

Novel Numerical Method Based on the Analog Equation Method for a Class of Anisotropic Convection-Diffusion Problems

L Zhang^{1,2}, FZ Wang^{3,4}, J Zhang^{3,5}*, YY Wang⁶, S Nadeem⁷ and TA Nofal⁸

¹Yellow River Institute of Hydraulic Research, YRCC, Zhengzhou, China, ²Henan Engineering Research Center of Hydropower Engineering Abrasion Test and Protection, Zhengzhou, China, ³Guangdong ATV Academy for Performing Arts, Dongguan, China, ⁴Nanchang Institute of Technology, Nanchang, China, ⁵School of Computer Science and Technology, Huaibei Normal University, Huaibei, China, ⁶School of Economics and Management, Nantong University, Nantong, China, ⁷Department of Mathematics, Quaid-i-Azam University, Islamabad, Pakistan, ⁸Department of Mathematic, College of Science, Taif University, Taif, Saudi Arabia

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*Correspondence:

J Zhang zj801106@163.com

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Zhang L, Wang FZ, Zhang J, Wang YY, Nadeem S and Nofal TA (2022) Novel Numerical Method Based on the Analog Equation Method for a Class of Anisotropic Convection-Diffusion Problems. Front. Phys. 10:807445. doi: 10.3389/fphy.2022.807445 In this work, a CMFS method based on the analogy equation method, the radial basis function and the method of fundamental solutions for linear and nonlinear convectiondiffusion equations in anisotropic materials is presented. The analog equation method is utilized to transform the linear and nonlinear convection-diffusion equation into an equivalent one. The expressions of the homogeneous solution and particular solution are derived by utilizing the radial basis function approximation and the method of fundamental solutions, respectively. By enforcing the desired solution to satisfy the original convection-diffusion equation with boundary conditions at boundary and internal collocation points yield a nonlinear system of equations, which can be solved by using the Newton-Raphson iteration or the Picard method of iteration. The error convergence curves of the proposed meshless method have been investigated by using different globally supported radial basis functions. Numerical experiments show that the proposed CMFS method is promising for anisotropic convection-diffusion problems with accurate and stable results.

Keywords: radial basis function, meshless method, convection-diffusion problems, nonlinear partial difference equation, boundary value problem

1 INTRODUCTION

Partial differential equations (PDEs) are generally utilized in understanding and modeling of a considerable lot of realism matters show up in applied science and material science. The PDE models are utilized in numerous fields, for example, plasma physics, hydrodynamics, finance, biology and nonlinear optic [1–3]. It is pointed that it is hard to tackle nonlinear problems most of the cases, especially analytically. Numerous strategies have been developed by the researchers for the numerical solution of complex problems (see [4–6] and the references therein). Among these techniques, one of the most attractive group of techniques is radial basis function (RBF) based techniques.

The RBF techniques are attractive in numerical simulation thanks to their simple, flexible, and truly meshfree features. These techniques have been successfully applied to diverse problems in a simple-to-implement fashion. The popular RBF-based numerical methods include the Kansa's method [7, 8], the method of fundamental solution (MFS) [9–12], the boundary knot method

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[13–15], the modified method of fundamental solutions [16–18] which have been well-developed and applied to a variety of boundary value problems.

In previous literatures, the polynomial RBFs, thin plate spline, Gaussians, and Multiquadrics are often used [19-21]. It is noted that the traditional RBF-based schemes are indirect and global in the sense that the expansion coefficients are used as the basic variables in the numerical solution procedure, while the global RBF interpolation leads to the full matrix. Roughly speaking, the RBF-based approaches can be classified as two types. The first category uses RBFs to approximate the particular solution of a partial differential equation (PDE) of interest, and the homogeneous part is obtained by means of numerical methods, like the boundary element method, the MFS, or the boundary knot method. The second category is the domain-type collocation methods. According to this approach, the RBF expansion is utilized directly for the unknown solution, and the collocation satisfies the governing equation and boundary conditions. The Kansa's method is a typical domain-type RBF approach. In this paper, we will develop a RBF-based technique of the first type to investigate anisotropic materials.

