Lagrangian numerical simulation of particulate flows

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Abstract

The Lagrangian numerical simulation (LNS) scheme presented in this paper is motivated by the multiphase particle-in-cell (MP-PIC). In this numerical scheme we solve the fluid phase continuity and momentum equations on an Eulerian grid. The particle motion is governed by Newton's law thus following the Lagrangian approach. Momentum exchange from the particle to fluid is modeled in the fluid phase momentum equation. Forces acting on the particle include drag from the fluid, body force and force due to interparticle stress. There is freedom to use different models for these forces and to introduce other forces. The effect of viscous stresses are included in the fluid phase equations. The volume fraction of the particles appear in the fluid phase continuity and momentum equations. A finite volume method is used to solve for the fluid phase equations on an Eulerian grid. Particle positions are updated using the Runge-Kutta scheme. This numerical scheme can handle a range of particle loadings and particle types.

The LNS scheme is implemented using an efficient three-dimensional time dependent finite volume algorithm. We use a Chorin-type pressure-correction based fractional-step scheme on a non-staggered cartesian grid. In this paper, we consider only incompressible Newtonian suspending fluid. However, the average velocity field of the fluid phase is not divergence-free because its effective density is not constant. Our pressure correction based fractional-step scheme accounts for varying properties in the fluid phase equations. This method can also account for suspending fluids with non-constant properties. The numerical scheme is verified by comparing results with test cases and experiments.

Key Words: Approximate factorization, Two-phase flow, Eulerian-Lagrangian numerical scheme (LNS), multiphase particle-in-cell (MP-PIC) method, particulate flows, three-dimensional time dependent finite volume approach, Chorin scheme, pressure-correction
1 Introduction

Numerical simulations of particulate flows find application in various settings; e.g. sedimenting and fluidized suspensions, lubricated transport, hydraulic fracturing of reservoirs, slurries, sprays etc. Numerical schemes based on mathematical models of separated particulate multiphase flow have used the continuum approach for all the phases (Gidaspow 1994) or a continuum approach for the fluid phase and a Lagrangian approach for the particles (Williams 1985).

Continuum/continuum approaches consider the particulate phase to be a continuous fluid interpenetrating and interacting with the fluid phase. These approaches are commonly used for dense particulate flows since it is convenient to model the interparticle stresses using spatial gradients of the volume fraction (Batchelor 1988, Gidaspow 1994). However, for multimodal simulations one has to consider each particle type (i.e. particles of different size or material) as a separate phase. This requires solving extra continuity and momentum equations for each additional phase. As a result, this formulation is more commonly used for two-phase rather than multi-mode flow simulations.

It is more convenient to use a Lagrangian description for the particle phase and an Eulerian approach for the fluid phase for multimodal simulations. In this formulation each computational particle (called a parcel) is considered to represent a group of particles possessing the same characteristics such as size, composition etc. Use of Lagrangian approach for the particle phase also solves the problem of numerical diffusion. It has been found that the required number of parcels to accurately represent the particle phase is not excessive (Dukowicz 1980).

The Eulerian/Lagrangian approach can be applied under various assumptions. In some problems, such as the dispersion of atmospheric pollutants, it may be assumed that the particles do not perturb the flow field. The solution then involves tracing the particle trajectories in a known velocity field (Gauvin, Katta and Knelman 1975). In other
problems the particles carry sufficient momentum to set the surrounding fluid in motion. In this case it is necessary to include the fluid-particle momentum exchange term in the fluid phase equation. However, the volume occupied by the particles in a computational cell in comparison with the volume of the fluid may still be neglected (Crowe, Sharma and Stock 1977). Dukowicz (1980) presented a time-splitting numerical technique that accounted for full coupling between fluid and particle phase. In this approach the volume occupied by the particles in a computational cell is not neglected and a momentum exchange term is included in fluid phase equation but the particles are assumed to be sufficiently dispersed so that particle collisions are infrequent. This approach is appropriate for many liquid spray applications. Particle collision frequencies are high when the particulate volume fractions are above 5% and should be accounted for in the numerical method. Lagrangian collision calculations are not suitable to resolve the interparticle stress arising from collisions in the Eulerian/ Lagrangian approach for dense particulate flows.

Andrews and O’Rourke (1996) and Snider, O’Rourke and Andrews (1998) presented a multiphase particle-in-cell (MP-PIC) method for particulate flows that accounts for full coupling between the fluid and particle phase as well as the interparticle stress due to collisions. The fluid phase is assumed to be inviscid where viscosity is significant on the scale of the particles and is used only in the particle drag formula. In this approach the particle phase is considered both as a continuum and as a discrete phase. Interparticle stresses are calculated by treating the particles as a continuum phase. Particle properties are mapped to and from an Eulerian grid. Continuum derivatives that treat the particle phase as a fluid are evaluated to model interparticle stress and then mapped back to the individual particles. This results in a computational method for multiphase flows that can handle particulate loadings ranging from dense to dilute and particles of different sizes and materials.

