



A fast computation technique for the direct numerical simulation of rigid particulate flows

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Abstract

In this paper, we present a computation technique for the direct numerical simulation of freely moving rigid bodies in fluids. We solve three-dimensional laminar flow problems using a control volume approach. The key feature of this approach is that the computational overhead (relative to a pure fluid solver) to solve for the motion of rigid particle is very small. The formulation is convenient for handling irregular geometries. We present results for the sedimentation of particles of different shapes. Convergence tests are presented to assess the order of accuracy of the numerical scheme. Various test cases are considered and the numerical results are compared with experimental values to validate the code. Due to the ability to perform fast computations, this method has been used for animations and its application to the direct numerical simulation of turbulent particulate flows merits investigation. The technique is not restricted to any constitutive model of the suspending fluid. Hence, it may potentially be used in Large Eddy Simulations (LES) or Reynolds Averaged Navier–Stokes (RANS) type simulations.

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1. Introduction

Numerical simulation techniques for solid–liquid flows, which do not use any model for fluid–particle interaction, have been developed over the past 12 years. In these methods, the fluid flow is governed by the continuity and momentum equations, whereas the particles are governed by the equation of motion for rigid body. The flow field around each individual particle is resolved. The hydrodynamic force between the particles and the fluid is obtained from the solution and is not modeled by any drag law. These

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simulations referred to as direct numerical simulation (DNS) of solid–liquid flows, can be applied in numerous settings; e.g., sedimenting and fluidized suspensions, lubricated transport, hydraulic fracturing of reservoirs, slurries, understanding particle–turbulence interaction, etc.

Hu and co-workers [11–13] developed a finite-element method based on unstructured grids to simulate the motion of large numbers of rigid particles in two and three dimensions in Newtonian and viscoelastic fluids. This approach is based on an arbitrary-Lagrangian–Eulerian (ALE) technique that uses a moving mesh scheme to handle the time-dependent fluid domain. A new mesh is generated when the old one becomes too distorted, and the flow field is projected on to the new mesh. A combined fluid–particle weak formulation is used, where the hydrodynamic forces and torques on the particle are not computed explicitly. Another numerical scheme based on the moving mesh technique was developed by Johnson and Tezduyar [14]. They used a space time finite element formulation and a fully explicit scheme in which the forces and torques on the particles were calculated explicitly to solve the equations of rigid motion.

Glowinski et al. [9] presented a distributed Lagrange-multiplier/fictitious-domain (DLM) method for the direct numerical simulation of the motion of many rigid particles in Newtonian fluids. Their finite element formulation permits the use of a fixed structured grid. This eliminates the need for remeshing the domain – a necessity in the unstructured grid-based methods. Structured grids also allow the use of fast and efficient solvers. In the DLM method, the entire fluid–particle domain is assumed to be a fluid and the flow in the particle domain is constrained to be a rigid-body motion by using a rigidity constraint. This leads to a Lagrange multiplier field in the particle domain. The constraint of rigid body motion is represented by $\mathbf{u} = \mathbf{U} + \boldsymbol{\omega} \times \mathbf{r}$, where \mathbf{u} is the velocity of the fluid at a point inside the particle domain, \mathbf{U} and $\boldsymbol{\omega}$ are the translational and angular velocities of the particle, respectively, and \mathbf{r} is the position vector of the point with respect to the particle centroid. The fluid–particle motion is treated implicitly using a combined weak formulation in which the mutual forces cancel.

A new DLM formulation for particulate flows was later presented by Patankar et al. [21]. In their approach, the rigid motion is imposed by constraining the deformation-rate tensor within the particle domain to be zero. This eliminates \mathbf{U} and $\boldsymbol{\omega}$ as variables from the coupled system of equations – thus enabling simulations of irregular shaped bodies without additional complexity. This formulation recognizes that the rigidity constraint results in a stress field inside a rigid solid just as there is pressure in an incompressible fluid. The DLM formulation of Glowinski et al. [9] and Patankar et al. [21] were implemented by using a Marchuk–Yanenko fractional-step scheme for time discretization. A finite-element method was used.

The ALE and DLM formulations, discussed above, have been implemented for laminar flow conditions. Although the DLM approach has been successfully used for computations with up to 1204 spheres in three dimensions – Pan et al. [19] – further speed-up of the solution procedure is desirable. This is critical for DNS of turbulent particulate flows or for performing simulations with thousands of particles where fast computations are essential for the problems to be computationally feasible. With these objectives in consideration, Patankar [20] presented an adapted version of his earlier DLM-based formulation – Patankar et al. [21] – for the fast computation of rigid particulate flows. The key issue addressed in the formulation was the fast implementation of the rigidity constraint, irrespective of the type of equations used to describe the fluid. The formulation eliminated the need for an iterative procedure to solve the rigid body projection step thereby allowing faster computations for particulate flows.

In this paper, we have implemented the formulation developed by Patankar [20] for the case of three-dimensional particles using a control volume approach. Our objective is to demonstrate the effectiveness of the approach in handling three-dimensional regular and irregular shaped particles without significant computational overhead. We report convergence tests to check the order of accuracy of the scheme. A fixed structured grid is used in both the fluid and the particle domains. The mathematical formulation for the problem is laid down in Section 2. The computational scheme is outlined in Section 3. The numerical implementation is outlined in Section 4. Section 5 describes the results obtained and the validation.

2. Mathematical formulation

Let Ω be the computational domain which includes both the fluid and the particle domain. Let $P(t)$ be the particle domain. Let the fluid boundary not shared with the particle be denoted by Γ . We will consider a Dirichlet boundary condition on Γ . At times we will assume a single particle in the domain for the simplicity of exposition. The body force will also be assumed to be constant so that there is no net torque acting on the particle. The governing equations for fluid motion are given by:

$$\rho_f \left(\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} \right) = \nabla \cdot \boldsymbol{\sigma} + \rho_f \mathbf{g} \quad \text{in } \Omega \setminus \overline{P(t)}, \tag{1}$$

$$\nabla \cdot \mathbf{u} = 0 \quad \text{in } \Omega \setminus \overline{P(t)}, \tag{2}$$

$$\mathbf{u} = \mathbf{u}_\Gamma(t) \quad \text{on } \Gamma, \tag{3}$$

$$\mathbf{u} = \mathbf{u}_i \quad \text{on } \partial P(t), \tag{4a}$$

$$\boldsymbol{\sigma} \cdot \mathbf{n} = \mathbf{t} \quad \text{on } \partial P(t), \tag{4b}$$

$$\mathbf{u}|_{t=0} = \mathbf{u}_0(\mathbf{x}) \quad \text{in } \Omega \setminus \overline{P(0)}, \tag{5}$$

where ρ_f is the fluid density, \mathbf{u} is the fluid velocity, \mathbf{g} is the acceleration due to gravity, \mathbf{n} is the outward normal on the particle surface, \mathbf{u}_i is the velocity at fluid–particle interface $\partial P(t)$, $\boldsymbol{\sigma}$ is the stress tensor and \mathbf{t} is the traction vector on the particle surface. The initial velocity \mathbf{u}_0 should satisfy (2). The boundary velocity, \mathbf{u}_Γ , in (3) should satisfy the compatibility condition due to (2). For an incompressible fluid the divergence-free constraint (2) gives rise to pressure in the fluid. The stress tensor is then given by:

$$\boldsymbol{\sigma} = -p\mathbf{I} + \boldsymbol{\tau}, \tag{6}$$

where \mathbf{I} is the identity tensor, p is the pressure and $\boldsymbol{\tau}$ is the extra-stress tensor given by,

$$\boldsymbol{\tau} = \mu \left[\nabla \mathbf{u} + (\nabla \mathbf{u})^T \right], \tag{7}$$

for a Newtonian fluid, where μ is the viscosity of the fluid.