Anisotropic materials, characterized by varied material properties along different directions, are ubiquitous in nature and difficult to be analyzed. If non-linear property is included, the problems become even more complicated. Previously, Shin and Elman [22] studied the effect of various element discretization strategies and iteration algorithms for nonlinear convectiondiffusion problems with variable velocity in isotropic materials. Torsten [23] used the anisotropic streamline-diffusion finite element method to analyze homogeneous convection-diffusion problems with dominant convection. Onyejekwe [24] applied Green element method to 2D transient convection-diffusion problems with linear reaction and variable velocity. The lattice Boltzmann method is proposed for general nonlinear anisotropic convection-diffusion equations [25,26]. The anisotropic nonlinear convection-diffusion equations are also investigated by the finite volume method [27], the finite element method [28], the finite difference method [29], the virtual element method [30]. Shang et al. [31] proposed a discrete unified gas kinetic scheme for a general nonlinear convection-diffusion equation. Cao and Zhang studied a nonlinear diffusion-convectionreaction equation with a variable coefficient which has applications in many fields [32].

Based on the above-mentioned investigations, we aim to apply the MFS, in combination with the RBF and the analog equation method (AEM) [33], to analyze anisotropic nonlinear convection-diffusion problems. First, the AEM is utilized to transform the PDE into an equivalent one. Then, the expressions of the homogeneous and particular solutions are derived by utilizing RBF approximation and the MFS, respectively. Finally, enforcing the desired solution to satisfy the original PDE with boundary conditions at boundary and internal collocation points yield a nonlinear system of equations, which can be solved by using the Newton-Raphson iteration or the Picard method of iteration. Here, we will focus on the construction of Picard method of iteration, which should be carefully constructed to reach convergence. The structure of this paper is arranged as follows. In Section 2, we give the depiction of nonlinear convection-diffusion problems in anisotropic media. Followed in Section 3, the detailed processes are derived to construct the proposed CMFS. Numerical investigation and results analysis are carried out in Section 4. Section 5 concludes this paper and provides some comments on the CMFS.

2 NONLINEAR CONVECTION-DIFFUSION PROBLEMS IN ANISOTROPIC MATERIALS

Consider an open-bounded domain $\Omega \subseteq \mathbb{R}^d$, where *d* represent problem's dimension. Assuming that Ω is bounded by a piecewise smooth boundary Γ which is consist of numerous sufficiently smooth segments in the Liapunov sense. The space coordinate $X \in \Omega \subseteq \mathbb{R}^d$ is utilized to denoted position of an arbitrary point. The description of convection-diffusion problems in anisotropic media can be expressed as partial differential equation

$$\sum_{i=1}^{d} \sum_{j=1}^{d} k_{ij} \frac{\partial^2 u(X)}{\partial x_i \partial x_j} - \mathbf{v} \cdot \nabla u(X) = f(X), \quad X \in \Omega$$
(1)

subjected to Dirichlet boundary condition

$$u(X) = \bar{u}(X), X \in \Gamma_1$$
(2a)

and Neumann boundary condition related boundary flux

$$q_n(X) = \bar{q}_n(X), X \in \Gamma_2$$
(2b)

where **v** denotes a velocity vector and u is the desired field. **K** = $[k_{ij}]$ is the constant material tensor. Particularly, the material constant tensor for 2D cases is given by

$$\mathbf{K} = \begin{bmatrix} k_{11} & k_{12} \\ k_{12} & k_{22} \end{bmatrix}.$$

The boundary flux is defined as $q_n = -\sum_{i=1}^d \sum_{j=1}^d k_{ij} \frac{\partial u}{\partial x_j} n_i$, where n_i represent components of the unit outward normal vector **n** to the boundary Γ . f(X) stands for the internal forcing function, \bar{u} and \bar{q}_n are specified values on the Dirichlet and Neumann boundary, respectively. For a well-posed problem, $\Gamma = \Gamma_1 \cup \Gamma_2$.

