Experimental study and transient CFD/DEM simulation in a fluidized bed based on different drag models

Ling Zhou, Lingjie Zhang, Ling Bai, Weidong Shi, Wei Li, Chuan Wang and Ramesh Agarwal

Gas–solid two-phase flow is the main phenomena in the chemical-looping combustion (CLC) fluidized bed system. Drag force generated from relative movement between phases is the main force that hinders the movement of the oxygen carrier particles. It is important to evaluate and understand the limitations and validity range of different drag models. In this paper, based on Discrete Elements Methods (DEM) coupled with Computational Fluid Dynamics (CFD), three kinds of drag models, Wen-Yu, Syamal-O’Brien, and Gidaspow, are used for the simulation of a laboratory-scale spouted fluidized bed. The numerical results based on these three models are compared and analyzed from the views of the bed height, bubble diameter, and pressure fluctuations. A spouted fluidized bed experiment rig was established to carry out the high-speed imaging experiment and validate the numerical results. All three drag models could basically describe the transient behavior of the bubble shape. In the first stage of whole fluidization process, Syamal-O’Brien drag model could predict well both the change of the bubble diameter and the bubble shape in good agreement with experiment results, but the model for the bubble collapse status is not a good prediction; Wen-Yu drag model underestimates the variations in bubble diameter and predicts the occurrence of the next fluidization stage prematurely; although, the Gidaspow drag model still slightly underestimates the change of bubble diameter, but the results of this drag model are in good agreement with the experiment results in terms of the bubble shape, and pressure fluctuations. Under the architecture of CFD/DEM, the Gidaspow drag model gives the better prediction of the inner flow in the dense gas–solid fluidized bed. The results provide reference for the further improved design of the fluidized bed by employing the CFD/DEM method.

1. Introduction

Reducing carbon emissions from fossil-fueled power plants has been an active area of research in recent years. One technology that appears to be very promising for high-efficiency low-cost carbon capture is chemical-looping combustion (CLC). CLC involves combustion of fuels (either gas or solid) by heterogeneous chemical reactions with an oxygen carrier, usually a particulate metal oxide. Gas–solid two-phase flow is the main phenomena in a CLC fluidized bed. It is important and difficult to study the characteristics of the gas–solid two-phase flow for further improved design of the fluidized bed, especially for the condition of high solid concentration.

Over decades, many researchers carried out plenty of laboratory-scale fluidized bed experiments to explore the characteristics of the flow inside the fluidized beds, such as the minimum fluidization velocity, the bed height, and particle aggregation, etc. There are many kinds of classifications of particles, however, the generally accepted method is the one summarized by Geldart, which is based on a huge number of experimental results. Among them, Geldart D particles are those with large diameter and high density. These particles are widely used in experimental studies because they can obviously present an aggregation state during fluidization process. However, restricted by the present experimental level, it is hard to obtain microscopic kinetic information during collision by experiment.

With the development of computer technology and parallel computing, numerical simulation has become the primary means of studying the inner flow characteristics of the fluidized bed. Currently, there are two main methods for the numerical simulation of gas–solid two-phase flows: the first is the Euler-Euler method, also known as the Two-Fluid Model (TFM) which is based on the classical hydrodynamics theory. In this method, each phase is considered as a continuum and specifies the particle amount in the mixture of two phases by the particle
volume fraction, and employs the constitutive equations of the granular theory to calculate the inter-particle forces. This model costs less computer facility and computational time-consuming, it mainly used in high mass loading situations and reaction occasions which focus on the energy exchange during reaction processes, such as the simulation of chemical-looping pilot plant and the simulation of chemical-looping combustion reactor with chemical reactions. The second approach is the Euler–Lagrange method, which regards the fluid phase as a continuum and the solid particles as the discrete phase. Then the continuity equation for the fluid phase and the Newtonian equations for the discrete phase are solved using the force-balance method to obtain the discrete phase trajectories. Compared to the Euler–Lagrangian method, the main disadvantage of Euler–Euler method is that this method does not take into account the properties of the particles, such as Young’s modulus, elastic modulus, particle diameter, etc., and ignores the collision between the particles and the particles-walls. Discrete Elements Methods (DEM) belongs to the deterministic model of the Euler–Lagrangian method, which is able to describe the translational and rotational motion of the particles by calculating the Newtonian equations. It also considers the collisions among the particles and the particles-walls. As a result, this model is widely used for calculating the dense gas–solid two-phase flow of solid phase volume fraction greater than 12%. According to the particles kinetic theory, particles will suspend in the bed when the forces exerted from the fluid phase equal gravity.