Particle motion can be represented in terms of translational and angular velocities using Newton’s second law. In the DLM formulation the particle is treated as a fluid that is subjected to an additional rigidity constraint. The governing equations for the particle motion are given by [21]:

$$\rho_s \left(\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} \right) = \nabla \cdot \boldsymbol{\sigma} + \rho_s \mathbf{g} \quad \text{in } P(t), \tag{8}$$

$$\nabla \cdot \mathbf{u} = 0 \quad \text{in } P(t), \tag{9}$$

$$\mathbf{D}[\mathbf{u}] = \frac{1}{2}(\nabla \mathbf{u} + \nabla \mathbf{u}^T) = \mathbf{0} \quad \text{in } P(t), \tag{10}$$

$$\mathbf{u} = \mathbf{u}_i \quad \text{on } \partial P(t), \tag{11a}$$

$$\boldsymbol{\sigma} \cdot \mathbf{n} = \mathbf{t} \quad \text{on } \partial P(t), \tag{11b}$$

$$\mathbf{u}|_{t=0} = \mathbf{u}_0(\mathbf{x}) \quad \text{in } P(0), \tag{12}$$

where ρ_s is the particle density. The initial velocity \mathbf{u}_0 should satisfy (10). The rigidity constraint (10) ensures that the velocity field is divergence-free. Hence (9) is a redundant equation but is retained in order to facilitate numerical implementation [21]. As noted earlier the incompressibility constraint (Eq. (9)), which is a scalar constraint; gives rise to a scalar field (pressure p). Similarly the rigidity constraint (Eq. (10)), which is a tensor constraint; gives rise to a tensor field (stress \mathbf{L}) in the particle domain. \mathbf{L} is a symmetric second-order tensor. Variational analysis of the above equations shows that the pressure p and stress \mathbf{L} are the distributed Lagrange multipliers due to the divergence-free and rigidity constraints, respectively [21]. \mathbf{L} is an additional stress field required inside the particle domain to maintain the rigid-body motion. The Lagrange multiplier stress \mathbf{L} has six scalar variables in a three-dimensional case. This is because Eq. (10) represents six scalar constraint equations at a point. A reformulation of Eq. (10) can reduce the number of Lagrange multiplier variables to three. The rigidity constraint can be implemented by imposing [21]:

$$\nabla \cdot (\mathbf{D}[\mathbf{u}]) = \mathbf{0} \quad \text{in } P(t), \quad (13)$$

$$\mathbf{D}[\mathbf{u}] \cdot \mathbf{n} = \mathbf{0} \quad \text{on } \partial P(t), \quad (14)$$

where (13) and (14) represent three scalar constraint equations at a point. As a result \mathbf{L} is no longer the Lagrange multiplier itself. It remains the stress field inside the particle due to the rigidity constraint. It can be represented in terms of a Lagrange multiplier λ by [21]:

$$\mathbf{L} = \mathbf{D}[\lambda], \quad (15)$$

where λ is a vector with three scalar components in a three-dimensional case. Hence, the stress inside the particle is given by [21]:

$$\boldsymbol{\sigma} = -p\mathbf{I} + \mathbf{D}[\lambda] + \boldsymbol{\tau}. \quad (16)$$

$\boldsymbol{\tau}$ is zero in the particle domain due to the rigidity constraint (Eqs. (13) and (14)). Nevertheless, the term is retained in the particle domain to facilitate the solution procedure [21].

Patankar et al. [21] solved the above fluid–particle equations using a finite element method. Patankar [20] presented an adapted version of this formulation thus allowing for fast computation of rigid particulate flows.

3. Computational algorithm

We present the computational algorithm based on the formulation by Patankar [20]. Assume a Newtonian fluid with constant viscosity μ . The momentum equation applicable in the entire domain can be written as:

$$(\rho_f(1-H) + H\rho_s) \left(\frac{\partial \mathbf{u}}{\partial t} + (\mathbf{u} \cdot \nabla) \mathbf{u} \right) = -\nabla p + \mu \nabla^2 \mathbf{u} + (\rho_f(1-H) + H\rho_s) \mathbf{g} + \mathbf{f} \quad \text{in } \Omega, \quad (17)$$

where \mathbf{f} is the additional term due to the rigidity constraint in Eqs. (13) and (14) in the particle domain. Note that $\mathbf{f} = \nabla \cdot \mathbf{D}[\lambda]$ according to Eqs. (8) and (16). The continuity equation applicable in the entire domain is given by:

$$\frac{\partial (\rho_f(1-H) + H\rho_s)}{\partial t} + \nabla \cdot [(\rho_f(1-H) + H\rho_s) \mathbf{u}] = 0 \quad \text{in } \Omega. \quad (18)$$

In Eqs. (17) and (18), H is a step function given by:

$$H = \begin{cases} 0 & \text{in } \Omega/\overline{P(t)}, \\ 1 & \text{in } P(t). \end{cases} \tag{19}$$

H moves with the particle, therefore,

$$\frac{DH}{Dt} = 0 \quad \text{on } \partial P(t), \tag{20}$$

where D/Dt is the material derivative. The continuity Eq. (18) combined with the definition of H (Eqs. (19) and (20)) gives a divergence free velocity field in the entire domain. Also Eq. (17) reduces to Eq. (1) in the fluid domain and Eq. (8) in the particle domain.

The governing equations can be solved by the following two fractional time steps [20]:

- (1) Solve for $\hat{\mathbf{u}}$ and \hat{p} satisfying the following equations in the entire domain:

$$\left. \begin{aligned} (\rho_f(1 - H) + H\rho_s)\left(\frac{\hat{\mathbf{u}} - \mathbf{u}^n}{\Delta t} + (\hat{\mathbf{u}} \cdot \nabla)\hat{\mathbf{u}}\right) &= -\nabla\hat{p} + \mu\nabla^2\hat{\mathbf{u}} + (\rho_f(1 - H) + H\rho_s)\mathbf{g}, \\ \nabla \cdot \hat{\mathbf{u}} &= 0. \end{aligned} \right\} \tag{21}$$

Set $\mathbf{u}^{n+1} = \hat{\mathbf{u}}$ and $p^{n+1} = \hat{p}$ in the fluid domain. Note that $\mathbf{f} = \mathbf{0}$ everywhere in this step.

