ORIGINAL ARTICLE

A coarse‑grid projection method for accelerating incompressible MHD fow simulations

Ali Kashef1,2

Received: 28 February 2020 / Accepted: 23 December 2020 / Published online: 2 February 2021 © The Author(s), under exclusive licence to Springer-Verlag London Ltd. part of Springer Nature 2021

Abstract

Coarse grid projection (CGP) is a multiresolution technique for accelerating numerical calculations associated with a set of nonlinear evolutionary equations along with stif Poisson's equations. In this article, we use CGP for the frst time to speed up incompressible magnetohydrodynamics (MHD) fow simulations. Accordingly, we solve the nonlinear advection–diffusion equation on a fne mesh, while we execute the electric potential Poisson equation on the corresponding coarsened mesh. Mapping operators connect two grids together. A pressure correction scheme is used to enforce the incompressibility constrain. The study of incompressible fow past a circular cylinder in the presence of Lorentz force is selected as a benchmark problem with a fxed Reynolds number but various Stuart numbers. We consider two diferent situations. First, we only apply CGP to the electric potential Poisson equation. Second, we apply CGP to the pressure Poisson equation as well. The maximum speed-up factors achieved here are approximately 3 and 23, respectively, for the frst and second situations. For the both situations, we examine the accuracy of velocity and vorticity fields as well as the lift and drag coefficients. In general, the results obtained by CGP are in an excellent to reasonable range of accuracy. The CGP results are signifcantly more accurate compared to the numerical simulations of the advection–difusion and electric potential Poisson equations on pure coarse scale grids.

Keywords Coarse grid projection · Multiresolution schemes · Magnetohydrodynamics · Ohm's law · Lorentz's force · Poisson's equation · Flow past a cylinder

1 Introduction

The study of external and internal [[12](#page-13-0)] magnetohydrodynamics (MHD) flows has been the subject of interest in scientifc areas (see, e.g., Mutschke et al. [[19](#page-13-1)]; Klein and Pothérat [[15](#page-13-2)]; Farooq et al. [\[4](#page-13-3)]) as well as industrial felds (see, e.g., Josserand et al. [\[10](#page-13-4)]; Irfan and Farooq [[8\]](#page-13-5); Shoaib et al. [\[26](#page-14-0)]). Numerous researchers have taken advantages of computational tools for the MHD fow simulations in desired conditions (Verron and Sommeria [\[27](#page-14-1)]; Mück et al. [\[18\]](#page-13-6); Yoon et al. [[28](#page-14-2)]; Lee et al. [[16\]](#page-13-7); Dousset and Pothérat [\[3\]](#page-13-8); Shah et al. $[25]$ $[25]$; Peng et al. $[20]$ $[20]$). From a mathematical perspective, one has to deal with a Poisson equation for

 \boxtimes Ali Kashefi kashef@stanford.edu

Department of Civil and Environmental Engineering, Stanford University, Stanford, CA 94305, USA

an electric potential feld to obtain a numerical solution to the solenoidal current feld governed by the Ohm law (see e.g., Section 2 of Lee et al. [[16\]](#page-13-7)). The challenge ahead is the fact that the elliptic Poisson equation is the most time-consuming component of the numerical procedure and demands high computational expenses. An approach to substantially lessen the central processing unit (CPU) times associated with the linear electric potential Poisson equation is using multiresolution techniques. To this purpose, we examine one of the recently introduced techniques, called Coarse Grid Projection (CGP) methodology (Lentine et al. [\[17](#page-13-10)]; San and Staples [[23\]](#page-14-4); San and Staples [\[24](#page-14-5)]; Jin et al. [[9\]](#page-13-11); Kashefi and Staples $[14]$ $[14]$; Kashefi $[13]$ $[13]$; Kashefi $[11]$ $[11]$), for the MHD flow computations for the frst time in this research paper.

Historically, the CGP algorithm was presented frst by Lentine et al. [[17\]](#page-13-10), when they applied this multiresolution method to inviscid fow simulations for video games. San and Staples [\[23\]](#page-14-4) implemented CGP in numerical computations of transient incompressible viscous flows for high Reynolds numbers. Furthermore, they used CGP for

¹ Department of Mechanical Engineering, Stanford University, Stanford, CA 94305, USA

quasigeostrophic ocean simulations (San and Staples [[24\]](#page-14-5)). Jin et al. [\[9](#page-13-11)] calibrated the fast fuid dynamics (FFD) models using CGP. Kashefi and Staples [\[14\]](#page-13-12) proposed a specific version of CGP for unstructured triangular fnite-element grids. Kashefi [\[13](#page-13-13)] discussed CGP as a partial mesh refinement tool for the incompressible fow simulations. Kashef [\[11](#page-13-14)] constructed a CGP framework for incremental pressure correction schemes. Lastly, the efect of CGP on the energy equation was studied by Kashefi $[12]$ $[12]$.

