi_4 \right) f_i + \left( \chi_3 - \frac{\epsilon}{\tau} \chi_1 + \frac{\epsilon}{\tau} \chi_4 - \frac{\epsilon^3}{\tau^3} \chi_2 \right) f_{xi} + (\chi_4 - \frac{\epsilon}{\tau} \chi_2) f_{xxi}$$

Then, the NRHOC difference scheme of constant-coefficient convection–diffusion Equation (6) is yielded.

$$-a \delta^2 u_i + c_i \delta_x u_i = f_i + c_i f_{xi} + c_2 f_{xxi}$$ \hspace{1cm} (17)

where

$$a = \epsilon \left( 1 + \frac{\epsilon}{\tau} \chi_1 - \frac{\epsilon}{\tau} \chi_3 \right)$$

$$c_1 = \epsilon \left( \frac{\epsilon}{\tau} \chi_1 - \frac{\epsilon}{\tau} \chi_2 + \chi_3 + \frac{\epsilon}{\tau} \chi_4 \right)$$

$$c_2 = \frac{\chi_4 - \frac{\epsilon}{\tau} \chi_2 - \frac{\epsilon}{\tau} \chi_3}{1 - \frac{\epsilon}{\tau} \chi_2 + \frac{\epsilon^3}{\tau^3} \chi_4}$$

$$\kappa = -\chi_3 \left( -\frac{\epsilon}{\tau} \chi_1 + \frac{\epsilon^3}{\tau^3} \chi_4 \right) - \chi_1 \left( \frac{\epsilon^3}{\tau^3} \chi_4 - \frac{\epsilon^4}{\tau^4} \chi_2 \right)$$

$$\chi_1 = \frac{h_1 h_R}{6}, \chi_2 = \frac{h_1 h_R (h_R - h_1)}{24}, \chi_3 = \frac{h_R - h_1}{3}, \chi_4 = \frac{h^2_R - h_L h_R + h_1^2}{12}$$
2.2. NRHOC Scheme for 1D Variable-Coefficient Convection–Diffusion Equations

In this subsection, the RHOC scheme for model Equation (5) will be derived based on difference scheme (17). Suppose that model Equation (5) has the following difference scheme:

\[-\lambda_i \delta_x^2 u_i + c_i \delta_x u_i = f_i + c_{i1} f_{xi} + c_{2i} f_{xxi}\]  \hfill (19)

Using Taylor’s formula and the original equation (Equation (5)), the modified differential equation of difference scheme (19) can be derived as

\[-e u_{xx} + c u_x = f + 2c_2 c_x u_{xx} + (c_1 c_x + c_2 c_{xx}) u_x + O(\eta + \omega)\]  \hfill (20)

where \(\eta = h_L h_R (h_R^2 - h_L h_R + h_L^2), \omega = \frac{(h_R - h_L) (h_R^2 + h_L^2)}{60}\).

To obtain the higher-order difference scheme of the original equation (Equation (5)), its modified differential equation is obtained using residual correction:

\[-B u_{xx} + D(x) u_x = f(x)\]  \hfill (21)

where

\[B = \epsilon - 2c_2 c_x, \quad D(x) = c + c_1 c_x + c_2 c_{xx}\]  \hfill (22)

By applying (17) to (21), we obtain the NRHOC scheme for one-dimensional variable-coefficient convection–diffusion Equation (5) as follows:

\[-\Lambda_i \delta_x^2 u_i + D_i \delta_x u_i = f_i + C_{i1} f_{xi} + C_{2i} f_{xxi}\]  \hfill (23)

where the coefficients are

\[\Lambda_i = B_i \left(1 + \frac{D_i^2}{B_i^4} X_1 - \frac{D_i^3}{B_i^6} X_3 \right) \frac{1}{1 - \frac{D_i^3}{B_i^6} X_2 + \frac{D_i^2}{B_i^4} X_4}\]

\[D_{i1} = -\frac{D_i}{B_i^2} X_1 - \frac{D_i^2}{B_i^4} X_2 + X_3 + \frac{D_i^3}{B_i^6} X_4 \frac{1 - \frac{D_i^3}{B_i^6} X_2 + \frac{D_i^2}{B_i^4} X_4}{1 - \frac{D_i^3}{B_i^6} X_2 + \frac{D_i^2}{B_i^4} X_4}, \quad D_{2i} = \frac{X_4 - \frac{D_i^3}{B_i^6} X_2 - \frac{K_{1i}}{B_i^2}}{1 - \frac{D_i^3}{B_i^6} X_2 + \frac{D_i^2}{B_i^4} X_4},\]

\[K_{1i} = -\chi_1 \left(-\frac{D_i^2}{B_i^4} X_1 + c_2 X_3\right) - \chi_1 \left(\frac{D_i^3}{B_i^6} X_4 - \frac{D_i^4}{B_i^8} X_2\right),\]

\[B_i = \epsilon - 2c_2 c_{xiL}, \quad D_i = c_i + c_{i1} c_{xi1} + c_{2i} c_{xxi1},\]

\[c_{i1} = \frac{-\epsilon_i}{2} X_1 - \frac{c_i}{\epsilon} X_2 + X_3 + \frac{c_i^2}{\epsilon} X_4, \quad c_{2i} = \frac{X_4 - \frac{c_i}{\epsilon} X_2 - \frac{K_{1i}}{c_{x1}}}{1 + \frac{c_i}{\epsilon} X_2 + \frac{c_i^2}{\epsilon^2} X_4},\]

\[\chi_1 = \frac{h_L h_R}{6} + X_2 = \frac{h_L h_R (h_R - h_L)}{24}, \quad \chi_3 = \frac{h_R - h_L}{3}, \quad \chi_4 = \frac{h_R^2 - h_L h_R + h_L^2}{12}.\]