It is observed that the smaller the determinant of **K**, the more asymmetric and anisotropic are the field and flux vectors, the more difficult it is to get an accurate numerical solution. Specially, if **K** is a diagonal matrix with completely equivalent diagonal elements, **Eq. 1** reduces to an isotropic convection-diffusion equation. In general, when material tensor **K** or velocity vector **v** is dependent on the unknown filed u, **Eq. 1** shows nonlinearity. Here, we restrict our attention to research such nonlinear cases, velocity **v** is considered as a function related to the desired field u, that is, $\mathbf{v} = \mathbf{v}(u)$.

3 THE CMFS MESHLESS METHOD

For the previous nonlinear problems, the proposed CMFS method is based on the combination of the AEM, globally

supported RBF approximation and the MFS. Detailed processes are given below.

3.1 The Analog Equation Method

The AEM, which was improved by Burlon et al [34], was first proposed by Katsikadelis for the solution of nonlinear problems. Using the analog equation method, **Eq. 1** can be converted into a Poisson type equation. Suppose u(X) is the sought solution of **Eq. 1**, which is a continuously differentiable function with up to two orders in Ω . Apply the Laplacian operator to u(X), we can derive

$$\nabla^2 u = b(X). \tag{3}$$

If the source distribution b(X) is known, then the solution of **Eq. 1** can be produced by solving the linear **Eq. 3** under the same boundary conditions (2a)-(2b), namely, **Eq. 3** is equivalent to **Eq. 1**.

Using to the principle of superposition, the solution of linear PDE Eq. 3 can be written in the form of the sum of the particular solution u^{par} and the homogeneous solution u^{hom} , which is given as

$$u = u^{\text{hom}} + u^{\text{par}},\tag{4}$$

Accordingly, u^{hom} and u^{par} should satisfy

$$\nabla^2 u^{\text{par}}(X) = b(X), X \in \Omega, \tag{5}$$

and

$$\begin{cases} \nabla^2 u^{\text{hom}}(X) = 0, & X \in \Omega\\ u^{\text{hom}}(X) = \bar{u} - u^{par}(X), & X \in \Gamma_1\\ q_n^{\text{hom}}(X) = \bar{q}_n - q_n^{par}(X), & X \in \Gamma_2 \end{cases}$$
(6)

respectively. Furthermore, we must note that the particular solution u^{par} satisfying **Eq. 5** does not need to satisfy boundary conditions, so it is not unique and must be coupled with the homogeneous solution u^{hom} and related boundary conditions. However, our aim is just to obtain the approximating expression rather than detailed numerical methods.

3.2 Radial Basis Function Approximation to the Particular Solution

This step is to derive the particular solution by RBF approximation. There are two schemes to fulfill this procedure. The standard approach to obtain the particular solutions is to integrate an operator by means of selected RBF. This scheme is just suitable to some simple operators and RBFs which is mathematically reliable. Another approach is a reverse differential process, which has no restriction on certain operators and RBFs, but the selected RBFs must be continuously differentiable with high orders. Here, the introduction of AEM replaces the original complicated operator with a simpler Laplacian operator, so it is feasible to evaluate the particular solution using the integrating process. To this end, the fictitious term introduced in **Eq. 3** can be approximated by

$$b(X) = \sum_{m=1}^{M} \alpha_m \phi(X, X_m), (X, X_m \in \Omega)$$
(7)

where *M* is the number of interpolation points inside the domain Ω . α_j (j = 1, 2, ..., M) denote coefficients to be determined, and ϕ represents the radial basis function.

Similarly, it is reasonable to express the particular solution $\boldsymbol{u}^{\mathrm{par}}$ as

$$u^{\text{par}}(X) = \sum_{m=1}^{M} \alpha_m \hat{u}(X, X)_m,$$
(8)

if the following relationship

$$\nabla^2 \hat{u}(X, X_m) = \phi(X, X_m), \tag{9}$$

holds, where \hat{u} is a corresponding particular basis function depending on the radial basis function ϕ and can be obtained by repeated integration.