The assumption of an inviscid fluid phase is not suitable for all applications, e.g. flow of particulate mixture in hydraulic fractures. In this paper we present an Eulerian/ Lagrangian numerical simulation (LNS) scheme for particulate flows in three-dimensional geometries. We extend the MP-PIC approach of Andrews and O’Rourke (1996) to include the effect of viscous stress in the fluid phase equations. Snider,
O’Rourke and Andrews (1998) used finite-volume SIMPLER algorithm (Patankar 1980) with staggered grid for velocity and pressure to solve for an unsteady two-dimensional flow of particles suspended in an incompressible fluid. Since time-splitting algorithms are better suited for unsteady flow calculations we use a finite-volume Chorin-type (Chorin 1968) pressure-correction based fractional-step scheme to solve the fluid phase equations in cartesian coordinate system. Our pressure-correction based fractional-step scheme accounts for varying properties in the fluid phase equations. We use a non-staggered grid for velocity and pressure (Rhie and Chow 1982). We do so because we intend to extend the scheme to three-dimensional curvilinear coordinate system in future. The non-staggered grid requires less storage memory than the staggered grids in addition to other advantages (Perić, Kessler and Scheuerer 1988, Zhang, Street and Koseff 1994).

The fluid and particle phases exchange momentum through the hydrodynamic forces acting on the particle surface. Models for these forces can be developed through experimental investigation. The development of direct numerical simulation (DNS) techniques for particulate flows (Hu et al. 1992, Hu 1996, Johnson and Tezduyar 1997, Glowinski, Pan, Hesla and Joseph 1999, Patankar, Singh, Joseph, Glowinski and Pan 1999) have given rise to an invaluable tool for modeling the hydrodynamic forces in many applications. It is straightforward to use these models in the LNS technique. One of our future objectives is to model the hydrodynamic force using our DNS capability. These models will then be implemented in the LNS scheme which is computationally less intensive than the DNS methods.

In the next section we will present governing equations for the Eulerian/ Lagrangian formulation. In section 3 the numerical scheme will be explained. This computational scheme will then be tested in section 4 by comparing calculated and experimentally measured sedimentation rates (Davis, Herbolzheimer and Acrivos 1982) for bimodal suspension. This test case was also used by Snider et al. (1998). This scheme will then applied to the sedimentation of particles initially placed at the top of the sedimentation column. As observed in the direct numerical simulation by Glowinski, Pan Hesla and Joseph (1999), we see a fingering motion of the particles which are heavier than the suspending fluid. This phenomenon is similar to the Rayleigh-Taylor instability of heavy fluid on top of a lighter fluid. Simulation results for the flow of a particulate mixture in a
reservoir fracture will be qualitatively compared with known experimental results of Barree and Conway (1995). Results on sedimentation in an inclined vessel will be reported. Conclusions will be stated in section 5.

2 Governing equations

We use equations of motion for the average properties of the fluid phase. Different forms of the averaged equations of motion for the fluid phase have been proposed in literature (see Joseph and Lundgren 1990 and references therein). Any of these equations of motion for the fluid phase can be used in this formulation without any fundamental difficulty. In the present work we use the fluid phase equations proposed by Tsuo and Gidaspow (1990). The continuity equation is given by

\[ \frac{\partial}{\partial t} (\rho_f, \theta_f) + \nabla \cdot (\rho_f, \theta_f, \mathbf{u}_f) = 0, \]

where \( \rho_f \) is the density of the suspending fluid, \( \theta_f \) is the fluid volume fraction and \( \mathbf{u}_f \) is the average velocity of the fluid phase. The average velocity field of the fluid phase is not divergence-free even if the suspending fluid is incompressible. The momentum equation for the fluid phase is given by

\[ \frac{\partial}{\partial t} (\rho_f, \theta_f, \mathbf{u}_f) + \nabla \cdot (\rho_f, \theta_f, \mathbf{u}_f, \mathbf{u}_f) = -\nabla p + \nabla \cdot \left[ \theta_f \mu_f \nabla \mathbf{u}_f \right] \\
+ \nabla \cdot \left[ \theta_f \mu_f (\nabla \mathbf{u}_f)^T \right] - \nabla \left\{ \frac{2}{3} \theta_f \mu_f \nabla \cdot \mathbf{u}_f \right\} - \mathbf{F}, \]

where \( p \) is the dynamic pressure in the fluid, \( \mu_f \) is the viscosity of the fluid phase and \( \mathbf{F} \) is the rate of momentum exchange per unit volume between the fluid and particle phases. Dukowicz (1980) had a viscous term similar to the second expression on the RHS of (2). The body force due to gravity acting on the fluid phase is balanced by the hydrostatic pressure and does not appear explicitly in the momentum equation. Expressions for \( \mathbf{F} \) will be presented later in this section. An equation of state for the fluid phase is in the form
\[ \rho = f(p, T), \]
\[ f(p, T) = \frac{p}{RT} \quad \text{for an ideal gas}, \]
\[ f(p, T) = \text{constant for a constant density fluid}, \]

where \( T \) is the temperature, \( R \) is the gas constant and \( f \) is some function of pressure and temperature.