Generally, calculations relevant to the velocity feld are performed on a grid with fne spatial resolutions, mainly due to the existence of the nonlinear convection term in the Navier–Stokes equations. On the other hand, the velocity feld appears in the right-hand side of the electric potential Poisson equation in the MHD fows. By applying the CGP approach to this set of equations, instead of solving the Poisson equation on the fne mesh, the velocity feld is restricted onto a corresponding coarsened grid and the potential Poisson equation is solved on the coarsened grid. Since the degree of freedom of Poisson's equation decreases, a considerable amount of CPU times is saved. The resulting electric potential is prolonged onto the fne grid and the Lorentz force is set as the source term of the Navier–Stokes equations. Note that Kashefi and Staples [[14](#page-13-12)] demonstrated that the CGP approach preserves the accuracy of the pressure gradient feld, instead of the pressure itself. On the other hand, the pressure gradient appears as an explicit variable in the Navier–Stokes equations. Hence, the velocity feld fdelity is preserved in CGP as well. Making an analogy, the electric potential is also an implicit variable in the Lorentz force, which means that we deal with the gradient of the electric potential feld, instead of the electric potential itself. Thus, it is expected that CGP preserves the accuracy of the Lorentz force, and consequently, the accuracy of the velocity feld. It is worthwhile to note that we use a non-incremental pressure projection scheme to execute the incompressibility constraint of the velocity feld. As a result, at each time step, a set of three decoupled elliptic equations has to be numerically solved: a linear electric potential Poisson equation, a nonlinear advection–difusion equation, and a linear pressure Poisson equation. We investigate the integrity of the unsteady incompressible MHD fows when CGP is only applied to the potential Poisson equation or simultaneously to both the potential and pressure Poisson equations.

The study of the flow around bluff bodies has been long undertaken for scientifc purposes as well as engineering applications (Zhao and Lu $[29]$ $[29]$; Ganta et al. $[5]$ $[5]$). As regards the MHD flows, numerous experimental and computational observations for the fuid fow past an obstacle have been reported in the literature (Mutschke et al. [\[19\]](#page-13-1); Yoon et al. [\[28](#page-14-2)]; Dousset and Pothérat [[3\]](#page-13-8); Ghosh et al. [[7\]](#page-13-16)). Due to the importance of this fundamental fuid mechanics problem, we peak the two-dimensional incompressible MHD fows

around a circular cylinder as the benchmark test case for the present numerical study. We perform the numerical simulations for a fxed Reynolds number and diferent values of the Stuart number. For each case, we evaluate the CPU time, lift and drag evolutions, and von Karman vortex street structure for the CGP versus non-CGP computations. We use the most generalized version of the CGP framework introduced by Kashefi and Staples $[14]$ $[14]$. In this way, we examine the CGP performance for the electric potential Poisson equation in an unstructured triangular fnite-element mesh.

This article is structured as follows. We present the governing equations of the fow feld in the presence of a magnetic feld in Sect. [2.1.](#page-1-0) Coarse grid projection methodology is provided in Sect. [2.2.](#page-2-0) Computational aspects of the present study are discussed in Sect. [2.3](#page-2-0). Resulting simulations with graphical and numerical explanations are illustrated in Sect. [3.](#page-4-0) Finally, conclusions of the work are given in Sect. [4.](#page-11-0)

2 Problem formulation

2.1 Governing equations

We consider an incompressible isothermal flow affected by the Lorentz force. Hence, the equations of the conservation of charge, momentum, and mass govern the fuid fow are as follows:

$$
\nabla \cdot \mathbf{j} = 0,\tag{1}
$$

$$
\rho \left[\frac{\partial u}{\partial t} + (u \cdot \nabla) u \right] - \mu \Delta u + \nabla p = \mathbf{j} \times \mathbf{B},\tag{2}
$$

$$
\nabla \cdot \mathbf{u} = 0,\tag{3}
$$

where u , j , and B denote the vector of the velocity, the current density, and the static magnetic feld, respectively. *p* stands for the pressure. ρ represents the fluid flow density and μ indicates the dynamic viscosity.

Ohm's law [[1\]](#page-13-17) says

$$
\mathbf{j} = \sigma(-\nabla\phi + \mathbf{u} \times \mathbf{B}),\tag{4}
$$

where ϕ is the electric potential and σ is the electric conductivity.

Substituting Eq. [\(4](#page-1-1)) in Eq. [\(1\)](#page-1-2), an electric potential Poisson equation is derived as:

$$
\Delta \phi = \nabla \cdot (\mathbf{u} \times \mathbf{B}). \tag{5}
$$

From a time integration point of view, we explicitly deal with the current density and the Lorentz force, while a semi-implicit frst-order scheme is used for the velocity and pressure felds. In the next stage, a non-incremental pressure correction scheme (Chorin [[2\]](#page-13-18)) is applied to the Navier–Stokes and mass continuity equations. That procedure yields to

$$
\Delta \phi^n = \nabla \cdot (\boldsymbol{u}^n \times \boldsymbol{B}^n),\tag{6}
$$

$$
\mathbf{j}^n = \sigma(-\nabla \phi^n + \mathbf{u}^n \times \mathbf{B}^n),\tag{7}
$$

$$
\rho \left[\frac{\tilde{u}^{n+1} - u^n}{\delta t} + (u^n \cdot \nabla) \tilde{u}^{n+1} \right] - \mu \Delta \tilde{u}^{n+1} = j^n \times B^n, \tag{8}
$$

$$
\Delta p^{n+1} = \frac{\rho}{\delta t} \nabla \cdot \tilde{\boldsymbol{u}}^{n+1},\tag{9}
$$

$$
\boldsymbol{u}^{n+1} = \tilde{\boldsymbol{u}}^{n+1} - \frac{\delta t}{\rho} \nabla p^{n+1},\tag{10}
$$

where δt indicates the time step and $\tilde{\mathbf{u}}^{n+1}$ is the intermediate velocity vector at time t^{n+1} .