The accuracy and efficacy of the interpolation depend on the choice of the radial basis function ϕ , which should provide an accurate approximation to b so that Eq. 9 can be derived analytically. During the past several decades, polynomial RBFs, MQs and TPS interpolation have got hot attention from the science and engineering communities. They have been investigated for Laplacian operator in R^2 and R^3 . It is noted that the MQs converge exponentially, TPS in $R^2O(h|\log h|)$, and traditional $1 + rO(h^{1/2})$, where h is minimum separation distance [35]. As is known to all, the better the approximation properties of RBFs corresponds with the worse conditioning. For example, MQs are spectrally convergent whereas their condition number increases exponentially when the data density increases. Additionally, the selection of shape parameter in MQs is also an important factor to approximation and its small variation may cause severe differences to solutions. Thus, we consider the implementation of polynomial RBFs and MOs approximation in this paper.

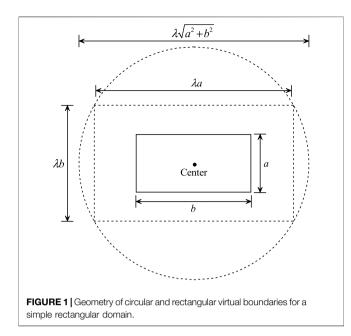
3.3 The Method of Fundamental Solutions for the Homogeneous Solution

Before introducing the MFS [36, 37], it is necessary to give the definition of the fundamental solutions. The fundamental solution $u^*(X, Y)$ for the Laplacian operator is defined to satisfy

$$\nabla^2 u^*(X,Y) + \delta_{XY} = 0, \qquad (10)$$

in an infinite domain, where δ_{XY} denotes the Dirac delta function which goes to infinity for the case of X = Y and equals to zero elsewhere. The detailed expression of $u^*(X,Y)$ for Laplace equation is

$$\begin{cases} u^*(X,Y) = \frac{1}{2\pi} \ln \frac{1}{r(X,Y)}, & d = 2\\ u^*(X,Y) = \frac{1}{4\pi r(X,Y)}, & d = 3 \end{cases}$$
(11)



The fundamental idea of the MFS is to place a virtual boundary outside the domain interested. Here, to obtain a weak solution of Laplace **Eq. 6**, collocation points X_n (n = 1, 2, L..., N) and source points Y_n (n = 1, 2, L..., N) are distributed on the physical boundary and the virtual boundary, respectively. Furthermore, suppose that there is a virtual source load φ_n $(1 \le n \le N)$ at each fictitious source point. The difference between the physical and virtual boundaries means that the homogeneous part u^{hom} at arbitrary field points X in the domain or on the physical boundary can be constructed by a linear combination of fundamental solutions in terms of fictitious sources φ_n $(1 \le n \le N)$, that is,

$$u^{\text{hom}}(X) = \sum_{n=1}^{N} \varphi_n u^*(X, Y_n), \left(X \in \Omega, Y_n \in \Gamma'\right)$$
(12)

which exactly satisfies Eq. 6.

The proper usage of the MFS must concern three problems. The first case is the number of collocation points distributed on the physical boundary. However, too many collocation points may aggravate the ill-conditioned matrix. The virtual boundary shape is another important aspect. Theoretically, the virtual boundary shape can be arbitrarily chosen in the calculation. In practical computation, the virtual boundary shape is usually selected as a circle or similar shape to the actual boundary to keep algorithm versatile [38]. For example, for the rectangular domain, the rectangular or circular virtual boundaries can be used (see **Figure 1**).

The location of the fictitious source points is also an interesting issue. It has been investigated in several literatures [39–41]. In this paper, we consider a new way to find out the proper location of the virtual boundary, a ratio parameter λ is introduced and can be defined as follows

$$\lambda = \frac{\text{characteristic length of the virtual boundary}}{\text{characteristic length of the physical boundary}}$$
(13)

For example, in Figure 1, if the length and height of the rectangle domain are a and b, respectively, the location of circular and similar rectangular virtual boundary can be measured by diameter $\lambda \sqrt{a^2 + b^2}$, λa and λb correspondingly. From the computation perspective, mathematical accuracy will decrease if the distance between the virtual and physical boundary turns out to be close, because of the singular disturbance of the fundamental solutions. Then again, in case when the source points are far away from the physical boundary then roundoff error in floating point arithmetic may be a major issue in such case, the coefficient matrix of the system of equations is closely to zero [38]. Therefore, parameter λ is commonly chosen to be in the range of 1.8-4.0 and 0.6-0.8 for internal and external problems respectively. Unless otherwise specified, a circular virtual boundary will be employed and parameter $\lambda = 3.0$ is used to determine its location in this paper.