We use the particle phase equations as given by Andrews and O’Rourke (1996); evolution of the particle phase is governed by a Liouville equation for the particle distribution function \( \phi(x_p, u_p, \rho_p, V_p, t) \)

\[ \frac{\partial \phi}{\partial t} + \nabla \cdot (\phi \cdot u_p) + \nabla \cdot (\phi \cdot A) = 0, \quad (4) \]

where \( x_p \) is the particle position, \( u_p \) is the particle velocity, \( \rho_p \) is the particle density and \( V_p \) is the particle volume. \( A \) is the particle acceleration given by

\[ A = \frac{du_p}{dt} = D_p \left( u_f - u_p \right) - \frac{1}{\rho_p} \nabla \rho_p + \left( 1 - \frac{\rho_f}{\rho_s} \right) g - \frac{1}{\theta_s \rho_p} \nabla \tau, \quad (5) \]

where \( \theta_s \) is the particle volume fraction. Equation (5) models acceleration due to hydrodynamic drag, dynamic pressure gradient, net buoyant force and gradients in the interparticle stress \( \tau \). Different appropriate models for force on the particle can also be used in (5). In this work we use \( D_p \) given by (Andrews & O’Rourke 1996)

\[ D_p = C_d \frac{3 \rho_f}{8 \rho_p} \frac{|u_f - u_p|}{R_p}, \]

where

\[ C_d = \frac{24}{\text{Re}} \left( \theta_f^{2.65} + \frac{\text{Re}^{2/3}}{6} \theta_f^{-1.78} \right), \quad \text{Re} = \frac{2 \rho_f |u_f - u_p| R_p}{\mu_f}, \quad R_p = \left( \frac{3 V_p}{4 \pi} \right)^{1/3}, \]
Cd is the drag coefficient, Re is the Reynolds number and Rp is the particle radius (we assume spherical particles). Particle-particle collisions are modeled by an isotropic interparticle stress given by (Harris and Crighton 1994)

$$\tau = \frac{P_s \theta_s^\beta}{\theta_{cs} - \theta_s},$$  

(7)

where $P_s$ has units of pressure, $\theta_{cs}$ is the particle volume fraction at close packing and $\beta$ is a constant. A discussion of the factors entering into the choice of $P_s$ and $\beta$ is given by Snider et al. (1998). In (7) it is assumed that acceleration of a particle due to interparticle stress is independent of its size and velocity. Any stress model that accounts for the effect of particle size and velocity can replace (7) when available. The particle volume fraction $\theta_s$ is defined by

$$\theta_s = \iiint \phi \, V_p \, dV_p \, d\rho_p \, d\mathbf{u}_p.$$  

(8)

Fluid volume fraction $\theta_f$ is then given by

$$\theta_f = 1 - \theta_s.$$  

(9)

The interphase momentum transfer function $F$ is given by

$$F = \iiint \phi \, V_p \rho_p \left[ D_p \left( \mathbf{u}_f - \mathbf{u}_p \right) - \frac{1}{\rho_p} \nabla P \right] \, dV_p \, d\rho_p \, d\mathbf{u}_p.$$  

(10)

Andrews and O’Rourke (1996) highlighted several important features of this formulation. The previous Eulerian/ Lagrangian formulations (e.g. Dukowicz 1980) ignored the interparticle stress term which is modeled in the present approach. It can be shown by deriving the average momentum equation for particle phase from (4) that this formulation
accounts for the kinematic stress that arises from local particle velocity fluctuations about the mean velocity. The effect of buoyancy driven currents is modeled in this formulation. This can be verified by adding the average momentum equations for the fluid and particle phases. This is also confirmed by our numerical simulation of the Rayleigh-Taylor instability (to be presented later) during the sedimentation of particles initially placed at the top in a sedimentation column.

3 Numerical scheme

We use a finite-volume method on an Eulerian grid to solve the fluid phase equations in cartesian coordinate system. A non-staggered grid for velocity and pressure (Rhie and Chow 1982) is used. The particle phase equations are solved by considering the motion of a finite number of computational particles which represent a sample of the total population of particles. Each computational particle, henceforth referred to as a parcel, is considered to represent a group of particles of identical size, velocity and position.

3.1 Interpolation scheme

In order solve the particle equation of motion it is necessary to interpolate variables to the particle position. Similarly the solution of fluid phase equations requires the calculation of variables on the Eulerian grid. This requires the interpolation of these variables from particle location to the Eulerian grid. This is accomplished by using bilinear interpolation function formed from the product of linear interpolation functions in the x, y and z directions (see Snider et al. 1998). The bilinear interpolation function $S_{ijk}(x)$ is unity at a given grid node $(i,j,k)$ which is at the cell center and decreases to zero at the 26 neighboring nodes and the domain beyond these neighboring nodes. The position $x_p$ of any particle can be located in a box defined by eight Eulerian grid nodes surrounding it. The sum of the eight interpolation functions, due to the surrounding nodes, at a particle location is unity.

The particle volume fraction on the Eulerian grid is calculated by
\[ \theta_{ijk} = \frac{1}{V_{ijk}} \sum_p N_p V_p S_{ijk}(x_p), \]  

(11)

where \( \theta_{ijk} \) is the particle volume fraction at grid node (i,j,k), \( V_{ijk} \) is the volume of the Eulerian cell (i,j,k) and \( N_p \) is the number of particles in a parcel. The fluid volume fraction \( \theta_{fijk} \) at grid node (i,j,k) immediately follows from (9).