2.2 Coarse grid projection methodology

The CGP scheme potentially is able to accelerate simulations associated with a set of decoupled equations if the following two conditions are satisfed. First, at least one linear equation exists in this set. Second, the unknown variable of the equation must appear implicitly (e.g., its gradient) in other equations of the set. If the conditions are satisfed, then CGP dramatically lowers the computational cost by executing the equation on a coarsened mesh while preserves the accuracy of variables presented explicitly in other equations. For the MHD flows, the electric potential Poisson equation is the linear elliptic equation and the gradient of the electric potential appears in the momentum equation, not the electric potential itself. Hence, these two conditions are satisfed. Note that the Poisson equation is the most time-consuming part of the simulation. Figure [1](#page-2-1) gives a schematic illustration of the CGP mechanism. As shown in Fig. [1,](#page-2-1) at each time step t^n , we map the velocity field of a fine grid u_f^n onto a corresponding coarsened grid and set u_c^n . The restricted velocity feld is used to solve Poisson's equation for the electric potential field on the coarsened grid ϕ_c^n . Then, we remap the resulting potential data ϕ_c^n onto the fine grid and set ϕ_f^n . The gradient of ϕ_f^n plays a role in the source term of the momentum equation.

In principle, the CGP scheme described can be potentially applied to three-dimensional MHD fows. In this article, however, we restrict our study to the two-dimensional flows

Fig. 1 Scheme of coarse grid projection methodology involving the restriction and prolongation of the velocity and electric potential data. Some parts of this fgure are reproduced from Fig. 1 of Ref. [\[14\]](#page-13-12)

in the $x - y$ plane with a constant magnetic field parallel to the *z*−axis such that:

$$
\mathbf{B} = \mathbf{B}_0 \hat{k},\tag{11}
$$

where B_0 is the constant. This condition has been applied to the two-dimensional confined jet flows by Lee et al. [\[16](#page-13-7)].

Equations (12) – (17) (17) (17) depict the CGP algorithm for the simulation of the two-dimensional incompressible MHD flow at each time step.

1. Restrict u_f^n and v_f^n onto the coarsened grid and obtain u_c^n and v_c^n .

2. Calculate ϕ_c^n on the coarsened mesh by solving

$$
\frac{\partial^2 \phi_c^n}{\partial x^2} + \frac{\partial^2 \phi_c^n}{\partial y^2} = B_o \left(\frac{\partial v_c^n}{\partial x} - \frac{\partial u_c^n}{\partial y} \right).
$$
 (12)

3. Prolong ϕ_c^n onto the fine grid and obtain ϕ_f^n .

4. Calculate \tilde{u}^{n+1}_f and \tilde{v}^{n+1}_f on the fine mesh by solving

means that CGP is only used for the electric potential Poisson equation.

2.3 Computational consideration

The CGP method and its relevant mapping functions depend signifcantly on the spatial discretization scheme. Here, we use the CGP framework as well as the restriction and prolongation operators introduced by Kashefi and Staples [[14\]](#page-13-12) for unstructured finite-element grids. This CGP confguration has been successfully utilized in the literature (Kashefi $[11]$ $[11]$; Kashefi $[13]$ $[13]$). Thus, we discretize the system of Eqs. ([12\)](#page-3-0)–[\(17\)](#page-3-1) using the Galerkin fnite-element method (Reddy $[21]$ $[21]$) with linear P_1 shape functions. According to the numerical studies investigated in San and Staples [\[23](#page-14-4)], Kashefi and Staples $[14]$, and Kashefi $[11]$, CGP was successful for up to three levels of the pressure Poisson grid coarsening. Following the experience, we practically consider four sequences of nested hierarchical grids, which in

$$
\rho \left(\frac{\tilde{u}_f^{n+1} - u_f^n}{\delta t} + u_f^n \frac{\partial \tilde{u}_f^{n+1}}{\partial x} + v_f^n \frac{\partial \tilde{u}_f^{n+1}}{\partial y} \right) - \mu \left(\frac{\partial^2 \tilde{u}_f^{n+1}}{\partial x^2} + \frac{\partial^2 \tilde{u}_f^{n+1}}{\partial y^2} \right) = \sigma B_0 \left(-\frac{\partial \phi_f^n}{\partial y} - B_0 u_f^n \right) \tag{13}
$$

$$
\rho \left(\frac{\tilde{v}_f^{n+1} - v_f^n}{\delta t} + u_f^n \frac{\partial \tilde{v}_f^{n+1}}{\partial x} + v_f^n \frac{\partial \tilde{v}_f^{n+1}}{\partial y} \right) - \mu \left(\frac{\partial^2 \tilde{v}_f^{n+1}}{\partial x^2} + \frac{\partial^2 \tilde{v}_f^{n+1}}{\partial y^2} \right) = \sigma B_o \left(\frac{\partial \phi_f^n}{\partial x} - B_o v_f^n \right) \tag{14}
$$

5. Restrict \tilde{u}_f^{n+1} and \tilde{v}_f^{n+1} onto the coarsened grid and obtain \tilde{u}_c^{n+1} and \tilde{v}_c^{n+1} .

6. Calculate p_c^{n+1} on the coarse mesh by solving

$$
\frac{\partial^2 p_c^{n+1}}{\partial x^2} + \frac{\partial^2 p_c^{n+1}}{\partial y^2} = \frac{\rho}{\delta t} \left(\frac{\partial \tilde{u}_c^{n+1}}{\partial x} + \frac{\partial \tilde{v}_c^{n+1}}{\partial y} \right). \tag{15}
$$

7. Prolong p_c^{n+1} onto the fine grid and obtain p_f^{n+1} . 8. Calculate u_f^{n+1} and v_f^{n+1} on the fine mesh via

$$
u_f^{n+1} = \tilde{u}_f^{n+1} - \frac{\delta t}{\rho} \frac{\partial p_f^{n+1}}{\partial x},\tag{16}
$$

$$
v_f^{n+1} = \tilde{v}_f^{n+1} - \frac{\delta t}{\rho} \frac{\partial p_f^{n+1}}{\partial y}.
$$
 (17)

Subscripts *f* and *c* indicate the value of the relevant variable on a fne gird and on a corresponding coarsened grid, respectively. *u* and *v* are, respectively, the velocity felds in *x* and *y* directions, while \tilde{u} and \tilde{v} are, respectively, the intermediate velocity felds in *x* and *y* directions. Note that one may omit steps (5) – (7) in the preceding algorithm, which each fne mesh with *M* element numbers is generated by uniformly subdividing elements of a coarse grid with *P* elements. In this way, *M* and *P* are correlated together with

$$
P = 4^{-k}M,\tag{18}
$$

where *k* is the coarsening level.