3.4 The Construction of Solving Equations

According to the above process, the solution u = u(X) to Eq. 3 can be expressed as

$$u(X) = \sum_{n=1}^{N} \varphi_n u_n^*(X) + \sum_{m=1}^{M} \alpha_m \hat{u}_m(X), (X \in \Omega)$$
(14)

which is also the solution of Eq. 1. Differentiating Eq. 14 yields

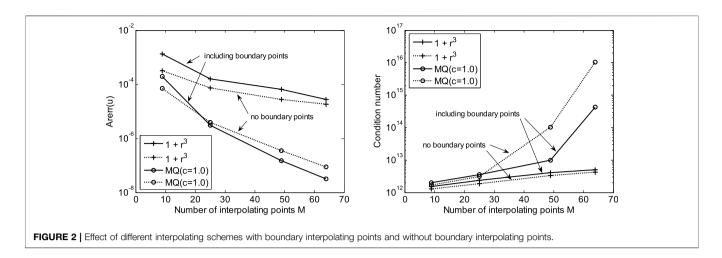
$$\frac{\partial u}{\partial x_i} = \sum_{n=1}^N \varphi_n \frac{\partial u_n^*(X)}{\partial x_i} + \sum_{m=1}^M \alpha_m \frac{\partial \hat{u}_m(X)}{\partial x_i},$$
(15)

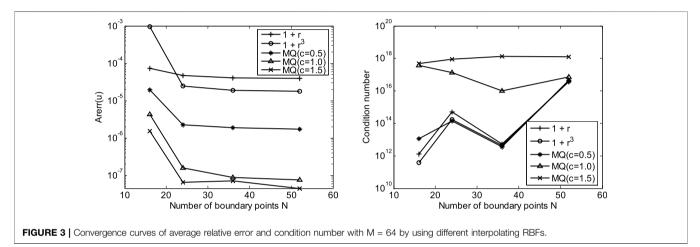
$$\frac{\partial^2 u}{\partial x_i \partial x_j} = \sum_{n=1}^N \varphi_n \frac{\partial u_n^*(X)}{\partial x_i \partial x_j} + \sum_{m=1}^M \alpha_m \frac{\partial \hat{u}_m(X)}{\partial x_i \partial x_j},$$
(16)

where i, j = 1, 2, ..., d and $u_n^*(X) = u^*(X, Y_n), \quad \hat{u}_m(X) = \hat{u}(X, X_m).$

There are two different approaches to determine the unknowns α_i and ϕ_i for nonlinear cases. If Eqs 14-16 directly satisfy the governing Eq. 1 at M interpolation points in the domain Ω and corresponding boundary conditions (2a)-(2b) at N boundary collocation points, a linear or nonlinear system of N + M equations can be obtained. The Newton-Raphson iteration [42] is a good choice to solve the nonlinear system. However, if we first linearize the nonlinear term, and then make Eqs. (14)-(16) satisfy Eq. (1) and boundary conditions (2a)-(2b) at related collocation points, the Picard method of iteration can be carried out. Compared to the Newton-Raphson iteration, the Picard method of iteration doesn't require computation of Jacobian matrix. However, the linearization of the nonlinear term in the Picard method of iteration is a key step. This process should be carefully considered to reach convergence. In the paper, the Picard method of iteration is constructed as follows.