An example of the interpolation of a variable from the grid node to particle position is given by

\[ u_{fp} = \sum_{\zeta=1}^8 s_{\zeta}(x_p) u_{f\zeta}, \]  

(12)

where \( u_{fp} \) is the fluid velocity at the particle location and \( \zeta \) is an index for the eight grid nodes bounding the particle.

The interphase momentum transfer \( F_{ijk} \) at a grid node (i,j,k) is evaluated by an interpolation scheme given by Snider et al. 1998. According to this scheme the expression for \( F_{ijk} \) is given by

\[ F_{ijk} = \frac{1}{V_{ijk}} \sum_p \left\{ \rho_p V_p N_p S_{ijk}(x_p) \left[ D_p \left( u_{fp} - u_p \right) - \frac{1}{\rho_p} \nabla p_p \right] \right\} \]  

\[ = \frac{1}{V_{ijk}} \sum_p \left\{ \rho_p V_p N_p S_{ijk}(x_p) \left[ D_p \left( u_{fijk} - u_p \right) - \frac{1}{\rho_p} \nabla p_{fijk} \right] \right\} \]  

(13)

This gives a less diffusive interpolation scheme and increases the diagonal dominance of the momentum equation of the fluid phase.

### 3.2 Numerical algorithm

We use an explicit update of the particle positions in which the translational motion of the particles is solved based on the fluid velocity and pressure fields at the end of the
previous time step. Motion of each particle type represents the motion of the parcel it represents. In our numerical algorithm we first solve for the particle equations of motion. This is followed by a solution of the fluid phase equations. The fluid properties $\rho_f$ and $\mu_f$ are constant.

3.2.1 Numerical scheme for particle motion

Given the solution at the end of $n$ time-steps i.e. given $u_p^n$, $x_p^n$ for particles in all the parcels, $u_p^n$ and $p^n$, compute $u_p^{n+1}$ and $x_p^{n+1}$ by the following procedure:

For particles in all the parcels:
Set $u_p^{n,0} = u_p^n$, $x_p^{n,0} = x_p^n$.

do k=1,K

$$u_p^{n,k} = u_p^{n,k-1} + \frac{\Delta t}{K} f_1, \text{ where}$$

$$f_1 = \left[ \left( D_p u_p^n - \frac{1}{\rho_p} \nabla p^n + \left( 1 - \frac{\rho_f}{\rho_p} \right) g \right) - D_p u_p^{n,k} - \frac{1}{\theta_s \rho_p} \nabla \tau \right] \text{at } x_p^{n,k-1} \right)$$

$$x_p^{n,k} = x_p^{n,k-1} + \frac{\Delta t}{K} \left( \frac{u_p^{n,k-1} + u_p^{n,k}}{2} \right)$$

$$u_p^{n,k} = u_p^{n,k,l} + \frac{\Delta t}{K} \left( f_1 + f_2 \right), \text{ where}$$

$$f_2 = \left[ \left( D_p u_p^n - \frac{1}{\rho_p} \nabla p^n + \left( 1 - \frac{\rho_f}{\rho_p} \right) g \right) - D_p u_p^{n,k} - \frac{1}{\theta_s \rho_p} \nabla \tau \right] \text{at } x_p^{n,k} \right)$$

10
enddo
Set $u_{p}^{n+1} = u_{p}^{n,K}$, $x_{p}^{n+1} = x_{p}^{n,K}$. Calculate $\theta_{s}^{n+1}$ by using (11).

Here, $\Delta t$ is the time-step. In this step the effective time-step for the motion of particles is reduced if the value of $K$ is greater than one. This is helpful to avoid instability due to the interparticle stress term. In our numerical simulations $K$ varies between 2 to 5. Value smaller than 2 led to instabilities. The choice of $K$ depends on the application and is usually set ad hoc based on numerical trials. Eventually a systematic choice of $K$ is desirable and is under investigation. Note that the original MP-PIC approach used an implicit scheme to avoid using small time steps. A similar scheme was used by Glowinski, Pan, Hesla and Joseph (1999) to solve for the motion of particles in their direct numerical simulation method for fluid-particle mixtures. Near the close packing limit we use a large but finite value for the particle stress. Our scheme is not completely robust near the close packing limit although we did simulate some challenging problems e.g. see fig. 4-7. Improvement of our scheme in this regard is the subject of our current investigation.