Difference Equation (23) is the rational compact difference scheme on nonuniform grids of model Equation (5). In particular, if \(h_L = h_R\), the difference equation is degenerated into difference Equation (A13) on the grid.
3. NRHOC Scheme for 2D Convection–Diffusion Equations

Let us consider the following 2D convection–diffusion equation:

\[
\begin{cases}
-\epsilon_1 u_{xx} - \epsilon_2 u_{yy} + c(x, y)u_x + d(x, y)u_y = f(x, y), & a < x < b, c < y < d \\
u(a, y) = \phi_1(y), u(b, y) = \phi_2(y) \\
u(x, c) = \phi_1(x), u(x, d) = \phi_2(x)
\end{cases}
\]  

(25)

where \(\epsilon_1, \epsilon_2\) are constant and \(\epsilon_1 > 0, \epsilon_2 > 0, c, d,\) and \(f\) are smooth enough functions of variables \(x\) and \(y\). It is a linearized model equation of a two-dimensional, steady, incompressible Navier–Stokes equation.

First, the constant-coefficient problem is considered; then, the NRHOC scheme of Equation (25) with variable coefficients can be obtained by modifying the difference equation.

3.1. NRHOC Scheme for 2D Constant-Coefficient Convection–Diffusion Equations

Let us consider the following two-dimensional convection–diffusion equation with constant coefficients:

\[-\epsilon_1 u_{xx} - \epsilon_2 u_{yy} + cu_x + du_y = f(x, y)\]  

(26)

We rewrite Equation (26) into the following equivalent equations:

\[
\begin{cases}
-\epsilon_1 u_{xx} + cu_x = f_1(x, y) \\
-\epsilon_2 u_{yy} + du_y = f_2(x, y)
\end{cases}
\]

(27)

where

\[
\begin{cases}
f_1(x, y) = f(x, y) + \epsilon_2 u_{yy} - du_y \\
f_2(x, y) = f(x, y) + \epsilon_2 u_{xx} - cu_x
\end{cases}
\]

(28)

Applying difference scheme (17) to (27) and (28) yields

\[
\begin{cases}
-\alpha_1 \delta_x^2 u_{ij} + c \delta_x u_{ij} = F_{1ij} \\
-\alpha_2 \delta_y^2 u_{ij} + d \delta_y u_{ij} = F_{2ij}
\end{cases}
\]

(29)

where

\[
\begin{cases}
F_{1ij} = f_{1ij} + c_1 f_{1xij} + c_2 f_{1xxij} \\
F_{2ij} = f_{2ij} + d_1 f_{2yij} + d_2 f_{2yyij}
\end{cases}
\]

(30)

We use Equation (28) to calculate the derivative and substitute it into (29); then, the two obtained equations are added together to obtain the NRHOC scheme of two-dimensional constant-coefficient model Equation (26) as follows:

\[
\begin{cases}
-\alpha_x \delta_x^2 - \alpha_y \delta_y^2 + c \delta_x + d \delta_y + E \delta_x^2 \delta_y + G \delta_y^2 \delta_x + H \delta_x \delta_y + R \delta_x \delta_y \} u_{ij} = F_{ij}
\end{cases}
\]

(31)

where

\[
\alpha_x = \epsilon_1 \frac{1 + \frac{\epsilon_2^2}{\epsilon_1} x_1 - \frac{\epsilon_2}{\epsilon_1} x_3}{1 - \frac{\epsilon_2^2}{\epsilon_1} x_2 + \frac{\epsilon_2}{\epsilon_1} x_4}, \alpha_y = \epsilon_2 \frac{1 + \frac{\epsilon_1^2}{\epsilon_2} x_1 - \frac{\epsilon_1}{\epsilon_2} x_3}{1 - \frac{\epsilon_1^2}{\epsilon_2} x_2 + \frac{\epsilon_1}{\epsilon_2} x_4}
\]

\[
c_1 = \frac{-\frac{\epsilon_2}{\epsilon_1} x_1 - \frac{\epsilon_2^2}{\epsilon_1} x_2 + x_3 + \frac{\epsilon_1^2}{\epsilon_2} x_4}{1 - \frac{\epsilon_2^2}{\epsilon_1} x_2 + \frac{\epsilon_2}{\epsilon_1} x_4}, c_2 = \frac{x_4 - \frac{\epsilon_2}{\epsilon_1} x_2 - \frac{\epsilon_1}{\epsilon_2} x_3}{1 - \frac{\epsilon_1^2}{\epsilon_2} x_2 + \frac{\epsilon_1}{\epsilon_2} x_4}
\]

\[
d_1 = \frac{-\frac{\epsilon_1}{\epsilon_2} x_1 - \frac{\epsilon_1^2}{\epsilon_2} x_2 + x_3 + \frac{\epsilon_2^2}{\epsilon_1} x_4}{1 - \frac{\epsilon_1^2}{\epsilon_2} x_2 + \frac{\epsilon_1}{\epsilon_2} x_4}, d_2 = \frac{x_4 - \frac{\epsilon_1}{\epsilon_2} x_2 - \frac{\epsilon_2}{\epsilon_1} x_3}{1 - \frac{\epsilon_1^2}{\epsilon_2} x_2 + \frac{\epsilon_1}{\epsilon_2} x_4}
\]
\[ E = - c_2 \varepsilon_2 - d_2 \varepsilon_1, G = c_2 d - d_1 \varepsilon_1, H = d_2 c - c_1 \varepsilon_2, R = c_1 d + d_1 c \]

\[ \kappa_1 = - \chi_3 \left( - \frac{c_2^2}{\varepsilon_1} \chi_1 + c_\chi_3 \right) - \chi_1 \left( \frac{c_3^3}{\varepsilon_1^3} \chi_4 - \frac{c_4}{\varepsilon_1} \chi_2 \right) \]

\[ \kappa_2 = - \chi_3 \left( - \frac{d_2^2}{\varepsilon_2} \chi_1 + d_\chi_3 \right) - \chi_1 \left( \frac{d_3^3}{\varepsilon_2^3} \chi_4 - \frac{d_4}{\varepsilon_2} \chi_2 \right) \]

\[ F_{ij} = \left( 1 + c_1 \delta_x + c_2 \delta_y^2 + d_1 \delta_y + d_2 \delta_y^2 \right) f_{ij} \]