We illustrate the restriction and prolongation functions addressed in the steps (1) , (3) , (5) , and (7) of the CGP algorithm. To save space, we describe the mapping technique in a general format. One may apply this scheme to the restriction of the velocity or intermediate velocity felds and to the prolongation of the electric potential or the pressure felds. Let us take a fine finite-element space V_j with the dataset $\{q_f\}$ resolved on that. The plan is to restrict the data into a coarse finite-element space V_{j-k} to obtain the dataset $\{q_c\}$. Note that *k* is a positive integer number indicating the coarsening level as mentioned earlier. Moreover, the dimension of each dateset is equal to the total number of degrees of freedom of the corresponding space. Obviously, the dimension of $\{q_f\}$ is greater than ${q_c}$. Now consider two random nodes α and β which belong, respectively, to spaces V_j and V_{j-k} . We enforce

$$
q_c(\beta) = q_f(\alpha) \quad \text{if} \quad \mathbf{x}(\beta) = \mathbf{x}(\alpha), \tag{19}
$$

where $\mathbf{x}(\beta)$ is the position vector of node β on the space V_{i-k} **x**(α) is the position vector of node β on the space V_i . Notice that all fnite-element spaces are located in one coordinate system. In other words, we simply use an injection procedure to perform the restriction. Taking the advantage of this interpolation, one may directly restrict the data from a desired space onto another non-nested space, which reduces the associated computational cost. Now we explain the prolongation platform. The goal is to prolong the dataset $\{q_c\}$ of the coarse space V_{j-k} onto the fine space V_j to set $\{q_f\}$. In fnite-element applications, a possible strategy for the construction of the prolongation operator is the implementation of the fnite-element shape function. In the current study, we use linear shape functions. Thus, we prolong the data linearly. For any node α on V_{i-k+1} , there exists a linear triangular element *E* with three nodes β_j ($1 \le j \le 3$) in the space of *V*_{*j*−*k*} such that $\mathbf{x}_{\alpha} \in V_E \subset V_{j-k}$. Thus $q_f(\alpha)$ can be computed as a linear combination of the nodal values of V_E such that

$$
q_f(\alpha) = \sum_{j=1}^3 q_c(\beta_j) \psi_j(\mathbf{x}_{\alpha}), \qquad (20)
$$

where ψ_j is a linear shape function of element *E*. In contrast with the restriction function, we can only prolong the data from a desired space V_{j-k} to the nested space V_{j-k+1} . For instance, one has to run the prolongation function three times for a simulation with three levels of coarsening $(k = 3)$. That is why the restriction operator is more costly in comparison with the prolongation operator. One may see

Sect. [2.3](#page-3-2) of Kashefi and Staples [[14\]](#page-13-12) for further elaboration. The efficiency of CGP strongly depends on the design of the restriction and prolongation operators. However, we demonstrate that the CGP methodology is quite profcient even using these basic data interpolation/extrapolation.

An object-oriented C++ code is developed. The GMRES(m) technique with $ILU(0)$ preconditioner [\[22](#page-14-8)] is used to numerically solve Eqs. (12) (12) (12) – (15) (15) . Gmsh [[6\]](#page-13-19) is used as the fnite-element mesh generator. All simulations are performed on a single Intel (R) Xeon (R) processor with 2.66 GHz clock rate and 64 Gigabytes of RAM.

3 Results and discussion

A rectangular box $[0, L] \times [0, H]$ is considered as the computational domain. A circle with a diameter of *d* and center of (x_c, y_c) represents a rigid circular cylinder with no-slip conditions. We generate the meshes with 108,352 nodes and 215,680 elements, 27,216 nodes and 53,920 elements, 6868 nodes and 13,480 elements, and 1749 nodes and 3370 elements, respectively, for $k = 0$, $k = 1$, $k = 2$, and $k = 3$. We show those grids for $k = 2$ and $k = 3$ in Fig. [2.](#page-4-1)

The inflow boundary condition is modeled by a free stream velocity

$$
u = u_{\infty}\hat{i} \tag{21}
$$

and the outfow by the natural Neumann condition

$$
\mu \nabla \mathbf{u} \cdot \mathbf{n} = 0. \tag{22}
$$

Fig. 2 Computational grid for the electric potential Poisson equation and the pressure Poisson equations **a** after two levels coarsening $(k = 2)$; **b** after three levels coarsening $(k = 3)$. This figure is reproduced from Ref. [[14](#page-13-12)]

The bottom and top of the domain are perfectly slipped. For the magnetic feld, we follow the boundary conditions described in Refs. [[3,](#page-13-8) [18,](#page-13-6) [19](#page-13-1)]. Accordingly, far from the cylinder, the electric feld vanishes.

The Reynolds number is determined as:

$$
\text{Re} = \frac{\rho du_{\infty}}{\mu}.
$$
 (23)

In this study, we set this dimensionless number to $Re = 100$. In the International Unit System, we set the density (ρ), cylinder diameter (*d*), and free stream velocity (u_{∞}) to 1.00; and the viscosity (μ) to 0.01. The length (L) and the height (*H*) of the rectangular box are, respectively, equal to 38 m and 32 m. For the center of the circle, we set $x_c = 8$ m and $y_c = 16$ m.