- (2) Solve for \mathbf{u}^{n+1} in the particle domain by projecting $\hat{\mathbf{u}}$ on to a rigid body motion. To do so, in the particle domain, we have

$$\rho_s \left(\frac{\mathbf{u}^{n+1} - \hat{\mathbf{u}}}{\Delta t} \right) = \mathbf{f}. \tag{22}$$

To solve for \mathbf{u}^{n+1} in step 2 one requires knowledge of \mathbf{f} . An equation for \mathbf{f} can be obtained by using Eqs. (13) and (14) [20]:

$$\left. \begin{aligned} \nabla \cdot (\mathbf{D}[\mathbf{u}^{n+1}]) &= \nabla \cdot \left(\mathbf{D} \left[\hat{\mathbf{u}} + \frac{\mathbf{f}\Delta t}{\rho_s} \right] \right) = \mathbf{0}, \\ \mathbf{D}[\mathbf{u}^{n+1}] \cdot \mathbf{n} &= \mathbf{D} \left[\hat{\mathbf{u}} + \frac{\mathbf{f}\Delta t}{\rho_s} \right] \cdot \mathbf{n} = \mathbf{0}. \end{aligned} \right\} \tag{23}$$

The above equation implies that $\hat{\mathbf{u}} + (\mathbf{f}\Delta t/\rho_s)$ is a rigid body motion, but it gives no information about what this motion should be. In fact, this rigid motion is the solution that we are looking for, because $\hat{\mathbf{u}} + (\mathbf{f}\Delta t/\rho_s) = \mathbf{u}^{n+1}$ from Eq. (22).

Patankar [20] argued the rigid body motion can be obtained by imposing an additional condition that, in the projection step, the total linear and angular momenta in the individual particle domain should be conserved. The following procedure can be used to get the required solution.

- (i) Split $\hat{\mathbf{u}}$ as: $\hat{\mathbf{u}} = \mathbf{u}_r + \mathbf{u}'$, where \mathbf{u}_r

$$\left. \begin{aligned} \mathbf{u}_r &= \mathbf{U} + \boldsymbol{\omega} \times \mathbf{r} \\ M\mathbf{U} &= \int_p \rho_s \hat{\mathbf{u}} \, d\mathbf{x} \quad \text{and} \quad \mathbf{I}_p \boldsymbol{\omega} = \int_p \mathbf{r} \times \rho_s \hat{\mathbf{u}} \, d\mathbf{x} \end{aligned} \right\}, \tag{24}$$

where M is the mass of the particle and \mathbf{I}_p is the moment of inertia of the particle. This computational step is cheap since it is merely an addition (integration).

- (ii) Since the linear and the angular momenta should be conserved in the projection step, set $\mathbf{u}^{n+1} = \mathbf{u}_r$ in the particle domain. This is the required solution. Note that $\mathbf{f} = -(\rho_s \mathbf{u}')/\Delta t$. The velocity is corrected in the particle domain (and the smeared interface). The influence of this correction is not propagated by diffusion into the fluid in this fractional step. First, we note that there is no sharp

discontinuity of velocity in the domain; rather the velocity correction is smeared on the length scale of the grid size. Second, the influence of the velocity correction in the particle domain does propagate by diffusion into the fluid during the computation of the first fractional step of the next time step. Third, the combination of the two fractional steps, i.e., the net scheme, is first-order [9,10] also to be discussed below. Smaller time steps are preferable – as is usually recommended for first-order schemes.

The key advantage of the above formulation is that the projection step only involves integrations in the particle domain with no iterations; hence, it is computationally cheap.

Some remarks on the uniqueness and similarity of this approach compared to the previous DLM formulations and implementations are in order. In the above approach, \mathbf{U} and $\boldsymbol{\omega}$ are not implicit variables in the coupled system of equations. This is unlike the formulation by Glowinski et al. [9] where additional equations of motion for \mathbf{U} and $\boldsymbol{\omega}$ are solved. For bodies with irregular shapes, the angular velocity equation is non-linear. The method in this paper, where additional angular velocity equations are not required, can be effective to simulate the motion of irregular shaped bodies without additional complexity.

The operator splitting scheme used above is like the Marchuk–Yanenko scheme used in the previous DLM implementations [9,10,21]. Here we use the control volume discretization while previous implementations were finite element based. Eq. (24) is useful to extend the DLM approach to a control volume discretization. Incorporation of a rigid particle solver in the conventional control volume based conservative schemes for the solution of Navier–Stokes equations, therefore, becomes straightforward.

Glowinski and co-workers [9,10] discussed different operator splitting schemes of the Marchuk–Yanenko type, primarily depending on how the viscous term is split. One version they considered is equivalent to keeping part of the viscous term in fractional step one (a modified Eq. (21)) and the other part in fractional step two (a modified Eq. (22)). In this version the rigid motion projection is done coupled with the diffusion equation in the entire domain. The computational cost of this approach, in our earlier implementations, was consequently larger; typically taking up to 30% of the total computational time.

The other version, used widely by e.g. [9,10,21], is equivalent to keeping the entire viscous term in the first fractional step. The rigid motion projection step is then similar to Eq. (22). As such, like the present scheme, the rigid motion projection gives velocity correction in the particle domain and near the smeared interface only. Thus, this aspect is also applicable to the earlier DLM implementations. As discussed above the effect of this correction diffuses into the fluid in the next fractional step. That the net scheme is first-order has been discussed by [9,10]. The same conclusions will apply to the formulation in this work. The numerical results, to be presented below, confirm that our scheme is first-order.

In earlier work, the rigid body projection step similar to Eq. (22) was implemented by a variational formulation (instead of Eq. (24)) that typically required 70–100 iterations (e.g. [9,10,21]). Our previous implementations of this approach usually took 5–10% computational time. In this work, the rigid motion projection is based on Eq. (24) above, where no iterations are necessary. The typical time spent in our implementations using Eq. (24) is less than 1% for all the cases presented in this paper thus speeding up the particle solver.

Dong et al. [5] compared the performance of their spectral DLM approach with the force-coupling method (FCM) by Maxey and Patel [17]. The FCM method is based on representing the particles by low-order force multipoles distributed over a finite volume. Dong et al. [5] reported that FCM is computationally cheaper than their DLM approach. The computational cost of their DLM approach was more in part because of the iterations involved in the rigid motion projection that was implicitly coupled with the diffusion term in the entire domain (they had the diffusion term in the particle projection step). In comparison, the cost of the fractional time stepping scheme such as that in this and in earlier works [9,10,21] is less. This cost reduction is similar to that obtained by using e.g. a Chorin type pressure-based fractional time stepping scheme instead of a fully implicit scheme for the solution of Navier–Stokes equations.

The method to impose the rigidity constraint presented here (without additional equations of motion for \mathbf{U} and $\boldsymbol{\omega}$), lays a theoretical framework that is similar to the first-order Chorin [3] type schemes for the incompressibility constraint. Even with a simple projection method represented by Eqs. (22)–(24), a second-order scheme for the rigidity constraint can be envisaged analogous to the second-order pressure correction based fractional step schemes [16] for the incompressibility constraint. This is relegated to future work.

The formulation above follows that presented by Patankar [20]. Independently, Glowinski and co-workers [10,8] have presented and discussed a rigid motion projection similar to Eq. (24) but including coupled variables \mathbf{U} and $\boldsymbol{\omega}$ and a density difference dependent term. They note that it works well for neutrally buoyant particles (i.e., fluid and particle densities are equal); otherwise the variational formulation based projection (instead of a scheme like in Eq. (24)) works better. There are at least two issues to consider here:

1. The variational formulation, although slower in terms of computational speed, is an error minimizing projection and thus may give better results, in general, than using Eq. (24). This issue can be minimized by improving the order of accuracy of the current time stepping scheme by drawing analogy with the Chorin type schemes. This is a topic of our continuing work.
2. We solve Eq. (21) by accounting for the density variation. That leads to a projection step (Eq. (22)) that does not involve a term dependent on the density difference. Unlike our present scheme, a formulation with a density difference dependent term in the rigid projection equation is discussed by Patankar [20], which has similarities to a scheme independently proposed by Glowinski et al. [10]. The density difference dependent term, which includes inertia and gravity effect, is an additional source term in the rigid projection step (see Patankar [20]). Thus, there is an additional velocity correction in the particle domain. The associated error is minimized in a variational formulation based method. This may improve its performance relative to Eq. (24). For neutrally buoyant particles there is no density difference dependent term in rigid projection, thus the advantage of the variational formulation based projection is reduced. This is consistent with the observation of Glowinski et al. [10]. In our case, even if the fluid and particle densities are not the same, this is not an issue since we do not have a density difference dependent term in Eq. (22).