### 3.2. NRHO Scheme for 2D Variable-Coefficient Convection-Diffusion Equations

Assume that model Equation (25) has a difference scheme at point \((x_j, y_j)\) like (31):

\[ \left\{ - \alpha_{xij} \delta_x^2 - \alpha_{yij} \delta_y^2 + c \delta x + d \delta y + E \delta_x^2 \delta_y + G \delta_y^2 \delta_y + H \delta_3 \delta_x + R \delta_3 \delta_y \right\} u_{ij} = F_{ij} \quad (33) \]

where the coefficients are

\[ \alpha_{xij} = \varepsilon_1 + \frac{c_{ij}^2}{\varepsilon_1^3} \chi_1 - \frac{c_{ij}^2}{\varepsilon_1^2} \chi_3 + \frac{c_{ij}^2}{\varepsilon_1^2} \chi_4, \alpha_{yij} = \varepsilon_2 + \frac{d_{ij}^2}{\varepsilon_2^2} \chi_1 - \frac{d_{ij}^2}{\varepsilon_2} \chi_3 + \frac{d_{ij}^2}{\varepsilon_2} \chi_4 \]

\[ c_{1ij} = - \frac{c_{ij}^2}{\varepsilon_1^2} \chi_1 - \frac{d_{ij}^2}{\varepsilon_2^2} \chi_2 + \frac{c_{ij}^2}{\varepsilon_1^2} \chi_4, c_{2ij} = \frac{\chi_4 - \frac{c_{ij}^3}{\varepsilon_1^3} \chi_2 - \frac{c_{ij}^2}{\varepsilon_1} \chi_3}{1 - \frac{c_{ij}^2}{\varepsilon_1^2} \chi_2 + \frac{c_{ij}^2}{\varepsilon_1^2} \chi_4} \]

\[ d_{1ij} = - \frac{d_{ij}^2}{\varepsilon_2^2} \chi_1 - \frac{d_{ij}^2}{\varepsilon_2^2} \chi_2 + \frac{d_{ij}^2}{\varepsilon_2^2} \chi_4, d_{2ij} = \frac{\chi_4 - \frac{d_{ij}^3}{\varepsilon_2^3} \chi_2 - \frac{d_{ij}^3}{\varepsilon_2} \chi_3}{1 - \frac{d_{ij}^2}{\varepsilon_2^2} \chi_2 + \frac{d_{ij}^2}{\varepsilon_2^2} \chi_4} \]

\[ E_{ij} = - c_{2ij} \varepsilon_2 - d_{2ij} \varepsilon_1, G_{ij} = c_{2ij} d_{ij} - d_{1ij} \varepsilon_1 \]

\[ H_{ij} = d_{2ij} c_{ij} - c_{1ij} \varepsilon_2, R_{ij} = c_{1ij} d_{ij} + d_{1ij} c_{ij} \]

\[ \kappa_{1ij} = - \chi_3 \left( - \frac{c_{ij}^2}{\varepsilon_1} \chi_1 + c_{ij} \chi_3 \right) - \chi_1 \left( \frac{c_{ij}^3}{\varepsilon_1^2} \chi_4 - \frac{c_{ij}^4}{\varepsilon_1^3} \chi_2 \right) \]

\[ \kappa_{2ij} = - \chi_3 \left( - \frac{d_{ij}^2}{\varepsilon_2} \chi_1 + d_{ij} \chi_3 \right) - \chi_1 \left( \frac{d_{ij}^3}{\varepsilon_2^2} \chi_4 - \frac{d_{ij}^4}{\varepsilon_2^3} \chi_2 \right) \]

\[ F_{ij} = \left( 1 + c_{1ij} \delta_x + c_{2ij} \delta_y^2 + d_{1ij} \delta_y + d_{2ij} \delta_y^2 \right) f_{ij} \]

Difference Equation (33) is expanded using Taylor’s formula to obtain the corresponding modified differential equation as follows:

\[ -au_{xx} - bu_{yy} + c(x, y) u_x + d(x, y) u_y - (2c_2 d_x + 2d_2 c_y) u_{xy} - 2c_2 c_x u_{xx} - 2d_2 d_y u_{yy} - (c_1 c_x + c_2 c_x + d_1 c_y + d_2 c_y) u_x - (c_1 d_x + c_2 d_x + d_1 d_y + d_2 d_y) u_y = f(x, y) + o(\eta + \omega) \quad (35) \]

where \( \eta = h_L h_R (h_L^2 + h_R^2), \omega = \frac{(h_R - h_L)(h_R^2 + h_L^2)}{60}. \)
To obtain the difference scheme with higher-order accuracy, adding the term
\[
[(2c_2 d_x + 2d_2 c_y)u_{xy} + 2c_2 c_x u_{xx} + 2d_2 d_y u_{yy} + (c_1 c_x + c_2 c_{xx} + d_1 c_y + d_2 c_{yy})u_x + (c_1 d_x + c_2 d_{xx} + d_1 d_y + d_2 d_{yy})u_y]
\]
to Equation (25) yields its modified differential equation,
\[
-Au_{xx} - Bu_{yy} + C(x,y)u_x + D(x,y)u_y = F(x,y)
\]
where
\[
A = \varepsilon_1 - 2c_2 c_x, C = c + c_1 c_x + c_2 c_{xx} + d_1 c_y + d_2 c_{yy} \\
B = \varepsilon_2 - 2d_2 d_y, D = d + c_1 d_x + c_2 d_{xx} + d_1 d_y + d_2 d_{yy} \\
F = f - (2c_2 d_x + 2d_2 c_y)u_{xy}
\]