3.2.2 Numerical scheme for fluid phase equations

We use a non-staggered cartesian grid to solve the fluid phase equations (Rhie and Chow 1982). The pressure and the cartesian velocity components are defined at the center of the control volume. Mass fluxes are defined at the mid-point of their corresponding faces of the control volume. We use a Chorin-type fractional-step method (Chorin 1968) based on a pressure correction scheme. This pressure correction based scheme has several features similar to the SIMPLE approach of Patankar and Spalding (1972). A fractional-step based pressure correction scheme for the turbulent flow of a Newtonian fluid was
presented by Comini and Del Giudice (1985). A variation of the fractional-step pressure correction approach was also presented by Choi, Choi and Yoo (1996). The fluid phase equations are solved by the following solution procedure:

(1) Given \( u^n_f, p^n, \theta^n_s, \theta^{n+1}_s, u^n_p \) and \( x^{n+1}_p \), compute the intermediate velocity \( u^*_f \) at the grid nodes by solving:

\[
\frac{\rho_f \theta^{n+1}_f}{\Delta t} u^*_f + D_f u^*_f + \mathfrak{S} u^*_f = -\nabla p^n + \frac{\rho_f \theta^n_f}{\Delta t} u^n_f \\
+ (\nabla u^n_f) \cdot \nabla \left( \theta^{n+1}_f \mu_f \right) - (\nabla \cdot u^n_f) \nabla \left( \theta^{n+1}_f \mu_f \right) \\
+ \theta^{n+1}_p \nabla p^n + \frac{1}{V} \sum_p \left[ \rho_p V_p N_p S_{ijk} \left( x^{n+1}_p \right) D_p u^{n+1}_p \right], \forall \text{ grid node } ijk,
\]

where \( P \) is defined by

\[
P = p - \frac{1}{3} \theta_f \mu_f \nabla \cdot u_f.
\]

and

\[
D_f = \frac{1}{V} \sum_p \left[ \rho_p V_p N_p S_{ijk} \left( x^{n+1}_p \right) D_p \right], \text{ for any grid node } ijk.
\]

\( \mathfrak{S} \) is the convection-diffusion operator whose operation on any vector \( v \) is given by

\[
\mathfrak{S} v = \nabla \cdot \left\{ \left( \rho_f \theta^{n+1}_f u^n_f - \theta^{n+1}_f \mu_f \nabla \right) v \right\}.
\]

Time discretization of (2) and appropriate rearrangement of the terms results in (18). A ‘half implicit’ expression, \( \nabla \cdot \left( \rho_f \theta^{n+1}_f u^n_f u^{n+1}_f \right) \), is used for the convection term. This expression is first-order accurate. Fluid phase velocity \( u^n_f \) in the convection term can be
replaced by a second-order accurate expression given by $2\mathbf{u}_i^n - \mathbf{u}_i^{n-1}$ (Turek 1996). We can employ approximate factorization technique (Beam and Warming 1976, Briley and McDonald 1977) in which (18) is factorized as

$$\left( \frac{C_u}{\Delta t} \left[ 1 + \frac{\Delta t}{C_u} \mathcal{S}_x \right] \left[ 1 + \frac{\Delta t}{C_u} \mathcal{S}_y \right] \left[ 1 + \frac{\Delta t}{C_u} \mathcal{S}_z \right] \right) (\mathbf{u}_i^n - \mathbf{u}_i^{n-1}) =$$

$$- \nabla \mathbf{P}^n + \frac{\rho_i}{\Delta t} \mathbf{u}_i^n \left( \theta_i^n - \theta_i^{n+1} \right) - D_i \mathbf{u}_i^n - \mathcal{S} \mathbf{u}_i^n \quad (22)$$

$$+ \left( \nabla \mathbf{u}_i^n \right) \cdot \nabla \left( \theta_i^{n+1} \mu_f \right) - \left( \nabla \cdot \mathbf{u}_i^n \right) \nabla \left( \theta_i^{n+1} \mu_f \right)$$

$$+ \theta_p^{n+1} \nabla \mathbf{P}^n + \frac{1}{V} \sum_{ijk} \rho_p \mathbf{v}_p \mathbf{N}_p \mathbf{S}_ijk \left( x_p^{n+1} \right) D_p \mathbf{u}_p^{n+1}, \forall \text{ grid node ijk},$$

where

$$C_u = \rho_i \theta_i^{n+1} + D_i \Delta t,$$

$$\mathcal{S}_x \mathbf{v} = \frac{\partial}{\partial x} \left\{ \left( \rho_i \theta_i^{n+1} \mathbf{u}_{fx} - \theta_i^{n+1} \mu_f \frac{\partial}{\partial x} \right) \mathbf{v} \right\},$$

$$\mathcal{S}_y \mathbf{v} = \frac{\partial}{\partial y} \left\{ \left( \rho_i \theta_i^{n+1} \mathbf{u}_{fy} - \theta_i^{n+1} \mu_f \frac{\partial}{\partial y} \right) \mathbf{v} \right\},$$

$$\mathcal{S}_z \mathbf{v} = \frac{\partial}{\partial z} \left\{ \left( \rho_i \theta_i^{n+1} \mathbf{u}_{fz} - \theta_i^{n+1} \mu_f \frac{\partial}{\partial z} \right) \mathbf{v} \right\}. \quad (23)$$

It can be easily shown that the error in this factorization is $O(\Delta t^3)$. An approximate factorization technique in conjunction with fractional-step method to simulate turbulent flows was developed by Kim and Moin (1985).