The numerical implementation will be presented in the following section.

4. Numerical implementation

The various steps involved in the numerical implementation of the algorithm above are described below. We use a control volume method based on staggered grids [22]. Main control volumes (Fig. 1(a)) are used for defining pressure and staggered control volumes are used for defining x (Fig. 1(b)), y (Fig. 1(c)) and z -direction velocities, respectively. The arrows in Fig. 1 indicate the locations where the respective velocity components are defined.

4.1. Computation of volume fraction ϕ

The first step is to calculate the volume fraction ϕ occupied by the solid particle in each of the main control volumes. This is accomplished by the following steps:

- The largest cuboid that can enclose the solid particle is identified. The main control volumes forming this cuboid are marked.

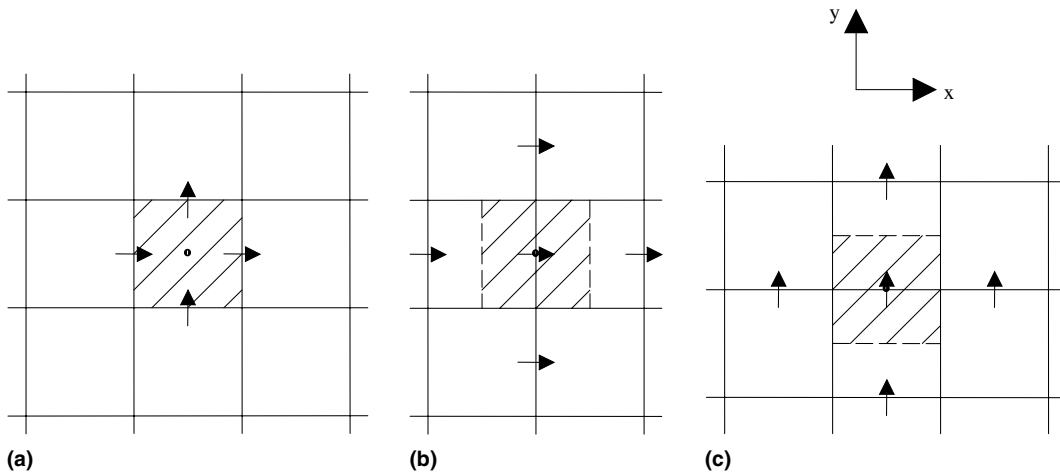


Fig. 1. The marked regions are: (a) main control volume (b) staggered control volume for u (x -direction) velocities and (c) staggered control volume for v (y -direction) velocities. The arrows indicate the locations where the respective velocity components are defined.

- Each of the marked main control volumes is subdivided into N number of finer volumes. The centre coordinates of each of these finer volumes in a given main control volume are checked to see if they lie within the boundaries of the solid particle. To illustrate this step further consider a disc; with radius r and thickness h as shown in Fig. 2. Let \mathbf{r}_c be the position vector of the centre of the disc (x_c, y_c, z_c) , \mathbf{r}_{fv} be the position vector of the centre of the finer volume (x_{fv}, y_{fv}, z_{fv}) and \mathbf{n} be the unit vector along the central axis of the disc.

To mark the finer volume as either being within the solid boundary or not we check for the following conditions.

- (i) We define length, L_p as

$$L_p = |(\mathbf{r}_{fv} - \mathbf{r}_c) \times \mathbf{n}|, \tag{25}$$

where $||$ denotes the magnitude of the vector. L_p must satisfy the following condition if \mathbf{r}_{fv} is inside the disc

$$L_p \leq r. \tag{26}$$

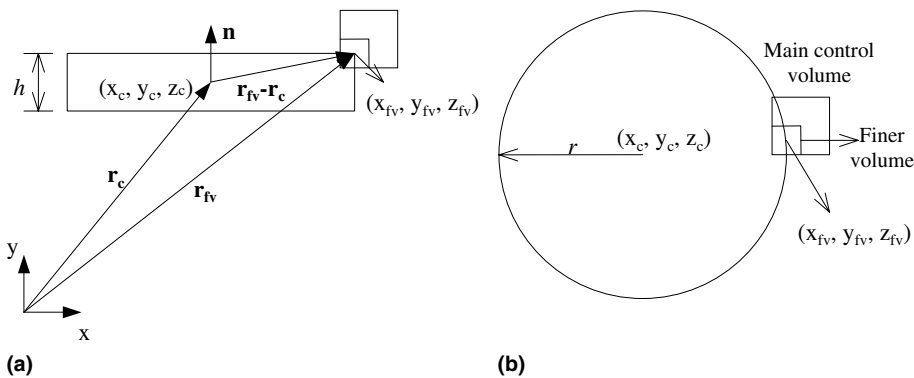


Fig. 2. (a) Side view and (b) top view of a disc along with typical main and fine control volumes.

(ii) (x_{fv}, y_{fv}, z_{fv}) should also lie within the planes bounding the disc,

$$\text{abs}[(\mathbf{r}_{fv} - \mathbf{r}_c) \cdot \mathbf{n}] \leq \frac{h}{2}. \quad (27)$$

If Eqs. (26) and (27) are satisfied; we mark the point to belong inside the disc. This scheme was used, e.g., in the calculations of the sedimentation of a disc, presented in Section 5. Similarly, other shapes were also considered.

- Out of the N fine control volumes in a given main control volume, let N_s have their centers inside the particle domain. The volume fraction φ for this main control volume is given by

$$\varphi = \frac{N_s}{N}. \quad (28)$$

In the above scheme, φ is calculated by a discretization of the main control volume. The accuracy of the approach is good when the fine control volume has an edge length that is small compared to the particle size. We have used the fine control volume edge length that is 1/10 of the edge length of the main control volume. The error relative to the exact value is less than 0.1%.

4.2. Solution of the fluid equations

As outlined in the section on computational algorithm, in the first fractional step, the fluid equations are solved in the entire domain. The fluid equation (21) are discretized using the control volume approach:

$$\int_V \left\{ \frac{\rho^{n+1} \hat{\mathbf{u}} - \rho^n \mathbf{u}^n}{\Delta t} + \nabla \cdot (\rho^{n+1} \hat{\mathbf{u}} \hat{\mathbf{u}}) - \mu \nabla^2 \hat{\mathbf{u}} + \nabla p^{n+1} - \rho^{n+1} \mathbf{g} \right\} dV = \mathbf{0}, \quad (29)$$

$$\int_V \left\{ \frac{\rho^{n+1} - \rho^n}{\Delta t} + \nabla \cdot (\rho^{n+1} \hat{\mathbf{u}}) \right\} dV = 0, \quad (30)$$

where V represents any control volume over which the integration is performed, and

$$\rho = \rho_f(1 - \varphi) + \rho_s \varphi. \quad (31)$$

$\hat{\mathbf{u}}$ is projected, in the next step, on to a rigid body motion in the particle domain. The volume fraction φ takes values between 0 and 1 for control volumes partially occupied by the solid particle, 1 for control volumes fully occupied by the solid particle and zero otherwise. The volume fraction φ is the same as function H defined earlier with the difference that it is smeared on the edge of the particle due to discretization.