(1) Assume an initial guess $u^{(0)}$ at M interpolating points





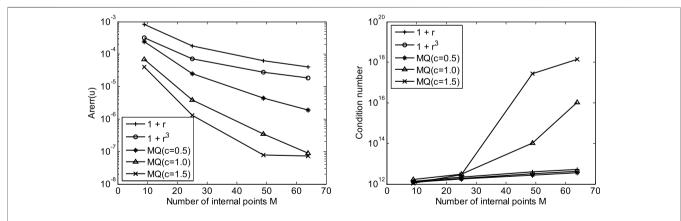


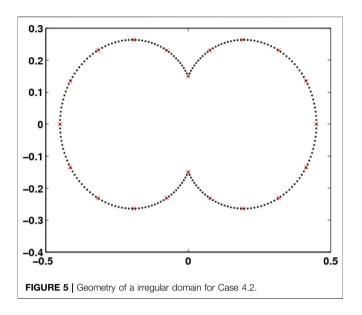
FIGURE 4 | Convergence curves of average relative error and condition number with N = 36 by using different interpolating RBFs.

(2) During an iteration k (k = 0, 1, 2, ...), given u^(k),
(a) Linearization of Eq. 1:

$$\sum_{i=1}^{d} \sum_{j=1}^{d} k_{ij} \frac{\partial^2 u(X)}{\partial x_i \partial x_j} + \mathbf{v} \left(u^{(k)} \right) \cdot \nabla u = f(X).$$
(17)

(b) Using the AEM-RBF-MFS to produce the following linear solving equations

 $\sum_{n=1}^{N} \varphi_n \left(\sum_{i=1}^{d} \sum_{j=1}^{d} k_{ij} \frac{\partial^2 u_n^*(X_s)}{\partial x_i \partial x_j} + \mathbf{v} (u^{(k)}) \cdot \nabla u_n^*(X_s) \right) \\ + \sum_{m=1}^{M} \alpha_m \left(\sum_{i=1}^{d} \sum_{j=1}^{d} k_{ij} \frac{\partial^2 \hat{u}_m(X_s)}{\partial x_i \partial x_j} + \mathbf{v} (u^{(k)}) \cdot \nabla \hat{u}_m(X_s) \right) \\ = f(X), X_s \in \Omega \text{ and } s = 1, ..., M$



$$\sum_{n=1}^{N} \varphi_{n} u_{n}^{*}(X_{l}) + \sum_{m=1}^{M} \alpha_{m} \hat{u}_{m}(X_{l}) = \bar{u}(X_{l}), X_{l} \in \Gamma_{1} \text{ and } l = 1, ..., N_{1},$$

$$\sum_{n=1}^{N} \varphi_{n} \left[\sum_{i=1}^{d} \sum_{j=1}^{d} k_{ij} \frac{\partial u_{n}^{*}(X_{l})}{\partial x_{j}} n_{i} \right] + \sum_{m=1}^{M} \alpha_{m} \left[\sum_{i=1}^{d} \sum_{j=1}^{d} k_{ij} \frac{\partial \hat{u}_{m}(X_{l})}{\partial x_{j}} n_{i} \right]$$

$$= -\bar{q}(X_{l}), X_{l} \in \Gamma_{2} \text{ and } l = 1, ..., N_{2}$$

- (c) The unknown coefficients φ_n and α_m can be obtained by solving the above system of linear equations.
- (d) Evaluating field values at M interpolating points by means of

$$u^{(k+1)}(X_s) = \sum_{n=1}^{N} \varphi_n u_n^*(X_s) + \sum_{m=1}^{M} \alpha_m \hat{u}_m(X_s), X_s \in \Omega \text{ and } s$$

= 1,..., M (18)

(e) Convergence criteria: if |u^(k+1) – u^(k)| ≤ ε, exit loop; else, let u^(k) = u^(k+1) for next iteration.

(3) Once the iteration converges, **Eq. 14** is used to evaluate quantities at arbitrary point in the domain and on the physical boundary.