Then, we apply (31) to (36), and the NRHOC scheme of two dimensional variable-coefficient convection–diffusion model Equation (25) is obtained as follows:
\[
\begin{align*}
\{-\Lambda_{xij}\delta_x^2 - \Lambda_{yij}\delta_y^2 + C_{ij}\delta_x + D_{ij}\delta_y + E_{ij}\delta_x^2 \delta_y^2 + G_{ij}\delta_x^2 \delta_y + H_{ij}\delta_x \delta_y + R_{ij}\delta_x \delta_y\}u_{ij} \\
= \{1 + C_{1ij}\delta_x + C_{2ij}\delta_x^2 + D_{1ij}\delta_y + D_{2ij}\delta_y^2\}F_{ij}
\end{align*}
\]
where the coefficients are
\[
\Lambda_{xij} = A_{ij} \left(1 + \frac{c_{ij}^2}{\Lambda_{xij}} \chi_1 - \frac{c_{ij}}{\Lambda_{xij}} \chi_3\right), \Lambda_{yij} = B_{ij} \left(1 + \frac{D_{ij}^2}{B_{ij}^2} \chi_1 - \frac{D_{ij}}{B_{ij}^2} \chi_3\right),
\]
\[
C_{1ij} = -\frac{c_{ij}}{\chi_2} \left(1 - \frac{C_{ij}^2}{\chi_4}\right), C_{2ij} = \frac{\chi_4 - \frac{C_{ij}^2}{\chi_4}}{1 - \frac{c_{ij}}{\chi_4}}
\]
\[
D_{1ij} = \frac{D_{ij}^2}{B_{ij}^2} \chi_2 + \frac{D_{ij}}{B_{ij}^4} \chi_4, D_{2ij} = \frac{\chi_4 - \frac{D_{ij}^2}{B_{ij}^2}}{1 - \frac{D_{ij}}{B_{ij}^4}}
\]
\[
E_{ij} = -C_{2ij}\Lambda_{yij} - D_{2ij}\Lambda_{xij}, C_{ij} = C_{2ij}D_{ij} - D_{1ij}\Lambda_{xij},
\]
\[
H_{ij} = D_{2ij}C_{ij} - C_{1ij}\Lambda_{xij}, R_{ij} = C_{1ij}D_{ij} + D_{2ij}C_{ij} + 2(c_{xij}d_{xij} + d_{xij}c_{yij})
\]
\[
\kappa_{1ij} = -\chi_3 \left(-\frac{c_{ij}^2}{\Lambda_{xij}} \chi_1 + C_{ij} \chi_3\right) - \chi_3 \left(-\frac{c_{ij}^3}{\Lambda_{xij}^3} \chi_1 - \frac{c_{ij}^4}{\Lambda_{xij}^4} \chi_2\right)
\]
\[
\kappa_{2ij} = -\chi_3 \left(-\frac{D_{ij}^2}{\Lambda_{yij}} \chi_1 + D_{ij} \chi_3\right) - \chi_3 \left(-\frac{D_{ij}^3}{\Lambda_{yij}^3} \chi_1 - \frac{D_{ij}^4}{\Lambda_{yij}^4} \chi_2\right)
\]
In particular, difference Equation (38) is the RHOC difference scheme on uniform grids of model Equation (25) when and only when $h_L = h_R$. 
4. Numerical Examples

In this section, we select several cases to further verify the high accuracy and validity of the NRHOC schemes and compare the results with those in the literature. For the one-dimensional linear problem, the Thomas algorithm is used to solve a system of tridiagonal linear equations directly. For one-dimensional nonlinear problems and two-dimensional problems, we use the BiCGSTab(2) iterative method \cite{39} to solve the discrete algebraic equations. The iterative process takes zero initial value and terminates when the condition
\[
\left\| u^{n+2} - u^n \right\|_2 \leq 10^{-14}
\]
is satisfied.

Above, \( n \) is the iteration number. The convergence order is calculated as
\[
\text{Rate} = \frac{\log(\text{err}(N_1)/\text{err}(N_2))}{\log(N_2/N_1)}
\]
where \( E(N_1) \) and \( E(N_2) \) are the maximum absolute errors calculated for two different grids \( N_1 \) and \( N_2 \).

For the one-dimensional problem, it is assumed that the defined interval to be solved is \([X_{\text{min}}, X_{\text{max}}]\), while the two-dimensional problem is the square region \([X_{\text{min}}, X_{\text{max}}] \times [Y_{\text{min}}, Y_{\text{max}}]\). We use the following grid generation function to obtain the computing grid nodes:
\[
x_i = X_{\text{min}} + i \frac{X_{\text{max}} - X_{\text{min}}}{N_x} + \frac{\lambda_x}{\pi} \sin \left( \pi \cdot i \frac{X_{\text{max}} - X_{\text{min}}}{N_x} \right)
\]
\[
y_j = Y_{\text{min}} + j \frac{Y_{\text{max}} - Y_{\text{min}}}{N_y} + \frac{\lambda_y}{\pi} \sin \left( \pi \cdot j \frac{Y_{\text{max}} - Y_{\text{min}}}{N_y} \right)
\]
where \( N_x \) and \( N_y \) are the numbers of sub-intervals along the x-direction and y-direction. \( \lambda_x \) and \( \lambda_y \) are the scaling parameters that control the density of grid points in the x-direction and y-direction, respectively. When \( -1 \leq \lambda_x, \lambda_y < 0 \), the computational nodes on the obtained nonuniform grid are dense near the left or lower boundary points of the interval; when \( 0 < \lambda_x, \lambda_y \leq 1 \), the computational nodes on the obtained nonuniform grid are dense near the right or upper boundary. The larger the absolute values of \( \lambda_x \) or \( \lambda_y \), the more grid points are concentrated in the area that needs to be dense, and vice versa.