The continuity Eq. (30) appears to be for a compressible fluid. This is because of smearing of φ at the particle boundary. Beyond the particle boundary region, the velocity field is indeed divergence free.

The SIMPLER (semi-implicit method for pressure linked equations – revised) algorithm by Patankar [22] is used for the solution of the discretized Eqs. (29) and (30). A fully implicit scheme is used for time discretization and a power law scheme is used for discretizing the flux term (convection-diffusion fluxes). A simple block correction based multigrid solver proposed by Sathyamurthy and Patankar [23] is used to solve the resulting linear equations.

4.3. Rigid body projection of the velocity field in the particle domain

The next step is to project the velocity field $\hat{\mathbf{u}}$ in the particle domain onto a rigid body motion \mathbf{u}_r . This is accomplished by numerically solving for the translational velocity \mathbf{U} and angular velocity $\boldsymbol{\omega}$ of the solid particle by using conservation of the angular and linear momentum given by Eqs. (32) and (33):

$$M\mathbf{U} = \int_P \varphi \rho_s \mathbf{u} \, d\mathbf{x}, \quad (32)$$

$$\mathbf{I}_p \boldsymbol{\omega} = \int_P \varphi \mathbf{r} \times \rho_s \mathbf{u} \, d\mathbf{x}. \quad (33)$$

For the case of a sphere, the moment of inertia \mathbf{I}_p , is a diagonal matrix with all the diagonal entries equal irrespective of the orientation of the main co-ordinate axes. For other geometries like a disc or a box it is not a diagonal matrix unless the principal axes of the object coincide with the main co-ordinate system. \mathbf{I}_p can be evaluated by computing the following integral over the particle domain.

$$\mathbf{I}_p = \int_P \rho_s [(\mathbf{r} \cdot \mathbf{r})\mathbf{I} - \mathbf{r} \otimes \mathbf{r}] \, d\mathbf{x}. \quad (34)$$

The velocity field \mathbf{u}_r in the particle domain is given by,

$$\mathbf{u}_r = \mathbf{U} + \boldsymbol{\omega} \times \mathbf{r}. \quad (35)$$

Linear interpolation of velocity is done to numerically calculate U and $\boldsymbol{\omega}$ according to Eqs. (32) and (33).

4.4. Explicit update of the particle position (translation) and orientation

The final step involves updating the positions of the particles with knowledge of their translational and angular velocities. Positions of the unit vectors along the three principal axes of a particle are updated by using the following expression:

$$\mathbf{e}_i^{n+1} = \cos(|\boldsymbol{\omega}|\Delta t)\mathbf{e}_i^n + (\mathbf{e}_i^n \cdot \hat{\boldsymbol{\omega}})(1 - \cos(|\boldsymbol{\omega}|\Delta t))\hat{\boldsymbol{\omega}} + \sin(|\boldsymbol{\omega}|\Delta t)\hat{\boldsymbol{\omega}} \times \mathbf{e}_i^n, \quad (36)$$

where \mathbf{e}_i^{n+1} is the updated vector equation of unit vector i and i ranges from 1 to 3; $\hat{\boldsymbol{\omega}}$ is a unit vector along the direction of the angular velocity $\boldsymbol{\omega}$. Eq. (36) represents updating the orientation of the particle.

The translational position \mathbf{X} , of a particle is updated by using the following explicit scheme:

Set $\mathbf{X}^{n+1,0} = \mathbf{X}^n$

Do $k = 1, K$

$$\mathbf{X}^{*n+1,k} = \mathbf{X}^{n+1,k-1} + \left(\frac{\mathbf{U}^n + \mathbf{U}^{n-1}}{2} \right) \left(\frac{\Delta t}{K} \right) \quad (37)$$

$$\mathbf{X}^{n+1,k} = \mathbf{X}^{*n+1,k} + M^{-1} \left(\frac{\mathbf{F}(\mathbf{X}^{n+1,k-1}) + \mathbf{F}(\mathbf{X}^{*n+1,k})}{2} \right) \frac{(\Delta t)^2}{2K^2} \quad (38)$$

enddo

Set $\mathbf{X}^{n+1} = \mathbf{X}^{n+1,K}$, where \mathbf{F} denotes the collision force acting on the particles to prevent them from penetrating each other or the walls of the domain. Further details on the modeling of this force can be found in Glowinski et al. [9].

5. Results and discussion

We determine the order of accuracy of the scheme and also validate the code. Convergence results for the fluid–particle solver are presented first. Drag on a sphere as a function of the Reynolds number is compared with experimental values. In the end, results for two spheres, many spheres and different geometries are presented.

5.1. Convergence results

In this sub-section we discuss results of the convergence test for the fluid and the fluid–particle solvers. Convergence test for the fluid solver only (i.e., without the particle projection step) was done to verify that, as expected, it is second-order with respect to the grid size for the problem with uniform fluid properties. We considered a Poiseuille flow inside a channel of square cross-section. The fluid solver was found to be second-order after plotting the error as a function of the grid size on a log-log plot.

Next we test the fluid–particle solver. We consider a single solid sphere of density 3 kg/m^3 and diameter 0.625 m falling under gravity in a rectangular channel of cross-section $2 \text{ m} \times 2 \text{ m}$ and height 8 m . The fluid viscosity is 0.05 kg/s and the density is 2 kg/m^3 . Simulations were carried out for increasing grid refinement. A blocking ratio of 0.3125 was considered. The blocking ratio α is defined as the ratio of sphere diameter d to edge length L of the cross-section of the channel

$$\alpha = d/L. \tag{39}$$

The terminal velocity of the sphere averaged over several time steps was compared with the results obtained experimentally by McNown et al. [18]. Fig. 3 shows a plot of the terminal velocity obtained from numerical simulations as a function of the grid size. In the limit of the grid size going to zero the value obtained (y intercept value) is in reasonably good agreement with the experimental values (approximately 6% difference). Ideally this difference should be zero if the experimental error is zero. It should be noted that the experimental value was read from a graph. The errors associated with the data, which were not given by McNown et al. [18], could account for the 6% difference in the zero grid size limit.

Brown and Lawler [1] systematically analyzed the sphere drag data available to date. They proposed correlations for the drag coefficient and the terminal velocity of a sphere falling in an infinite medium by systematically eliminating the wall effects in the experimental data. The wall effect is best represented in the graph by Fidleris and Whitmore [6]. Using the correlations by Brown and Lawler [1] and the wall effect given by Fidleris and Whitmore [6] we calculated the terminal velocity of the sphere, for the case considered here, to be 1.2722 m/s compared to the zero grid size limit value of 1.2453 m/s in Fig. 3. The discrepancy according to this comparison is 2%. A discrepancy of 5% with the experimental data is reported even with the best available curve fits [1].