The Stuart number or the magnetic interaction parameter is expressed as:

$$
N = \frac{\sigma \mathbf{B}_\text{o}^2 d}{\rho u_\infty}.\tag{24}
$$

The constant magnetic field (B_0) is set to 1.00 in the International Unit System. We vary the electric conductivity (σ) of the fluid to set the Stuart number. The numerical computations are performed until time *t* = 150 s with the fixed time increment of $\delta t = 0.05$ s.

We demonstrate the grid resolution of each simulation using the label of *M* ∶ *P*. *M* and *P* illustrate, respectively, the spatial resolution of the advection–difusion and electric potential Poisson solvers. When the CGP scheme is applied to both the electric potential and the pressure Poisson equations, *P* represents the mesh resolution of the pressure Poisson equation as well.

We study the effect of the CGP strategy on the incompressible MHD flow for two different Stuart numbers: $N = 0.01$ and $N = 0.50$. Figure [3](#page-5-0) shows the time evolution of the lift (C_L) and drag (C_D) coefficients for these Stuart numbers when the fow feld is simulated using the standard algorithm $(k = 0)$ with the spatial resolution of 215,680:215,680. The magnitude of both the lift and drag forces increase as the Lorentz force increases. Similar observations have been reported by [[10\]](#page-13-4). It is worth noting that this increment is considerable even when $N = 0.01$ in comparison to $N = 0.00$.

We present the numerical results of the current study in two diferent sections. Section [3.1](#page-5-1) gives the outcomes when the CGP method is only applied to the electric potential Poisson solver. Section [3.2;](#page-9-0) however, provides the results when the CGP algorithm is simultaneously used to execute both the electric potential and pressure Poisson solvers.

3.1 Applying CGP to the electric potential Poisson equation

Figure [4](#page-6-0) provides a visual comparison between the obtained vorticity fields with and without the CGP method for $N = 0.01$ and $N = 0.50$ at time $t = 150$ s. At each row of these fgures, we compare the resulting felds with three different resolutions: full fne scale, CGP scale, and full coarse scale. The CGP method provides more detailed data compared to that modeled with a pure coarse grid resolution for all levels of coarsening.

Fig. 3 A comparison between the time evolution of **a** lift coefficient for Stuart numbers of $N = 0.00$, $N = 0.01$, and $N = 0.50$, and **b** drag coefficient for Stuart numbers of $N = 0.00$, $N = 0.01$, and $N = 0.50$. For all the test cases, the standard algorithm ($k = 0$) is used

Engineering with Computers (2022) 38 (Suppl 2):S1539–S1553 S1545

Fig. 4 Vorticity fields at $t = 150$ s for the Stuart number of $N = 0.01$ and $N = 0.50$ when the CGP solver is only used for the electric potential Poisson equation

To more precisely investigate the performance of the CGP idea, we plot the vorticity distribution along the horizontal centerline, behind the cylinder and in the wake region at time $t = 150$ s in Fig. [5](#page-7-0) for $N = 0.01$ and $N = 0.50$. As can be seen from Fig. [5](#page-7-0), the resulting vorticity field for $N = 0.01$ captured from the full fne scale simulation performed on

the 215,680:215,680 grid resolution and the vorticity distribution for all coarsening levels of the CGP simulations are identical. Note that the outputs of the full coarse mesh include spurious fuctuations at the end of domain. In addition, the felds computed by the CGP and standard highresolution (215,680:215,680) algorithms oscillate with the same phase lag. As depicted in Fig. [5](#page-7-0) by increasing the Lorentz force $(N = 0.50)$, the performance of CGP varies. For one and two levels $(k = 1$ and $k = 2)$ of coarsening when $N = 0.50$ $N = 0.50$ $N = 0.50$ (see Fig. 5c, d), although the phases of periodic variation of the CGP and standard high-resolution (215,680:215,680) schemes are not equal to each other, CGP filters artificial fluctuations contaminating the flow field simulated with the pure coarse resolutions.

An exact measurement of the lift and drag forces are usually needed for engineering designs. To examine the efficiency of the CGP framework from this perspective, we plot the time evolution of the lift and drag coefficients of the flow past a cylinder for Stuart numbers of $N = 0.01$ and $N = 0.50$ for diferent spatial grid resolutions in Fig. [6](#page-8-0). To save space, we simply exhibit six diferent situations as a few examples in Fig. [6](#page-8-0). In all cases, the predictions of CGP are close to the full fne scale results, while they are signifcantly more accurate than the full coarse outputs.

 $(k = 0, 2, \text{ and } 0)$; **b** vorticity for the Stuart number of $N = 0.01$ for CGP ($k = 0$, 3, and 0); **c** vorticity for the Stuart number of $N = 0.50$ for CGP $(k = 0, 1, \text{ and } 0)$; **d** vorticity for the Stuart number of $N = 0.50$ for CGP ($k = 0, 2,$ and 0)

Fig. 5 Comparison of vorticity in the wake of the flow over a cylinder at $t = 150$ s when the CGP solver is only used for the electric potential Poisson equation for diferent values of the electric potential Poisson, the advection–difusion, and the pressure Poisson grid resolutions. **a** Vorticity for the Stuart number of $N = 0.01$ for CGP

Fig. 6 A comparison between the time evolution of lift and drag coeffcients when the CGP solver is only used for the electric potential Poisson equation for diferent values of the electric potential Poisson, the advection–difusion, and the pressure Poisson grid resolutions. **a** Drag coefficient for the Stuart number of $N = 0.01$ for CGP $(k = 0, 1, and 0)$; **b** drag coefficient for the Stuart number

of $N = 0.01$ for CGP ($k = 0, 2,$ and 0); **c** drag coefficient for the Stuart number of $N = 0.01$ for CGP ($k = 0$, 3, and 0); **d** lift coefficient for the Stuart number of $N = 0.01$ for CGP ($k = 0, 3,$ and 0); **e** drag coefficient for the Stuart number of $N = 0.50$ for CGP $(k = 0, 1, and 0)$; **f** drag coefficient for the Stuart number of $N = 0.50$ for CGP ($k = 0, 2,$ and 0)