4 NUMERICAL EXPERIMENTS

In this section, the convergence and accuracy of the CMFS are numerically examined by solving anisotropic nonlinear convectiondiffusion problems. Since the Kansa's method is a traditional RBFbased approach, comparisons are made between the CMFS and the Kansa's method. To measure the accuracy of the approximation, the relative error *rerr* (u), average relative error *Arerr* (u) are defined as below

$$rerr(u) = \left|\frac{u_j - \tilde{u}_j}{u_j}\right|_j, Arerr(u) = \sqrt{\frac{\sum_{j=1}^{L} (u_j - \tilde{u}_j)^2}{\sum_{j=1}^{L} (u_j)^2}}, \quad (19)$$

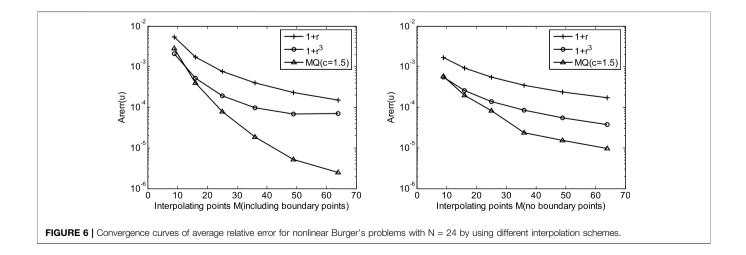
where u_j and \tilde{u}_j denote the analytical and numerical results, respectively. *L* is the total number of tested points. In our computation, tested points with number L = 10000 are uniformly distributed in the square domain.

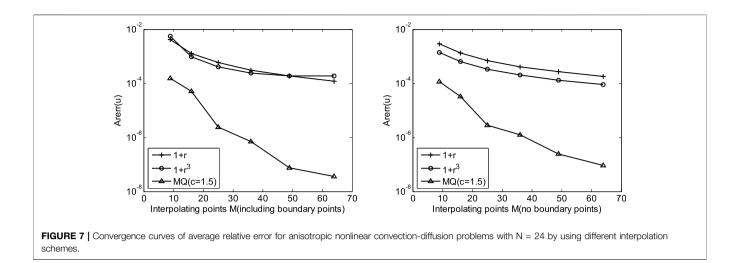
In order to investigate the condition number and convergence of the proposed CMFS, the linear anisotropic convection problem is tested in the first case, and then, nonlinear Burger's equation and nonlinear anisotropic convection-diffusion equation are subsequently examined.

4.1 2D Linear Anisotropic Convection-Diffusion Problems

We first consider the linear anisotropic convection-diffusion equation in a square domain $\Omega = \{(x, y) | 0 \le x, y \le 1\}$

$$\frac{\partial^2 u}{\partial x_1^2} + \frac{\partial^2 u}{\partial x_1 \partial x_2} + \frac{\partial^2 u}{\partial x_2^2} - \mathbf{v} \cdot \nabla u = 0, \qquad (20)$$





where $k_{11} = k_{22} = 1$ and $k_{12} = 1/2$ in the constant tensor **K**, and velocity vector **v** is given by $\mathbf{v} = \begin{bmatrix} -\frac{\sqrt{3}}{2} & -\frac{3+\sqrt{3}}{4} \end{bmatrix}^{\mathrm{T}}$.

The analytical solution shown is given as

$$u(x_1, x_2) = \exp\left(-\frac{\sqrt{3}}{2}x_1\right) + \exp\left(\frac{1}{2}x_1 - x_2\right), \quad (21)$$

for this case, which is also used to derive the Dirichlet boundary condition.

For the selection of interpolating points, different opinions always exist. **Figure 2** shows the average relative error of polynomial $1 + r^3$ and MQs whether interpolating points M include boundary collocation points or not. Form **Figure 2**, we can see the MQs is better than $1 + r^3$.

For definite internal points number M = 64, the solution accuracy of the CMFS with different RBFs and its condition numbers with increasing boundary points are shown in Figure 3. It is observed from Figure 3 that accuracy does not show significant improvement when the number of interpolating points maintains. Besides, Figure 4 displays the solution accuracy of the CMFS with different RBFs and its condition numbers with increasing boundary points in solving anisotropic convectiondiffusion problems when the number of boundary collocation points N = 36 is set. We can scrutinize higher convergence rate of the solution accuracy with increasing internal interpolating points than convergence rate with increasing boundary points. Compared to conventional polynomial RBFs, the MQs generally leads to huge condition number of 10¹⁸ with corresponding high solution accuracy 10⁻⁷. As mentioned in pervious literatures, the better the approximation properties of the RBFs, the worse the conditioning number [43]. In addition, we note that the solution accuracy by using MQs is better than using polynomial RBFs with two or three orders of magnitude. The choice of the RBFs shape parameter is investigated in some literatures [44].