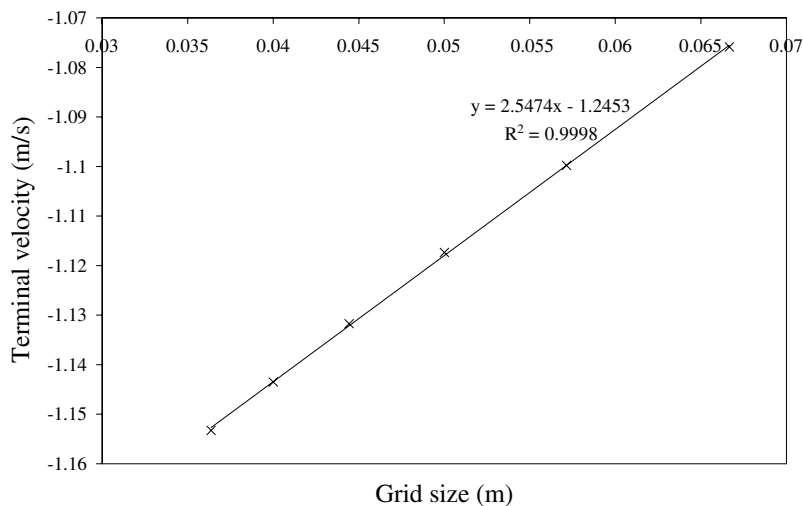


Fig. 3. Plot of the terminal velocity as a function of grid size for $\alpha = 0.3125$.

The inset equation for the trend line shows that the algorithm is first-order with respect to the grid size. The fluid solver was second-order for a constant property fluid. The reduction in the accuracy can be attributed to the simple linear interpolation used, in the rigid body projection step and, to account for the different material property (density) in the particle domain. It can also be caused by the explicit update of the particle positions. The explicit scheme is first-order with respect to time stepping. This directly influences the calculation of the convection term in the momentum equation, thus affecting the spatial accuracy.

Being first-order is not a limitation of the approach and as discussed before a second-order scheme for the rigidity constraint can be envisaged analogous to the second-order pressure correction based fractional step schemes for the incompressibility constraint.

The accuracy of the solution in the vicinity of the particle boundary depends on the interpolation schemes used in the integrations in Eqs. (32) and (33). A better interpolation scheme, compared to the currently used linear interpolation, could improve the accuracy of the method.

The time spent in the computation of the volume fraction and the particle projection step is less than 1% of the total time spent for all the computations presented here. Thus, we see that this approach has very little computational overhead for the calculation of particle motion compared to the fluid solver. This is true even when the number of particles are increased as in the case of results presented in Section 5.3.

To further validate the code, different test cases were considered. The first test case involved checking the drag on a sphere of radius 0.2 m settling at finite Reynolds numbers. A blocking ratio of 0.2 was taken. The drag coefficient C_D and the Reynolds number Re of the flow are given by,

$$C_D = \frac{8F_D}{\rho_f U^2 \pi d^2}, \quad (40)$$

$$Re = \frac{\rho_f U d}{\mu}, \quad (41)$$

where U is the terminal velocity of the sphere, and the drag on the sphere, F_D , is given by

$$F_D = \frac{\pi}{6} d^3 (\rho_s - \rho_f) g. \quad (42)$$

The Reynolds number was varied by changing the fluid and particle densities, and the fluid viscosity. The domain size is the same as that used in the previous case. Table 1 gives the values of drag coefficient C_D as a function of the Reynolds number Re obtained numerically and experimentally [18] for a blocking ratio of 0.2 and a grid size of 0.05.

For a blocking ratio of 0.3125 and the same grid size as that used in forming Table 1, the percentage difference is approximately 11%. As stated above, in the limit of the grid size approaching zero, for a blocking ratio of 0.3125, the difference reduces to 6%. Similarly, the difference in Table 1 will decrease as the grid size decreases further.

Table 1

Comparison of the numerical and experimental [18] values of drag coefficients for varying Reynolds number at a blocking ratio of 0.2

Reynolds number	Drag coefficient		Difference (%)
	Numerical	Experimental [18]	
41.8	2.29	2	13.01
40.4	2.28	2	12.26
33	2.56	2.12	14.61
32.4	2.49	2.12	12.27
11.7	4.9	4.27	12.97

We have further verified our rigid body projection scheme presented here but applied to steady state Stokes flow around a periodic array of spheres. The computed drag coefficient was found to be in good agreement with the analytical values within 2–5%. This is published elsewhere [24,25].

Mordant and Pinton [26] presented experimental data for the sedimentation of spheres. Their Reynolds numbers were typically large and the blockage ratio was very small (~ 0.005), i.e., the wall effects were negligible. In fact their data agree with those compiled by Brown and Lawler [1]. Our minimum blockage ratio was 0.2 below which it is computationally expensive for our current capability. However, Mordant and

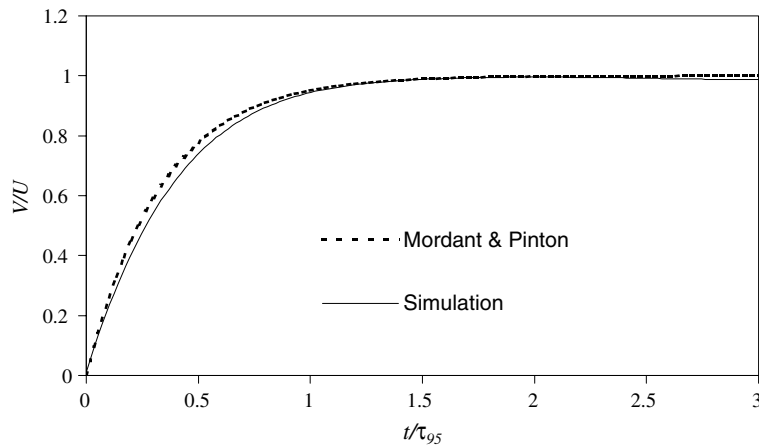


Fig. 4. Plot of the sphere velocity V versus time compared with the experimental result of Mordant and Pinton [26]. U is the terminal velocity and τ_{95} is the time it takes for the sphere to reach 95% of the terminal velocity. The finest grid case depicted in Fig. 3 is considered here.

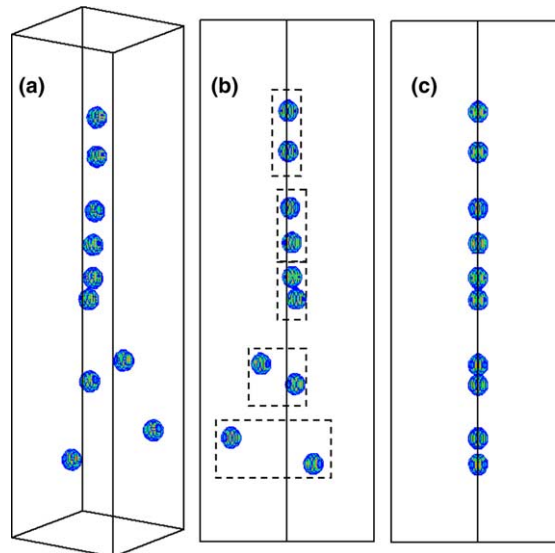


Fig. 5. Drafting, kissing and tumbling for two sedimenting spheres. (a) A 3D view. (b) A view in the diagonal plane in which tumbling occurs. (c) A view in the other diagonal plane. The two-spheres at same times are identified by dashed boxes in figure (b).