When the grid-stretching parameter is equal to 0, the corresponding computational grid is a uniform grid. To represent uniformity, in the following numerical experiments, we use RHOC to represent the rational higher-order compact difference scheme on a uniform grid.

Example 1. \(-\varepsilon u_{xx} + \frac{1}{1+x} u_x = f(x), 0 < \varepsilon \leq 1, 0 < x < 1.\)

For this equation, the boundary conditions are \( u(0) = 1 + 2^{-1/\varepsilon}, u(1) = e + 2 \). The exact solution is \( u(x) = e^x + 2^{-1/\varepsilon} (1 + x)^{1+1/\varepsilon} \). \( f(x) \) is determined by the exact solution. When \( \varepsilon \) is small, the solution to the equation has a boundary layer at \( x = 1 \). Table 1 shows that both the FOC scheme and the NRHOC scheme are stable and effective when \( \varepsilon = 1, 10^{-2}, \) but when \( \varepsilon = 10^{-3} \) and \( \varepsilon = 10^{-5} \), the error of the FOC scheme is large and no longer decreases as the number of grids increases, while the NRHOC scheme can obtain very exact solutions with appropriate values of grid parameters and can reach the theoretical fourth order. From Figures 1 and 2, it is seen that the NRHOC scheme has obvious advantages over the FOC scheme in terms of computational accuracy and resolution.
Table 1. Comparison of maximum absolute error and convergence order, Example 1.

<table>
<thead>
<tr>
<th>ε</th>
<th>Nodes</th>
<th>FOC [7]</th>
<th>NRHOC</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>Error</td>
<td>Rate</td>
</tr>
<tr>
<td></td>
<td></td>
<td>(-8)</td>
<td>–</td>
</tr>
<tr>
<td>1</td>
<td>11</td>
<td>2.18</td>
<td>–</td>
</tr>
<tr>
<td></td>
<td>21</td>
<td>1.38</td>
<td>3.98</td>
</tr>
<tr>
<td></td>
<td>41</td>
<td>8.63</td>
<td>4.00</td>
</tr>
<tr>
<td>10^-2</td>
<td>21</td>
<td>3.35</td>
<td>–</td>
</tr>
<tr>
<td></td>
<td>41</td>
<td>2.83</td>
<td>3.57</td>
</tr>
<tr>
<td></td>
<td>81</td>
<td>1.65</td>
<td>4.10</td>
</tr>
<tr>
<td>10^-3</td>
<td>21</td>
<td>3.60</td>
<td>–</td>
</tr>
<tr>
<td></td>
<td>41</td>
<td>7.72</td>
<td>–1.10</td>
</tr>
<tr>
<td></td>
<td>81</td>
<td>3.05</td>
<td>1.34</td>
</tr>
<tr>
<td>10^-5</td>
<td>201</td>
<td>4.52</td>
<td>–</td>
</tr>
<tr>
<td></td>
<td>401</td>
<td>2.18</td>
<td>1.05</td>
</tr>
<tr>
<td></td>
<td>801</td>
<td>1.03</td>
<td>1.08</td>
</tr>
</tbody>
</table>

Figure 1. Comparison of computed and exact solutions, and error distribution (ε = 10⁻³, N = 41, λ = 0.95), Example 1.

Example 2. \(-\varepsilon u_{xx} + u u_x = f(x), 0 < \varepsilon \leq 1, 0 < x < 1.\)

The exact solution is \(u(x) = e^{-x} + e^{(x-1)(1+\varepsilon)}\), and the boundary conditions and source item \(f(x)\) are given by the exact solution. When \(\varepsilon\) is small, the solution has a boundary layer at \(x = 1\). We calculated the solution for parameter \(\varepsilon = 10^{-1}, 10^{-2}, 10^{-3}, 10^{-4}, 10^{-5}\) using three difference schemes, FOC [7], RHOC, and NRHOC. As shown in Table 2, when \(\varepsilon = 10^{-1}, 10^{-2}\), the computational accuracy of all schemes can reach the theoretical fourth-order accuracy, but when \(\varepsilon = 10^{-3}, 10^{-4}, 10^{-5}\), the computational errors of the FOC scheme and the RHOC scheme are large, while the NRHOC scheme can still obtain the fourth-order accuracy solution with appropriate values of the grid-scaling parameters.
Figure 2. Comparison of computed and exact solutions, and error distribution ($\varepsilon = 10^{-5}, N = 221, \lambda = 1$), Example 1.

Figures 3 and 4 give a comparison of the computed and exact solutions, and the error distribution over the computational region using the three difference schemes when $\varepsilon = 10^{-3}$ and $\varepsilon = 10^{-5}$, respectively. As shown in Figures 3 and 4, the computational solutions obtained with the FOC scheme and the RHOC scheme all produce nonphysical oscillations near $x = 1$, while the NRHOC scheme matches well with the exact solution because there are enough computational nodes distributed in the boundary layer.

Example 3. $-(u_{xx} + u_{yy}) + Re u_x = 0, \quad 0 \leq x, y \leq 1$.