Table 1 Comparison of total CPU times between the standard and CGP algorithms for two diferent Stuart numbers when the CGP solver is only used for the electric potential Poisson equation

k	Resolution	$N = 0.01$		$N = 0.50$	
		CPU(s)	Speed- up	CPU(s)	Speed- up
Ω	215680: 215680	3,148,020.0	1.000	3,711,894.0	1.000
1	215680 : 53920	957,878.0	3.286	1,272,233.0 2.917	
\mathcal{L}	215680: 13480	886,935.0 3.549		1,214,636.0	3.055
Ω	53920: 53920	252,062.0	1.000	150,768.0	1.000
0	13480 : 13480	25,561.7 1.000		16,867.6 1.000	

The total CPU time consumed by the CGP scheme and the standard algorithm as well as the leading speed-up factors for diferent mesh resolutions and for the Stuart numbers of $N = 0.01$ and $N = 0.50$ are tabulated in Table [1.](#page-9-1) We restrict the data to the case of well-resolved solutions. Hence, the data corresponded to the three levels $(k = 3)$ of grid coarsening is not shown in Table [1](#page-9-1). Accordingly, the maximum achieving speed-up factor is 3.549. In general, the accelerating factors obtained for $N = 0.01$ are higher than for $N = 0.50$ in the equivalent CGP resolution, as by increasing the Lorentz force the resulting system of equations becomes stifer.

Figure [7](#page-9-2) compares the CPU times consumed by various components of the problem for the Stuart number of $N = 0.01$. As can be seen from Fig. [7](#page-9-2), the electric potential Poisson equation is the most time-consuming component of the process when the standard approach $(k = 0)$ is used. Surprisingly, this cost overcomes even the computational price devoted to the pressure Poisson equation. By the coarsening level increment, the electric potential Poisson equation

expense lessens dramatically so that its portion becomes less than 1% after just two levels of coarsening. For three levels of coarsening, the electric potential Poisson solver cost becomes insignifcant, while the pressure Poisson equation consumes the majority of the computational resources. That is our motivation to apply the CGP technique to the pressure Poisson equation to obtain the maximum possible speed-up factor for the simulation of incompressible MHD flows.

3.2 Applying CGP to the electric potential Poisson equation and the pressure Poisson equation

As we discussed in the previous section, the CGP confguration is profcient to reduce the computational cost associated with the electric potential Poisson equation while it preserves the accuracy of the velocity feld in the presence of an external magnetic feld. On the other hand, a considerable number of researchers in the literature [\[11](#page-13-14)–[14,](#page-13-12) [17,](#page-13-10) [23\]](#page-14-4) demonstrated the efficiency of the CGP algorithm when it was applied to the solution of the pressure Poisson equation in pressure correction schemes to numerically solve the incompressible Navier–Stokes equations. The main concern of this section is to assess the capability of CGP in terms of both accelerating the computations and preserving the accuracy of the velocity feld when one applies CGP to both the pressure Poisson and electric potential Poisson equations simultaneously in the numerical simulations of incompressible MHD felds. Similar to the previous section, we consider simulations with Stuart numbers of $N = 0.01$ and $N = 0.50$. But to save space, we mainly present the results of $N = 0.01$.

Figure [8](#page-10-0) exposes the vorticity feld computed by the standard algorithm $(k = 0)$ and the CGP scheme for $k = 1$, 2, and 3 for the Stuart number of $N = 0.01$ at time $t = 150$ s. From a visual point of view, the fidelity of the CGP

Fig. 7 Percent CPU times for the electric potential Poisson part, the advection–difusion part, and the pressure Poisson part for the standard and CGP algorithms for the Stuart number of $N = 0.01$ when the

CGP solver is only used for the electric potential Poisson equation. For all the test cases, the mapping cost is less than 0.001% and is not shown

Fig. 8 Vorticity fields at $t = 150$ s for the Stuart number of $N = 0.01$ when the CGP solver is applied to both the electric potential Poisson equation and the pressure Poisson equation

vorticity felds are preserved for one and two levels of the grid coarsening $(k = 1, \text{ and } 2)$ and they are close to the full fine scale mesh resolution ($k = 0$, 215,680:215,680). More notably, the CGP vorticity fields for $k = 1$ (215,680: 53,920) and $k = 2$ (215,680: 13,480) have a higher level of accuracy compared to the results obtained by the corresponding full coarse resolutions (e.g., 53,920:53,920 and 13,480:13,480). For three coarsening levels $(k = 3)$; however, the fidelity of the vorticity feld computed by a CGP simulation with the 215,680:3370 resolution is only preserved in a reasonable range but yet signifcantly provides more accurate information compared to the simulations performed on the coarse scale resolution of 3370:3370.

As can be seen from Fig. [9,](#page-11-1) we realize that at some cases the CGP methodology leads to a phase lag between the outputs with 215,680:215,680 and 215,680:53,920 or 215,680:13,480 grid resolutions. Our numerical experiments express that these types of phase lags between the outputs of the CGP and standard mechanisms also depend on the time increment (δt) chosen. Similar observations are reported by Kashefi and Staples $[14]$ $[14]$ when they studied the CGP effect on the simulation of the incompressible fow past a cylinder.