Pinton [26] data can be used to verify the transient behavior of sedimenting particles similar to the test done by Dong et al. [5] for their spectral DLM code. This comparison is shown in Fig. 4. The agreement is found to be good. Similar agreement was seen for all the sphere cases considered here.

In the remaining sections we will consider test cases to verify that the numerical scheme captures the key qualitative aspects of some important flows.

5.2. Drafting, kissing and tumbling for two interacting spheres sedimenting in a Newtonian fluid

The case of two spheres falling in a Newtonian fluid was considered to further validate our method. An important mechanism in this case is known as “drafting, kissing and tumbling”: see Fortes et al. [7]. The leading particle creates a wake of low pressure. The trailing particle is caught in its wake. It experiences lower drag hence falls faster than the leading one. This phenomenon is called drafting. The increased speed of the trailing particle impels a kissing contact with the leading particle. In kissing contact the two spheres

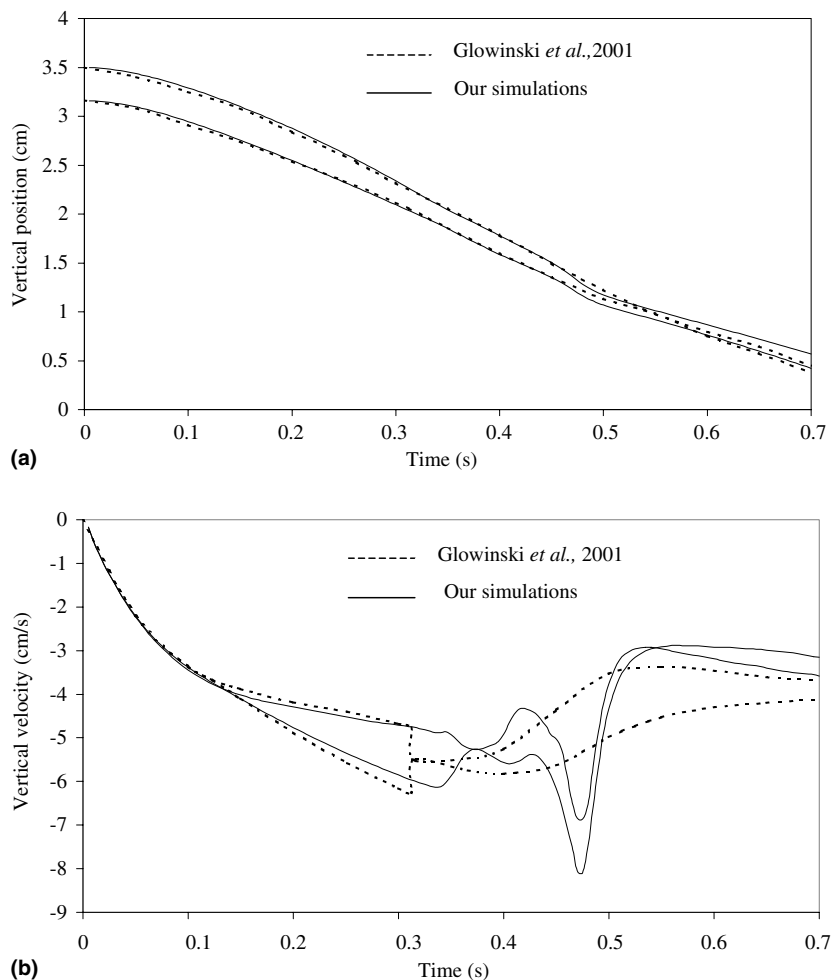


Fig. 6. Plots for drafting, kissing and tumbling of spheres shown in Fig. 5. Comparison of: (a) the vertical positions of the spheres and (b) the vertical velocities of the spheres with the results by Glowinski et al. [10].

form a long body with the line of center along the stream. This state is unstable in a Newtonian fluid and as a result the particles tumble under the influence of a couple which tries to bring them in broadside stable state. The tumbling particles in a Newtonian fluid induce anisotropy of suspended particles since on the average the line of centers between particles must be across the stream.

Fig. 5 shows numerical results for the case of two spheres of density 1.14 g/cm^3 and radius 0.083 cm falling under gravity in a rectangular channel of cross section $1 \text{ cm} \times 1 \text{ cm}$ and height 4 cm . A Newtonian fluid of viscosity 0.01 g/cm s and density 1 g/cm^3 is considered. The spheres were initially located at heights of 3.5 and 3.16 cm , respectively. This case is chosen to be identical to that presented by Glowinski et al. [10]. As can be seen from the snap shots taken at times $0.05, 0.25, 0.35, 0.5$ and 0.7 s the spheres exhibit the known drafting, kissing and tumbling behavior. Tumbling occurs in one of the diagonal planes and the two spheres move on either side of the centerline.

Figs. 6(a) and (b) show comparison, of the vertical positions and velocities of the spheres, with the results of Glowinski et al. [10]. As expected the comparison is found to be good up to the kissing of the particles. It is known that the tumbling phenomenon is the realization of an instability. Different modes of tumbling have been reported in literature e.g. tumbling in a diagonal plane [10] or a plane parallel to the channel walls [4]. Particle positions with respect to the centerline after tumbling can also be different e.g. both particles on same side of the centerline [10] or both particles on opposite side of the centerline [4]. Thus the details of the motion after kissing depend on how the tumbling occurs; an exact agreement in this part is not expected. The mode of tumbling depends on how the instability grows in a particular numerical implementation. This is clear from Figs. 6(a) and (b), where our results are different from those by Glowinski et al. [10] in the tumbling and later phases. In both cases the tumbling occurs in the diagonal plane, however, in our case the particles move on either side of the centerline while in case of Glowinski et al. [10] they move to the same side of the centerline. This leads to different velocity and position plots. In the position plot we see that the lagging particle remains lagging in our case while in case of Glowinski et al. [10] the lagging particle moves ahead (hence the cross-over in dotted lines in Fig. 6(b)).

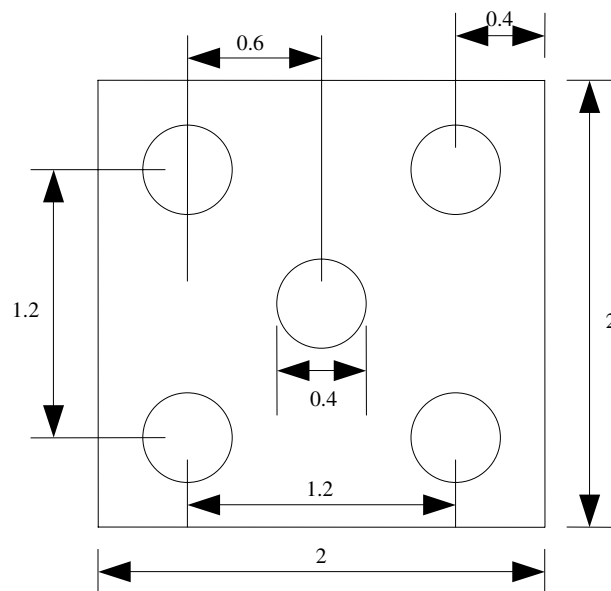


Fig. 7. Top view of the arrangement of the spheres in a single layer (all units are in meters).