We can solve either (18) or (22) during our simulations. Both approaches are first-order with respect of temporal discretization and gave similar results. Solving (18) does not involve the additional error of factorization whereas using (22) offers a faster solution procedure. We use the power-law upwinding scheme (Patankar 1980) for this convection-
diffusion problem giving a first-order discretization in computational space. Boundary values of the intermediate velocity are the same as the velocity specified at the boundary. This is permissible for a first-order pressure correction based scheme where (26) (the velocity correction equation to be presented later) shows that the resulting error is $O(\Delta t^2)$ (Choi, Choi and Yoo 1997). The solution of (18) is obtained by a block-correction-based multigrid method (Sathyamurthy and Patankar 1994). This method employs a multilevel correction strategy and is based on the principle of deriving the coarse grid discretization equations from the fine grid discretization equations. The solution of (22) is non-iterative and faster involving solution of three tridiagonal matrices.

(2) Given $\mathbf{u}_i^*$ at the grid nodes, compute the intermediate velocity $(\mathbf{u}_i^*)_{cf}$ on cell faces by linear interpolation. Value of $(\mathbf{u}_i^*)_{cf}$ on the boundary cell faces is calculated by linear extrapolation of the values of $\mathbf{u}_i^*$ at the interior grid nodes. Other upwind interpolation methods such as the QUICK formulation (Leonard 1979) can be used. We consider only linear interpolation scheme in the present work for this computational step.

(3) Given $(\mathbf{u}_i^*)_{cf}$ and $P^n$, compute $P^{n+1}$. Correction of cell face velocity is given by

$$\frac{(\rho_i \theta_i^{n+1} u_i^{n+1})_{cf} - (\rho_i \theta_i^{n+1} u_i^*)_{cf}}{\Delta t} = -\nabla P', \quad (24)$$

where $P' = P^{n+1} - P^n$ is the pressure correction defined at grid node. The pressure correction equation is obtained by using (24) in the continuity equation (1). The discrete form of the equation is then given by

$$\frac{\partial}{\partial x} \left( \frac{\partial}{\partial x} P' \right) + \frac{\partial}{\partial y} \left( \frac{\partial}{\partial y} P' \right) + \frac{\partial}{\partial z} \left( \frac{\partial}{\partial z} P' \right) = \frac{1}{\Delta t} \left\{ (\rho_i \theta_i^{n+1}) - (\rho_i \theta_i^n) \right\}$$

$$+ \frac{\partial}{\partial x} \left[ (\rho_i \theta_i^{n+1} u_{fx}^*)_{cf} \right] + \frac{\partial}{\partial y} \left[ (\rho_i \theta_i^{n+1} u_{fy}^*)_{cf} \right] + \frac{\partial}{\partial z} \left[ (\rho_i \theta_i^{n+1} u_{fz}^*)_{cf} \right], \quad (25)$$
where $\delta/\delta x$, $\delta/\delta y$ and $\delta/\delta z$ represent discrete difference operators in the computational space. We use the velocity specified at the boundary while setting up the pressure correction equation (25). Thus pressure correction at the boundary does not appear in (25) (Patankar 1980). We therefore do not need a boundary condition for pressure correction to solve (25). We use the block-correction-based multigrid method (Sathyamurthy and Patankar 1994) to solve (25). To obtain pressure correction at the boundary we apply (24) at the boundary cell faces where both $(u_{f}^{n+1})_{cf}$ and $(u_{f}^{*})_{cf}$ are known. $p^{n+1}$ follows directly from the solution for $P'$. Velocities at the internal cell faces at the end of the present time-step are computed using (24). These cell face velocities are used to calculate the mass flux in the ‘half implicit’ convection term in the next time step.

(4) Given $P'$, compute $u_{f}^{n+1}$ and $p^{n+1}$. The velocity at grid nodes is corrected by

$$\frac{\left(\rho_{f}\theta_{f}^{n+1}u_{f}^{n+1}\right) - \left(\rho_{f}\theta_{f}^{n+1}u_{f}^{*}\right)}{\Delta t} = -\nabla P',$$

where the gradient of pressure correction at the grid node is calculated by a central difference scheme. The pressure $p^{n+1}$ then follows from (19). The value of pressure in the domain is calculated with respect to the value at some reference point inside the computational space. Convergence is said to be achieved if the residue is less than $10^{-10}$.

4 Numerical results

We validate the numerical scheme by comparing calculated sedimentation rates with the values measured in the experiments of Davis, Herbolzheimer and Acrivos (1982) for a bimodal suspension. The sedimentation column in the experiment was vertical, 100 cm tall and has a square cross-section with each side 5 cm wide. The calculation domain in our simulations have x, y and z dimensions equal to 5 cm, 125 cm and 5 cm, respectively. Gravity acts in the negative y-direction. The suspending fluid is Newtonian with the density and viscosity being 992 kg/m$^3$ and 0.0667 kg/(m-s), respectively. Particles of two
different densities are used in the calculations. The density of the heavy particles is 2990 kg/m³; their diameters vary uniformly between 177 µm to 219 µm. The density of lighter particles is 2440 kg/m³ and their diameters range uniformly between 125 µm to 150 µm. The initial concentration of the heavy particles is 0.01 and that of lighter particles is 0.03. The particles are initially placed randomly with uniform distribution up to a height of 100 cm of the sedimentation column. To model the interparticle stress we choose $P_s = 100$ Pa, $\beta = 3$ and $\theta_{cs} = 0.6$.