Notwithstanding this efect, the key fact is that although dampened fows around the cylinder surfaces and spurious fuctuations at the end of the domain can be observed in the numerical results of the corresponding coarse scale, they are completely removed by the CGP strategy as shown in Fig. [9.](#page-11-1)

Lift and drag are both two critical engineering quantities in the industry. Thus, is it vital to validate the CGP performance for the computations of these two forces. In Sect. [3.1](#page-5-1) of this article, we examined the CGP platform for the total lift and drag calculations when we applied CGP only to the electric potential Poisson equation. Kashef and Staples [[14\]](#page-13-12) applied CGP to the pressure Poisson equation to simulate incompressible fows past a circular cylinder. Through their numerical outcomes, they showed that the pressure drag/lift forces have a lower sensitivity to the grid resolution in comparison with the viscous drag/lift forces (e.g., see Fig. 11 in Ref. [\[14\]](#page-13-12)). The same experience is repeated here, when we applied the CGP methodology to both the pressure Poisson and electric potential Poisson equations. With this in mind, we display the time evolution of the viscous drag and viscous lift for the Stuart number of $N = 0.01$ in Fig. [10](#page-12-0). With reference to Fig. [11](#page-13-20), the viscous

Fig. 9 Comparison of vorticity in the wake of the fow over a cylinder at $t = 150$ s when the CGP solver is applied to both the electric potential Poisson equation and the pressure Poisson equation for diferent values of the electric potential Poisson, the advection–difusion, and

drag and lift coefficients computed by the CGP algorithm are more accurate than those coefficients obtained by standard coarse scale simulations. Specifcally, the CGP performance for the viscous drag coefficient (C_{Df}) calculation is signifcantly higher than the full coarse scale computation as can be seen from Fig. [10.](#page-12-0) For instance, we observe from Fig. [10d](#page-12-0) that although the output of CGP with three coarsening level (*k* = 3, 215,680:3370) has 2.3% error in the viscous drag coefficient with reference to the standard full fne scale computation (215,680:215,680), the error of full coarse scale simulation (3370:3370) is 34.5% in respect to the same reference.

So far, we investigated the accuracy of the velocity and vorticity felds and the viscous drag and lift forces computed by CGP. Another important aspect of the CGP applications is speeding up the computations. Table [2](#page-13-21) and Fig. [11](#page-13-20) allocate the relevant information.

The information collected in Table [2](#page-13-21) demonstrates the CPU time of CGP and non-CGP simulations along with their achieved speed-up. Similar to Table [1,](#page-9-1) we only present the data relevant to the well-resolved solutions in Table [2](#page-13-21). In contrast with the trend tabulated in Table [1,](#page-9-1) by increasing the Lorentz force the speed-up factor increases. In this case, the maximum speed-up is approximately a factor of 23. As expected we experience higher levels of accelerations in comparison with the data tabulated in Table [1](#page-9-1), as CGP is right now applied to both the electric potential Poisson and the pressure Poisson equations. Note that the speed-up factors reported here are not absolute and depend on available computational resources such as CPU models, number of cores, and memory capacities.

the pressure Poisson grid resolutions. **a** Vorticity for the Stuart number of $N = 0.01$ for CGP ($k = 0, 1,$ and 0); **b** vorticity for the Stuart number of $N = 0.01$ for CGP ($k = 0, 2,$ and 0)

As can be observed in Fig. [11,](#page-13-20) the portion of computational expenses associated with both the pressure Poisson and the electric potential Poisson equations becomes less than 7% only by means of two levels of coarsening. According to the information provided in Fig. [11,](#page-13-20) the electric potential Poisson equation is stifer than the pressure Poisson equation even after three levels of coarsening $(k = 3)$.

4 Conclusions

In the present work, the CGP multiresolution algorithm was used for the frst time to obtain a numerical solution to incompressible MHD flows. CGP saved the computational resources mainly due to solving the electric potential Poisson equation on a coarser grid relative to that one used for the simulation of the advection–difusion equation. The maximum speed-up in the case of using CGP for the electric potential Poisson equation was approximately a factor of 3. By utilizing CGP for both the electric potential and pressure Poisson equations, the maximum achieved speedup was approximately a factor of 23. The CGP performance was investigated by the examination of the structure of the von Karman street for the vorticity feld, the velocity feld in the wake region, and the drag and lift coefficient evolution. Our numerical experiments demonstrated that CGP was able to maintain the accuracy of quantities of interest with reference to standard algorithms with high-resolution grids for both the advection–difusion and the electric potential Poisson equation, while reduced computational expenses dramatically. Notably, the CGP results were signifcantly more

Fig. 10 A comparison between the time evolution of viscous lift and viscous drag coefficients for the Stuart number of $N = 0.01$ when the CGP solver is applied to both the electric potential Poisson equation and the pressure Poisson equation for diferent values of the electric potential Poisson, the advection–difusion, and the pressure Poisson grid resolutions. **a** Viscous lift coefficient for CGP $(k = 0, 1, \text{ and } 0)$;

b viscous drag coefficient for CGP $(k = 0, 1, \text{ and } 0)$; **c** viscous lift coefficient for CGP $(k = 0, 2, \text{ and } 0)$; **d** viscous drag coefficient for CGP $(k = 0, 2, \text{ and } 0)$; **e** viscous lift coefficient for CGP $(k = 0, 3, \text{ and } 0)$; **f** viscous drag coefficient for CGP $(k = 0, 3, \text{ and } 0)$

Fig. 11 Percent CPU times for the electric potential Poisson part, the advection–difusion part, and the pressure Poisson part for the standard and CGP algorithms for the Stuart number of $N = 0.01$ when the

Table 2 Comparison of total CPU times between the standard and CGP algorithms for two diferent Stuart numbers when the CGP solver is applied to both the electric potential Poisson equation and the pressure Poisson equation

accurate than standard algorithms with low-resolution grids for both the above-mentioned equations.