The exact solution is

$$u(x, y) = e^{\frac{Re}{2} \sin(\pi y)} \left[ 2 e^{-\frac{Re}{2} \sinh(\sigma x)} \sinh(\sigma(1 - x)) \right] \sinh(\sigma x)$$

$$\sigma = \sqrt{\pi^2 + Re^2 / 4}$$

The boundary conditions are determined by the exact solution. When the Reynolds number ($Re$) is large, the solution to this equation has a boundary layer at $x = 0$. We calculated the solutions for Reynolds numbers $Re = 1, 10, 100, 1000, 5000, 7500$ using the NRHOC scheme and the HOC scheme [31]. Table 3 lists the computational error and convergence order when different values of $Re$ are taken. The computational data in the table show that compared with the HOC scheme, the NRHOC scheme has higher computational accuracy.
Figure 3. Comparison of computed and exact solutions, and error distribution ($\varepsilon = 10^{-3}$, $N = 81$, $\lambda = 1$), Example 2.
Figure 4. Comparison of computed and exact solutions, and error distribution ($\epsilon = 10^{-5}$, $N = 321$, $\lambda = 1$), Example 2.
Table 2. Comparison of maximum absolute error and convergence order, Example 2.

<table>
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<th>NRHOC</th>
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<td>Rate</td>
<td>Error</td>
<td>Rate</td>
</tr>
<tr>
<td>ε = 10^{-1}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>41</td>
<td>9.72 (−6)</td>
<td>–</td>
<td>1.16 (−5)</td>
</tr>
<tr>
<td>81</td>
<td>6.28 (−7)</td>
<td>3.95</td>
<td>7.68 (−7)</td>
</tr>
<tr>
<td>161</td>
<td>3.99 (−8)</td>
<td>3.98</td>
<td>4.89 (−8)</td>
</tr>
<tr>
<td>321</td>
<td>2.50 (−9)</td>
<td>4.00</td>
<td>3.07 (−9)</td>
</tr>
<tr>
<td>ε = 10^{-2}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>41</td>
<td>1.06 (−2)</td>
<td>–</td>
<td>2.03 (−2)</td>
</tr>
<tr>
<td>81</td>
<td>1.30 (−3)</td>
<td>3.03</td>
<td>7.50 (−4)</td>
</tr>
<tr>
<td>161</td>
<td>3.32 (−4)</td>
<td>3.03</td>
<td>3.74 (−4)</td>
</tr>
<tr>
<td>321</td>
<td>2.44 (−5)</td>
<td>3.76</td>
<td>3.03 (−5)</td>
</tr>
<tr>
<td>ε = 10^{-3}</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>41</td>
<td>6.17 (+0)</td>
<td>–</td>
<td>4.26 (+0)</td>
</tr>
<tr>
<td>81</td>
<td>1.93 (+0)</td>
<td>1.68</td>
<td>3.37 (+0)</td>
</tr>
<tr>
<td>161</td>
<td>2.59 (−1)</td>
<td>2.89</td>
<td>6.68 (−1)</td>
</tr>
<tr>
<td>321</td>
<td>1.52 (−2)</td>
<td>4.09</td>
<td>3.47 (−2)</td>
</tr>
<tr>
<td>ε = 10^{-4}</td>
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<td></td>
</tr>
<tr>
<td>81</td>
<td>3.70 (+1)</td>
<td>–</td>
<td>1.93 (+1)</td>
</tr>
<tr>
<td>161</td>
<td>1.83 (+1)</td>
<td>1.02</td>
<td>9.65 (+0)</td>
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<tr>
<td>321</td>
<td>8.43 (+0)</td>
<td>1.12</td>
<td>5.02 (+0)</td>
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<tr>
<td>641</td>
<td>3.03 (+0)</td>
<td>1.48</td>
<td>3.26 (+0)</td>
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<tr>
<td>ε = 10^{-5}</td>
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<td>160</td>
<td>1.86 (+2)</td>
<td>–</td>
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<td>320</td>
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<td>4.80 (+1)</td>
</tr>
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<td>640</td>
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<td>1.00</td>
<td>2.39 (+1)</td>
</tr>
<tr>
<td>1280</td>
<td>2.31 (+1)</td>
<td>1.01</td>
<td>1.20 (+1)</td>
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Table 3. Comparison of maximum absolute error and convergence order, Example 3.

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<th>NRHOC</th>
</tr>
</thead>
<tbody>
<tr>
<td>Error</td>
<td>Rate</td>
<td>Error</td>
</tr>
<tr>
<td>Re = 1, λ = 0.00</td>
<td></td>
<td></td>
</tr>
<tr>
<td>11 × 11</td>
<td>2.15 (−6)</td>
<td>–</td>
</tr>
<tr>
<td>21 × 21</td>
<td>1.29 (−7)</td>
<td>4.07</td>
</tr>
<tr>
<td>41 × 41</td>
<td>7.94 (−9)</td>
<td>4.02</td>
</tr>
<tr>
<td>Re = 100, λ = 0.85</td>
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<td></td>
</tr>
<tr>
<td>21 × 21</td>
<td>2.23 (−1)</td>
<td>–</td>
</tr>
<tr>
<td>41 × 41</td>
<td>5.07 (−2)</td>
<td>2.13</td>
</tr>
<tr>
<td>81 × 81</td>
<td>5.26 (−3)</td>
<td>3.27</td>
</tr>
<tr>
<td>Re = 1000, λ = 0.90</td>
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<td></td>
</tr>
<tr>
<td>128 × 64</td>
<td>8.16 (−1)</td>
<td>–</td>
</tr>
<tr>
<td>256 × 128</td>
<td>6.66 (−1)</td>
<td>0.29</td>
</tr>
<tr>
<td>512 × 256</td>
<td>4.43 (−1)</td>
<td>0.59</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Nodes</th>
<th>HOC [31]</th>
<th>NRHOC</th>
</tr>
</thead>
<tbody>
<tr>
<td>Error</td>
<td>Rate</td>
<td>Error</td>
</tr>
<tr>
<td>Re = 5000, λ = 0.945</td>
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</tr>
<tr>
<td>128 × 64</td>
<td>8.16 (−1)</td>
<td>–</td>
</tr>
<tr>
<td>256 × 128</td>
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<td>0.29</td>
</tr>
<tr>
<td>512 × 256</td>
<td>4.43 (−1)</td>
<td>0.59</td>
</tr>
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5. Conclusions

In this paper, we develop rational fourth-order compact difference schemes on uniform and nonuniform grids for convection–diffusion equations.