In order to check the convergence of the numerical scheme we perform two simulations with different grid size, number of parcels and time steps. In Case A there are 10 control volumes in the x and z directions and 50 control volumes in the y direction. There are 9000 parcels of each type giving a total of 18000 parcels. The number of particles in each parcel is chosen so that the total particle volume in each parcel is the same for heavy and light parcel types, respectively. The time step is 0.05 s. To start the simulation from rest we use smaller time steps beginning from 0.00078125 s and increasing every subsequent time step by a factor of two until it becomes 0.05 s. For Case B we double the number of control volumes and parcels in the domain. The time step is reduced by half. Here we present convergence results for the numerical scheme based on the solution procedure without approximate factorization. Similar convergence tests were also performed for the method of approximate factorization and are not reported here.

Figure 1 shows the transient interface levels of the two types of particles. Here the comparison is made between the LNS calculations from Cases A and B and the experimental data of Davis et al. (1982). We see that they are in good agreement thus validating the calculations by the present numerical procedure. Figure 2 shows the parcel positions in the channel at different times for Case B. Figure 3 shows the particle positions at $t = 300$ s calculated from Cases A and B. We see that they have almost identical particle distributions. The coarse mesh simulation required around 4 MB memory and took less than 8 s CPU time for completing one time step on a SGI machine.

We now apply the scheme to the sedimentation of particles initially placed at the top in a sedimentation column. This is similar to the presence of heavy fluid on top of a light fluid. Such a configuration should then give rise to a Rayleigh-Taylor instability. This instability was indeed reproduced in the direct numerical simulation of particles falling a
Newtonian fluid by Glowinski et al. (1999). To ensure that our numerical scheme is able to capture such instabilities we simulated this case. The x, y and z dimensions of the sedimentation column are 10 cm, 50 cm and 5 cm, respectively. Gravity acts in the negative y-direction. The density and viscosity of the Newtonian suspending fluid are 1000 kg/m³ and 0.00667 kg/(m-s), respectively. The particle density is 2500 kg/m³ and the diameter is 126 µm. Particles are initially placed in the top half of the sedimentation column in a regular array and their volume fraction is 0.1. We use the same parameters as before to model the interparticle stress. There are 17280 parcels in the calculation domain with each parcel having 6910 particles. There are 10 control volumes in the x-direction, 40 in the y-direction and 5 in the z-direction. The time step after the initial transient is 0.05 s.

Figure 4 shows the particle positions in the sedimentation column at different times. Figure 5 shows the three-dimensional view at t = 17.5 s. It is seen that a Rayleigh-Taylor instability develops leading to particles falling faster near the walls (Figure 4a). Particles near the axis of the channel are retarded due to an adverse pressure gradient acting on them (Figure 6). The falling particles further ‘pull’ the particles from the top. This motion eventually leads to the entrainment of fluid in the fluid-particle suspension (Figure 4b). The ‘bubble’ of fluid rises and breaks open at the top of the column leading to the ejection of particles (Figure 4b, Figure 5 and Figure 4c). This also leads to the formation of local particle clusters (Figure 4d and Figure 7b). The particles eventually begin to settle steadily to the bottom of the column to a no-motion state. Figure 7 shows the fluid concentration at different times in the x-y plane and averaged with respect to the z-direction. In the final no-motion state (Figure 7d) the extra weight of the particles is balanced by the interparticle stress.

The present scheme is used to simulate the flow of solid-liquid mixtures in fractured reservoirs. Barree and Conway (1995) reported experimental and numerical study of convective proppant transport. We consider the flow of solid-liquid mixture in a channel or slot that is 7.5 m long (x-direction), 25 cm high (y-direction) and 6.3 mm wide (z-direction). Gravity acts in the negative y-direction. We have 150 control volumes in the x-direction, 10 control volumes in the y-direction and 5 control volumes in the z-direction. The fluid and particle properties and the parameters for interparticle stress...
model are the same as that used in the simulation of the sedimentation column. There are initially no particles in the channel. We simulate two cases. Figure 8 shows the particle concentration at different times in the x-y plane and averaged with respect to the z-direction for the first case. In this case the fluid-particle mixture is introduced uniformly, at a volumetric flow rate of 19.69 cm$^3$/s and particle concentration of 10%, from the top half of the channel at the left end. It is seen that a gravity tongue develops which is identical to the one seen in the experiments of Barree and Conway (1995). The particles continue to fall, impinge on the bottom and flow with an increased rate of lateral transport at the bottom of the channel. This is in agreement with the experimental observation. In the second case the fluid-particle mixture is introduced uniformly at the left end of the channel at a volumetric flow rate of 78.75 cm$^3$/s and 10% particle concentration. Once again we see formation of the gravity tongue that traverses along the channel length (Figure 9). It is known from experiments (Kern, Perkins and Wyant 1959) that, in such flows, particles settle and form a bed at the channel bottom. It is seen in Figure 9 that a bed of particles is beginning to form. In these simulations there were around 200,000 parcels in the computational domain.