Our simulations are for a mesh size of 1/40 cm and a time step of 0.005 s. A collision scheme similar to that by Glowinski et al. [9,10] was used to prevent particles from overlapping. The parameters in the collision model per the definitions in Glowinski et al. [9] are $\varepsilon = 3 \times 10^{-5}$ and $\rho = 0.05$ (note that ρ is not the density) in CGS units. The mesh and time step of Glowinski et al. [10] is more refined; correspondingly they are able to choose parameters such that the collision is more like those of hard spheres unlike ours.

5.3. Sedimentation of many spheres

Simulation was done for the sedimentation of 10 spheres to show that the scheme is capable of handling many spheres. Fig. 7 shows a top view of the arrangement of the spheres in one layer.

Two such layers at heights 7.65 m and 7.05 m were considered. Spheres of density 5 kg/m^3 and radius 0.2 m fall under gravity in a rectangular channel of cross section $2 \text{ m} \times 2 \text{ m}$ and height 8 m. A Newtonian fluid of viscosity 0.05 kg/m s and density 3 kg/m^3 is considered.

Fig. 8(a) shows a three dimensional view of the sedimenting spheres and Fig. 8(b) shows the side views. Snapshots were taken at times 0.25, 2.25, 4.25 and 6.25 s.

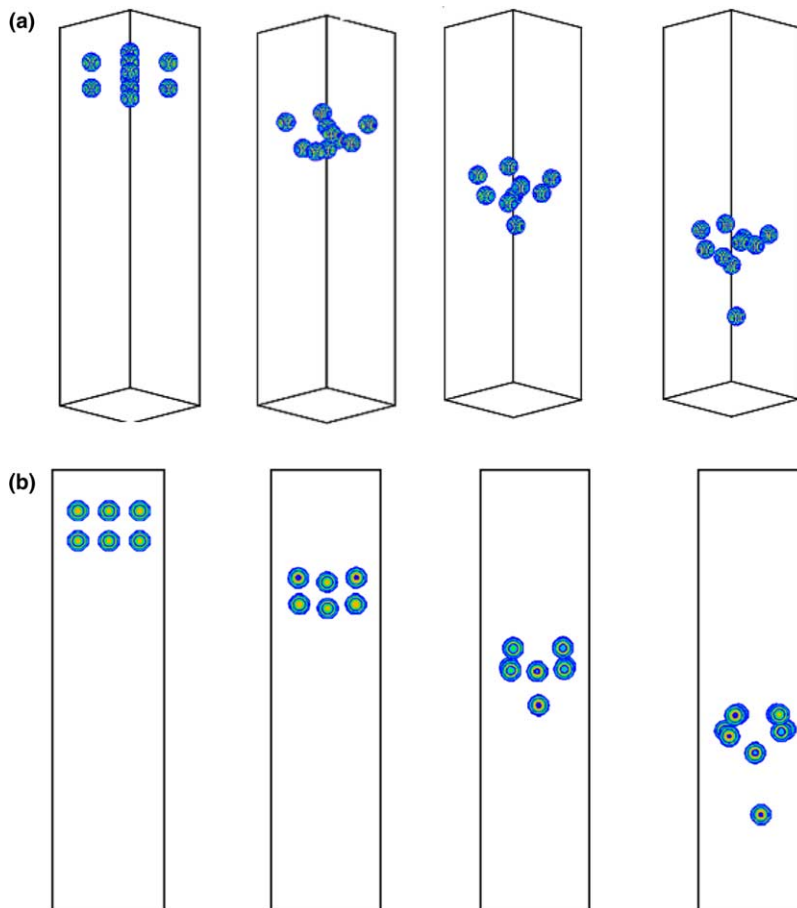


Fig. 8. (a) Three dimensional view of sedimenting spheres. (b) Side view of the sedimenting spheres.

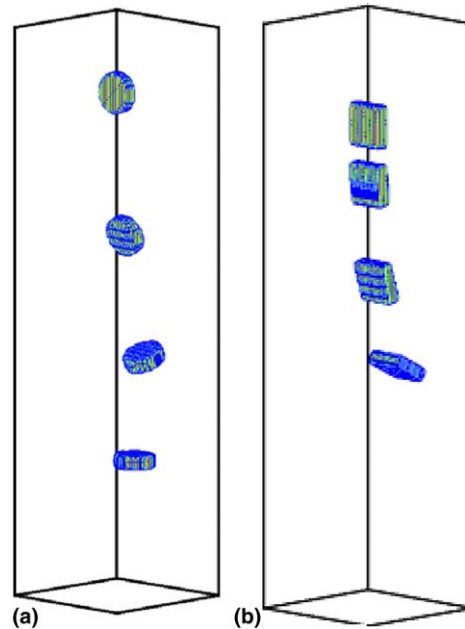


Fig. 9. (a) A disc falling in a rectangular channel filled with a Newtonian fluid. (b) A plate falling in a rectangular channel filled with a Newtonian fluid.

As can be seen from Figs. 8(a) and (b) the sedimenting spheres exhibit instability similar to Rayleigh Taylor instability. The inner spheres fall faster than the outer ones.

5.4. Different geometry results

Numerical tests were done for different geometries (other than a sphere) sedimenting in a Newtonian fluid. These were carried out to demonstrate that the scheme is capable of handling irregular geometries without significant increase in computational overhead. A disc and a plate were considered with initial long side position. The initial long side position changed to broad side position during the fall. The channel dimensions were the same as that used for the sphere case. The fluid properties were taken as: viscosity 0.05 kg/m s and density 1 kg/m^3 . The particle density was 2.725 kg/m^3 . Fig. 9(a) shows a disc of radius 0.3 m and thickness 0.2 m falling under gravity. As is evident from the snap shots of the fall at times $0.25, 2, 4$ and 6 s , the initial unstable long side position changes to broad side. Fig. 9(b) shows a similar trend for the case of a square plate of cross section $0.6 \text{ m} \times 0.6 \text{ m}$ and thickness 0.2 m . In this case snap shots were taken at $0.25, 1, 2$ and 3 s .

6. Conclusion

We have presented a numerical scheme for the fast computation of rigid particulate flows. Convergence tests showed the fluid–particle solver to be first-order. Code validation was done by comparing numerical results with known experimental data. It is found that the method is effective in handling three-dimensional regular and irregular geometries without significant additional computational overhead. The computational time required for the particle solver is less than 1% for the cases considered here.

The proposed numerical method does not include additional equations of motion for the particle translational and angular velocities. The application to irregular shaped bodies is straightforward. The capability for fast computation is rooted in a simple projection scheme for rigid motion (Eq. (24)). At the same time the formulation lays a theoretical framework analogous to the Chorin [3] type fractional step schemes for incompressible fluids. Improvement in the performance of this method is possible by developing higher order time stepping schemes analogous to the pressure correction based fractional step schemes [16] for the incompressibility constraint.

The ability to perform fast computations is suitable for the application of this method to turbulent particulate flows. This should be investigated especially in light of a method reported by Kajishima et al. [15]. It is evident [20], that the method presented here, may be used to improve upon the earlier method by Kajishima et al. [15]. Since the technique is not restricted to any constitutive model of the suspending fluid it may also be used in large eddy simulations (LES) or Reynolds averaged Navier–Stokes (RANS) type simulations.

The ease of handling irregular shaped bodies and the fast computation capability has enabled the application of this technique in animation. Carlson et al. [2] have applied this technique to animate the interplay between rigid bodies and fluid. They call it the rigid fluid method. They have reported several animations of objects of different shapes.

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