(1) The fourth-order rational compact difference scheme on uniform grids (the RHOC scheme) for one-dimensional convection–diffusion equations was developed by modifying the strongly dissipative polynomial fourth-order compact difference scheme of the one-dimensional convection–diffusion equation in reference [7]. The dispersion and dissipation errors of the RHOC scheme are significantly smaller than those of the FOC scheme; es-
especially, the larger Pe is, the more obvious the advantage of the RHOC scheme in terms of resolution.

(2) To efficiently solve the large-gradient-variation and boundary-layer problems, we extended the obtained difference scheme on uniform grids to nonuniform grids and constructed a rational higher-order compact difference scheme for one-dimensional convection–diffusion equations on nonuniform grids (the NRHOC scheme).

(3) A fourth-order rational compact difference scheme on nonuniform grids for two-dimensional convection–diffusion equations was developed by using the dimensionality reduction method and a partial differential equation correction technique.

(4) Several typical numerical cases were selected for solution to demonstrate the advantages of the NRHOC scheme proposed in this paper in terms of computational accuracy, validity, and stability. Numerical experiments show that the NRHOC scheme has better scale resolution and is more fit to solve convection-dominated problems precisely.

In follow-up research, we aim to focus on the following three aspects: (i) based on the present NRHOC difference scheme of convection–diffusion equations, conducting high-accuracy numerical simulations of flow and heat transfer problems [40]; (ii) combining the present scheme and adaptive methods to improve the computational accuracy of convection–diffusion with convective dominance [41]; (iii) theoretically analyzing the stability and convergence of the proposed differential scheme [42,43]. We look forward to reporting the results of the corresponding studies in the future.

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Data Availability Statement: No new data were created or analyzed in this study. Data sharing is not applicable to this article.

Conflicts of Interest: The authors declare no conflict of interest.

Appendix A

The 1D convection–diffusion equation with constant coefficients is

$$-\varepsilon u'' + cu' = f(x) \quad (A1)$$

where $\varepsilon$ and $c$ are constants and $\varepsilon > 0$.

Generally, we divide the interval [0,1] into N equal parts, and the coordinates of the inserted points are $x_i = ih$, where $h = x_{i+1} - x_i$ is the mesh step size. Let $u_i = u(x_i)$, $c_i = c(x_i)$, $f_i = f(x_i)$, $i \in \{0, 1, 2, ..., N\}$. Using Taylor’s formula,

$$\left( \frac{\partial u}{\partial x} \right)_i = \delta x u_i - \frac{h^2}{6} \left( \frac{\partial^3 u}{\partial x^3} \right)_i + O(h^4) \quad (A2)$$

$$\left( \frac{\partial^2 u}{\partial x^2} \right)_i = \delta^2 x u_i - \frac{h^2}{12} \left( \frac{\partial^4 u}{\partial x^4} \right)_i + O(h^4) \quad (A3)$$

where $\delta x u_i = \frac{u_{i+1} - u_i}{2h}$, $\delta^2 x u_i = \frac{u_{i+1} - 2u_i + u_{i-1}}{h^2}$ are the second-order central difference operators for the first- and second-order derivatives of $u(x_i)$.

Substituting (A2) and (A3) into (A1) yields

$$-\varepsilon \delta^2 x u_i + c_i \delta x u_i - \frac{c_i h^2}{6} \left( \frac{\partial^3 u}{\partial x^3} \right)_i + \frac{c_i h^2}{12} \left( \frac{\partial^4 u}{\partial x^4} \right)_i + O(h^4) = f_i \quad (A4)$$
Then, we directly calculate the third-order and fourth-order derivative functions with Equation (A1) to obtain

\[ u_{xxx} = \frac{c}{\epsilon} u_{xx} - \frac{1}{\epsilon^2} f \]

(A5)

\[ u_{xxxx} = \frac{c^2}{\epsilon^2} u_{xx} - \frac{C}{\epsilon^3} f - \frac{1}{\epsilon^4} f_{xx} \]

(A6)

We rewrite model Equation (6) as

\[ u_{xx} = \frac{c}{\epsilon} u_x - \frac{1}{\epsilon} f \]

(A7)

We substitute (A7) into (A6):

\[ u_{xxxx} = \frac{c^3}{\epsilon^3} u_x - \frac{c^2}{\epsilon^3} f - \frac{c}{\epsilon^4} f_x - \frac{1}{\epsilon^5} f_{xx} \]

(A8)

We substitute (A5) and (A8) into (A4):

\[ -\epsilon \delta^2_x u_i + \epsilon \delta_x u_i - \frac{c^2 h^2}{6 \epsilon} u_{xx i} + \frac{c^3 h^2}{12 \epsilon^2} u_{x i} + O(h^4) = \left(1 + \frac{c^2 h^2}{12 \epsilon^2} \right) f_i + \left( -\frac{c h^2}{12 \epsilon} \right) f_{xi} + \frac{h^2}{12} f_{xx i} \]

(A9)

Then, we substitute (A2) and (A3) into (A9):

\[ -\epsilon \left(1 + \frac{c h^2}{6 \epsilon} \right) \delta^2_x u_i + \epsilon \left(1 + \frac{c h^2}{12 \epsilon^2} \right) \delta_x u_i + \frac{c h^2}{72 \epsilon} \left( \frac{\partial^3 u}{\partial x^3} \right)_i - \frac{c^2 h^2}{12 \epsilon^2} \left( \frac{\partial^3 u}{\partial x^3} \right)_i + O(h^6) = \left(1 + \frac{c^2 h^2}{12 \epsilon^2} \right) f_i + \left(-\frac{c h^2}{12 \epsilon}\right) f_{xi} + \frac{h^2}{12} f_{xx i} \]