We perform inclined sedimentation calculations using our numerical scheme. Acrivos and Herbolzheimer (1979) performed experiments to calculate the sedimentation rates in inclined columns. Experiments were run with the container tilted at different angles. Following Snider et al. (1998) we perform calculations in a two-dimensional domain. Our code for three-dimensional domains is used to perform calculations in two-dimensions. The calculation domain in our simulations have x and y dimensions equal to 5 cm and 60 cm, respectively. There are 32 control volumes in the x-direction and 72 control volumes in the y-direction. Gravity acts at an angle of 35° with the negative y-direction. Suspending fluid properties are the same as bimodal sedimentation. Density of the particles is 2420 kg/m$^3$; their diameters vary uniformly between 130 µm to 142 µm. Initial concentration of the particles is 0.1. Initially, the particles are randomly placed with uniform distribution upto a height of 52.33 cm along the y-axis (the mixture-fluid interface is tilted at an angle to vessel walls). We use Model A for interparticle collision with the same parameters as before. There are 18111 parcels in the calculation domain.
The number of particles in each parcel is chosen so that the total particle volume in each parcel is the same. The time step is same as in bimodal simulations.

Figure 10 compares the transient interface levels of the particles from experiment (Acrivos and Herbolzheimer 1979) and simulation. Figure 11 shows the particle positions at different times.

The mixture-fluid interface can form wave instabilities similar to those of a fluid flowing down an inclined plane. Herbolzheimer (1983) presented photographs of waves at the interface in inclined sedimentation. Snider et al. (1998) simulated wave instability at the interface. We perform the same simulation as Snider et al. (1998) and reproduce the wave observed by them in their simulations. Fluid viscosity is changed to 0.0188 Pa-s, particle diameter is 132 µm and particle density is 2440 kg/m³. Column inclination is 20°. Particles are filled in the column upto a height of 40 cm. All other parameters are the same as the inclined sedimentation simulations above. Figure 12 shows the formation of wave on the mixture-fluid interface similar to that reported by Snider et al. (1998).

Animations of some simulations reported here and some other simulations can be seen at our web site http://www.aem.umn.edu/Solid-Liquid_Flows.

5 Conclusion

In this paper we have described an Eulerian-Lagrangian numerical simulation (LNS) scheme for the flow of particulate mixtures. We have implemented an efficient three-dimensional time dependent finite volume algorithm. A Chorin-type pressure-correction based fractional-step scheme was used to solve the fluid phase equations on a non-staggered cartesian grid. The reported scheme can account for suspending fluids with non-constant properties.

The numerical scheme was tested through convergence tests for the bimodal sedimentation of particles in a vessel. Results on the Rayleigh-Taylor instability of particles sedimenting in a fluid from a height were presented. They were in good agreement with the direct numerical simulation results reported in literature. Calculations were done for inclined sedimentation in a vessel. Wave instability on the fluid-mixture interface during inclined sedimentation was observed. We also simulated the flow of
particulate mixture in a fracture. The calculated results were in good qualitative agreement with the experimental observations.

In summary, we present a numerical scheme that extends the MP-PIC method in the following way (even if the viscous terms in the fluid phase equations are to be neglected): (a) A fractional step algorithm for fluid-particle equations and (b) collocated grid for pressure and velocity.

The numerical scheme is not limited to the particular model used for viscous stress terms in the fluid phase. It has the flexibility to use different models for these terms. The primary objective behind adding a viscous stress term was to provide a numerical framework that has the flexibility to incorporate different models for various applications. The way models are tested is through comparisons with experiments. Every model must pass this test. If the prediction of a model disagrees with experiments then it is not valid. If the predictions agree with few experiments it doesn’t mean that it is valid either. Our model with viscous terms agree qualitatively with some experiments; so we are encouraged to look further.

Acknowledgments

We acknowledge the support from NSF under KDI/NCC grant NSF/CTS-9873236 and STIM-LAB.

References


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Figure Captions

Figure 1. Transient interface levels for bimodal batch sedimentation of particles.
Figure 2. Calculated (Case B) parcel positions at different times for bimodal batch sedimentation of particles.
Figure 3. Comparison of parcel positions in bimodal batch sedimentation for Cases A and B at t = 300s.
Figure 4. Parcel positions at different times depicting the Rayleigh-Taylor instability of particles falling in a sedimentation column.
Figure 5. A three-dimensional view of parcel positions in a sedimentation column.
Figure 6. Contour plot for pressure averaged in the z-direction in a sedimentation column.
Figure 7. Contour plot of fluid volume fraction averaged in the z-direction at different times in a sedimentation column. The particles are fully settled at t = 275s.
Figure 8. Contour plot of particle volume fraction averaged in the z-direction at different times for flow of fluid-particle mixture in a fracture. Particles are introduced uniformly at the top half of the channel entrance to the left.
Figure 9. Contour plot of particle volume fraction averaged in the z-direction at t = 79.2s for flow of fluid-particle mixture in a fracture. Particles are introduced uniformly at the channel entrance to the left.
Figure 10. Transient interface level during inclined sedimentation of particles.
Figure 11. Parcel positions at different times during inclined sedimentation.
Figure 12. Formation of wave at the fluid-mixture interface during inclined sedimentation.
Figure 6
Figure 7

(a) $t = 12.5\text{s}$

(b) $t = 20\text{s}$
Figure 7

(c) $t = 150\text{s}$

(d) $t = 275\text{s}$
Figure 9