Acknowledgements AK wishes to thank Dr. Peter Minev for helpful discussions. Moreover, the author wishes to thank the reviewer for the insightful comments.

References

- 1. Boozer AH (1986) Ohm's law for mean magnetic felds. J Plasma Phys 35:133–139
- 2. Chorin AJ (1968) Numerical solution of the Navier-Stokes equations. Math Comput 22:745–762
- 3. Dousset V, Pothérat A (2008) Numerical simulations of a cylinder wake under a strong axial magnetic feld. Phys Fluids 20:017104
- 4. Farooq U, Lu D, Munir S, Ramzan M, Suleman M, Hussain S (2019) MHD flow of Maxwell fluid with nanomaterials due to an exponentially stretching surface. Sci Rep 9:7312
- 5. Ganta N, Mahato B, Bhumkar YG (2019) Analysis of sound generation by flow past a circular cylinder performing rotary oscillations using direct simulation approach. Phys Fluids 31:026104

CGP solver is applied to both the electric potential Poisson equation and the pressure Poisson equation. For all the test cases, the mapping cost is less than 0.001% and is not shown

- 6. Geuzaine C, Remacle JF (2009) Gmsh: a 3-D fnite element mesh generator with built-in pre-and post-processing facilities. Int J Numer Methods Eng 79:1309–1331
- 7. Ghosh S, Sarkar S, Sivakumar R, Sekhar T (2019) Forced convection magnetohydrodynamic fow past a circular cylinder by considering the penetration of magnetic feld inside it. Numer Heat Transf Part A Appl 1:1–18
- 8. Irfan M, Farooq MA (2019) Magnetohydrodynamic free stream and heat transfer of nanofuid fow over an exponentially radiating stretching sheet with variable fuid properties. Front Phys 7:186
- 9. Jin M, Liu W, Chen Q (2014) Accelerating fast fuid dynamics with a coarse-grid projection scheme. HVAC&R Res 20:932–943
- 10. Josserand J, Marty P, Alemany A (1993) Pressure and drag measurements on a cylinder in a liquid metal fow with an aligned magnetic feld. Fluid Dyn Res 11:107
- 11. Kashef A (2018) A coarse-grid incremental pressure projection method for accelerating low Reynolds number incompressible flow simulations, Iran. J Comput Sci 1:1-11
- 12. Kashef A (2020) A coarse grid projection method for accelerating free and forced convection heat transfer computations. RM 75:33
- 13. Kashef A (2018) Coarse grid projection methodology: a partial mesh refnement tool for incompressible fow simulations. Bull Iran Math Soc 1:1–5
- 14. Kashef A, Staples AE (2018) A fnite-element coarse-grid projection method for incompressible fow simulations. Adv Comput Math 44:1063–1090
- 15. Klein R, Pothérat A (2010) Appearance of three dimensionality in wall-bounded MHD fows. Phys Rev Lett 104:034502
- 16. Lee H, Ha M, Yoon H (2005) A numerical study on the fuid fow and heat transfer in the confned jet fow in the presence of magnetic feld. Int J Heat Mass Transf 48:5297–5309
- 17. Lentine M, Zheng W, Fedkiw R (2010) A novel algorithm for incompressible fow using only a coarse grid projection. ACM Trans Graph (TOG) 10:114
- 18. Mück B, Günther C, Müller U, Bühler L (2000) Three-dimensional MHD flows in rectangular ducts with internal obstacles. J Fluid Mech 418:265–295
- 19. Mutschke G, Gerbeth G, Shatrov V, Tomboulides A (1997) Twoand three-dimensional instabilities of the cylinder wake in an aligned magnetic feld. Phys Fluids 9:3114–3116
- 20. Peng Y, Alsagri AS, Afrand M, Moradi R (2019) A numerical simulation for magnetohydrodynamic nanofuid fow and heat

transfer in rotating horizontal annulus with thermal radiation. RSC Adv 9:22185–22197

- 21. Reddy JN (1993) An introduction to the fnite element method. McGraw-Hill, New York
- 22. Saad Y, Schultz MH (1986) GMRES: a generalized minimal residual algorithm for solving nonsymmetric linear systems. SIAM J Sci Stat Comput 7:856–869
- 23. San O, Staples AE (2013) A coarse-grid projection method for accelerating incompressible fow computations. J Comput Phys 233:480–508
- 24. San O, Staples AE (2013) An efficient coarse grid projection method for quasigeostrophic models of large-scale ocean circulation. Int J Multiscale Comput Eng 11:1
- 25. Shah Z, Babazadeh H, Kumam P, Shafee A, Thounthong P (2019) Numerical simulation of magnetohydrodynamic nanofuids under the infuence of shape factor and thermal transport in a porous media using CVFEM. Front Phys 7:164. [https://doi.org/10.3389/](https://doi.org/10.3389/fphy) [fphy](https://doi.org/10.3389/fphy)
- 26. Shoaib M, Akhtar R, Khan MAR, Rana MA, Siddiqui AM, Zhiyu Z, Raja MAZ (2019) A novel design of three-dimensional MHD fow of second-grade fuid past a porous plate. Math Probl Eng 2019:1
- 27. Verron J, Sommeria J (1987) Numerical simulation of a twodimensional turbulence experiment in magnetohydrodynamics. Phys Fluids 30:732–739
- 28. Yoon H, Chun H, Ha M, Lee H (2004) A numerical study on the fuid fow and heat transfer around a circular cylinder in an aligned magnetic feld. Int J Heat Mass Transf 47:4075–4087
- 29. Zhao M, Lu L (2018) Numerical simulation of fow past two circular cylinders in cruciform arrangement. J Fluid Mech 848:1013–1039

Publisher's Note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.