(A10)

We rewrite model Equation (A1) again as

\[ u_x = \frac{c}{\epsilon} u_{xx} + \frac{1}{\epsilon} f \]

(A11)

We calculate the second derivative directly:

\[ u_{xxx} = \frac{c^3}{\epsilon^3} u_x + \frac{1}{\epsilon} f \]

(A12)

By substituting (A12) into (A10) and omitting the higher-order terms, we obtain

\[ -\alpha \delta^2_x u_i + \epsilon \delta_x u_i = \left(1 + \frac{\alpha^2 h^2}{6 \epsilon^2} \right) f_i + \left( -\frac{c h}{12 \epsilon} \right) f_{xi} + \frac{h^2}{12} \left(1 + \frac{\alpha^2 h^2}{6 \epsilon^2} \right) f_{xx i} \]

Then, both sides of the above equation are divided by \( 1 + \frac{c h^2}{12 \epsilon^2} \) simultaneously to obtain a fourth-order compact difference equation for model Equation (A1).

\[ -\alpha \delta^2_x u_i + \epsilon \delta_x u_i = f_i + c_i f_{xi} + c_2 f_{xx i} \]

(A13)

where the coefficients are

\[ \alpha = \frac{1 + \frac{c h^2}{6 \epsilon^2}}{1 + \frac{c h^2}{12 \epsilon^2}}, \quad C_1 = -\frac{h}{12} \frac{c h}{12 \epsilon}, \quad C_2 = \frac{h^2}{12} \frac{1 + \frac{c^2 h^2}{6 \epsilon^2}}{1 + \frac{c^2 h^2}{12 \epsilon^2}} \]

(A14)

Appendix B

In order to explain the effect of the scheme modification in this paper in more detail, Fourier analysis is carried out for difference equation (A13). Fourier analysis, also known
as modified wave number analysis, provides an effective means to quantitatively analyze the resolution characteristics of different difference approximations. The Fourier transform and its inverse transform of the function defined in one-dimensional space are as follows:

\[
\hat{f}(k) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} f(x) e^{-ikx} \, dk
\]

\[
f(x) = \frac{1}{2\pi} \int_{-\infty}^{+\infty} \hat{f}(k) e^{ikx} \, dk
\]

where \( l = \sqrt{-1} \). Using the Fourier transform and Euler formula, the following imaginary and real parts of RHOC scheme (A13) are obtained:

\[
\kappa = \frac{1}{h} \sqrt{\frac{\epsilon Pe}{2}} \left( 1 + h^2 k_2^2 \left[ \frac{6 + \frac{\epsilon^2 Pe^2}{12 + \epsilon^4 Pe^2}}{36 + 3\epsilon^2 Pe^2} \right] k_1^2 - 4\epsilon h^2 \left[ \frac{6 + \frac{\epsilon^2 Pe^2}{12 + \epsilon^4 Pe^2}}{12 + \epsilon^4 Pe^2} \right] k_2 k_1 \right)
\]

\[
\kappa^2 = \frac{1}{h^2} \sqrt{\frac{\epsilon^2 Pe}{2}} \left( 1 + \frac{\epsilon^2 h^2 k_2^2}{6} \right) k_1^2 - \frac{\epsilon^2 h^4}{12 + \epsilon^4 Pe^2} k_2^2 - \epsilon \frac{\epsilon^2 h^2 k_2^2}{12 + \epsilon^4 Pe^2} k_1^2
\]

They are called effective (or numerical) wave numbers. Above, \( k_1 = \frac{\sin \kappa h}{\kappa}, k_2 = \frac{1 - \cos \kappa h}{h^2} \).

Using the same analysis method, the effective (or numerical) wave numbers of the FOC scheme are

\[
\kappa_{FOC} = \frac{1}{h} \sqrt{\frac{c^2 Pe}{2}} \left( 1 + \frac{h^2 k_2^2}{6} \right) k_1 - \frac{c^2 Pe}{6} \left( 1 + \frac{h^2 k_2^2}{12} \right) k_1 k_2
\]

\[
\kappa^2_{FOC} = \frac{1}{h^2} \sqrt{\frac{2c^2 Pe}{12}} \left( 1 + \frac{h^2 k_2^2}{6} \right) k_2 + \frac{c^2 Pe^2}{12} k_1^2
\]

Figures A1 and A2 show the relationship between the effective (or numerical) wave number and the exact wave number for different mesh Reynolds numbers (Pe) on the interval \( 0 \leq \kappa h \leq \pi \). The results show that in the case of \( Pe = 0.1 \) and \( Pe = 1 \), there is almost no difference between \( \kappa h \) (corresponding to the dispersion error) and \( \kappa^2 h^2 \) (corresponding to the dissipation error) of the RHOC scheme and the FOC scheme. However, when the mesh Reynolds number (Pe) equals 10 and 100, the dispersion error and dissipation error of the RHOC scheme are smaller than those of the FOC scheme. Especially, the larger the value of Pe, the more obvious the advantage of the RHOC scheme in resolution.

![Figure A1](image-url)
Figure A1. Graph of variable $s^2 h^2$ changing with $kh$: (a) $Pe = 0.1$, (b) $Pe = 1$, (c) $Pe = 10$, (d) $Pe = 100$.

Figure A2. Graph of variable $kh$ changing with $kh$: (a) $Pe = 0.1$, (b) $Pe = 1$, (c) $Pe = 10$, (d) $Pe = 100$.

References


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