4.2 Nonlinear Inviscid Burger's Equation

In this case, the nonlinear form for steady-state situation is considered

$$\nabla^2 u + u \frac{\partial u}{\partial x} = 0, \qquad (22)$$

where the variable *u* represents a velocity term. A particular solution for this problem is $u(x, y) = \frac{2}{x^3}$ which is also the exact solution when imposed as a boundary condition. An peanut irregular domain is considered for this case (**Figure 5**). The initial guess of $u^{(0)}$ is selected as one.

Figure 6 shows the convergence curves of the CMFS using the Picard method of iteration with different RBFs. All convergence curves can be found with the increase of internal collocation points M. Meanwhile, we also observe that MQ can reach more accuracy when a certain boundary points are employed to implement RBF approximation, instead of pure interior points. However, this phenomenon seems to disappear for 1 + r and $1 + r^3$. Furthermore, the usage of pure internal interpolating points improves accuracy in the case of small M. The similar fact that MQ have better accuracy than polynomials is observed, this is eliminated in this case.

4.3 Nonlinear Anisotropic Convection-Diffusion Problem

Consider a 2D anisotropic convection-diffusion problem depicted by

$$k_{11}\frac{\partial^2 u}{\partial x^2} + 2k_{12}\frac{\partial^2 u}{\partial x \partial y} + k_{22}\frac{\partial^2 u}{\partial y^2} + u\left(\frac{\partial u}{\partial x} + \frac{\partial u}{\partial y}\right) = f(x, y), \quad (23)$$

with $k_{11} = k_{22} = 1$ and $k_{12} = 0.5$ in the same square domain as Example 4.2. The analytical expression $u = 0.5 \sin(x + y)$ is used with Dirichlet boundary only, the right-handed source function is chosen to be

$$f(x, y) = -1.5\sin(x + y) + 0.25\sin[2(x + y)].$$
(24)

Here, the similar Picard iteration scheme is employed and the initial guess of u also is set to one. Figure 7 displays the average relative error curve at 1,000 test points against the different interpolating collocation schemes. It is found that whether the

boundary collocation is included or not, the CMFS convergence curve converges stably and quickly in solving the nonlinear anisotropic convection-diffusion problem than the polynomial RBFs 1 + r and $1 + r^3$. And the solution accuracy of the CMFS is averagely three orders of magnitude larger than the polynomial RBFS 1 + r and $1 + r^3$. However, the instability for RBF $1 + r^3$ should be observed when the RBF approximation involves the usage of the boundary collocation points.

5 CONCLUSION

In this paper, the CMFS, which is composed by the analog equation method, radial basis function approximation, and the method of fundamental solutions, is applied to the nonlinear anisotropic convection-diffusion problems. Numerical results reveal the efficiency and stability of the CMFS for the tested three cases. In a word, the proposed CMFS has the following features:

- 1) For linear cases, the approach is just one-step simple scheme. Meanwhile, no inverse of a matrix is involved.
- 2) The simple fundamental solution of Laplacian operator is employed, rather than one of the original PDE.
- 3) There are simple process and theoretical basis, so it is easy to program.
- 4) The related Picard iteration process is developed for nonlinear cases in isotropic and anisotropic media, respectively.

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- 5) The proposed method is a truly meshfree method and no integrals are needed.
- 6) It can be easily extended to solve anisotropic unhomogeneous problems with variable parameter or the other problems [45].

DATA AVAILABILITY STATEMENT

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

AUTHOR CONTRIBUTIONS

LZ supported found of the manuscript; FW developed the conceptualization, and wrote the manuscript; JZ worked on the mathematical development, performed the numerical analysis; SN and TN analyzed the data, and wrote the manuscript. All authors have read and agree to the published version of the manuscript.

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