In describing the hydrodynamics of gas–solid flow systems, drag force which is exerted by the fluid phase is one of the main contributing factors. Different scholars based on numerous experiment results had established a variety of drag models. Until now, there are mainly three drag models are used to calculate the gas–solid flow in a fluidized bed, which are Syamlal–O’Brien model, Wen-Yu model and Gidaspow model. Owing to the relatively complicated experiment situations and inner flow characteristics, there are still no clear guidance in choosing the appropriate drag model for different experiment situations. Many researches used Two-Fluid Model (TFM) to simulate the gas–solid two-phase flow. Lu et al. developed a new non-spherical particle drag model based on TFM to simulate the supercritical water fluidized bed. These simulation results were validated by comparing with experimental data and the bed expansions agreed well with experimentally measured data. Askaripour et al. conducted the 3D modeling and simulation of bubbling fluidized beds using various drag models based on TFM. They explored a wide range of solid size, static bed height and results of expansion ratio and solids velocity were compared with experimental data. It was found that Wen-Yu drag model provided the best prediction of expansion ratio and solids velocity. Wang et al. proposed a new drag model based on TFM which incorporated the contribution of meso-scale structure effects for Geldart-B particles, the simulation results obtained by the new model showed better agreement with the experimental results than Gidaspow model. However, it is also apparent that these simulation results obtained by TFM contained considerable deviations, so Euler–Lagarange model gets more attention recently. Ayeni et al. analyzed the inadequacies of the traditional drag laws in predicting pressure drop and velocity, and proposed a new drag law based on energy and force balance in fluidized bed which took into account additional energy for heterogeneous structure. And this model enforced the macroscopic conservation principles while attempting to still accommodate local variations in drag due to cluster formation mechanisms.

Until now, the investigations of drag models used in dense gas–solid two phase flow are insufficient, especially for the Euler–Lagrangian method. In this paper, by coupling the CFD and DEM, we investigated three most commonly used drag models to discuss their applicability and defect in dense gas–solid two phase flow. The inner-flow characteristics, such as the bed height, bubble diameter, particle velocity distribution and pressure fluctuations are compared among these different drag models. Furthermore, these prediction results are compared with the high-speed imaging experimental results, which could serve as references for the further application and improvement of drag models.

2. Mathematical modeling

2.1 Mass and momentum balances

DEM can be considered as a kind of soft sphere model, which considers the particle penetration depth and volume fraction of discrete phase and continuous phase in infinitesimal calculating element simultaneously.

Let $\varepsilon_g$ denote the porosity; the motion of the fluid phase obeys the laws of governing mass and momentum conservation equations and that can be written as:

$$\frac{\partial}{\partial t}(\varepsilon_g \rho_g) + \nabla \cdot (\varepsilon_g \rho_g \vec{u}) = 0$$  \hspace{1cm} (1)

$$\frac{\partial}{\partial t}(\varepsilon_g \rho_g \vec{u}_g) + \nabla \cdot (\varepsilon_g \rho_g \vec{u}_g \otimes \vec{u}_g) = -\varepsilon_g \nabla p + \nabla \cdot (\varepsilon_g \tau_g) + \varepsilon_g \rho_g \vec{g} + \vec{F}_{DEM}$$ \hspace{1cm} (2)

where $\rho_g$ is the gas density, $\vec{u}_g$ is the velocity vector, $p$ is the static pressure, $\tau_g$ is the stress tensor, $g$ is the gravitational acceleration, and $F_{DEM}$ is the external force generated by all the moving particles which acts on the infinitesimal fluid element (e.g. the force generated due to interaction of the fluid phase with the discrete phase). Conservation of mass and momentum of fluid phase can also be applied in dispersed phase.

Considering that the existence of a pressure field is a necessary condition, when solving the Navier–Stokes equations of the fluid phase, the pressure field and the velocity field must be coupled via the continuity equation. Thus the SIMPLE algorithm in ANSYS-Fluent 16.0 is employed to ensure this coupling. Furthermore, for high solid concentration fluidized bed, the standard $k-\varepsilon$ turbulence model is employed to simulate the turbulent flow in the reactor; it is satisfactory since the more complex turbulence models such as the full Reynolds Stress Model (RSM) or Large Eddy Simulation (LES) are computationally complex and expensive with limited benefit in accuracy of the simulations.
The motion of particles in a fluidized bed are mainly moving and rotating. The equations of particle movement and rotation are:

\[
m \frac{d\vec{u}_p}{dt} = \sum F_i = F_g + F_{saf} + F_{drag} + F_{con} + F_{con}^{s} + F_{adh} \tag{3}
\]

\[
I_p \frac{d\vec{w}_p}{dt} = T_p = I_p \vec{F}_{con} \tag{4}
\]

where \( m \) is the mass of the particle, \( \vec{u}_p \) is the particle velocity and \( F_i \) is the sum of various forces acting on the particle including gravity, Saffman force, drag force, short range force and the adhesive forces. Specifically, the adhesive forces contain either with or without physical contact. The adhesive forces with physical contact are based on the formation of a physical bond between particles or particles and wall, such as solid bridges. The adhesive forces without physical contact include van der Waals and electrostatic forces.

### 2.2 Discrete element method

There are plenty of forces that can cause particles moving, among which the most important forces are gravity, drag force and short-range forces. In this paper, elastic-damper model is used to calculate the short-range forces during particle collision. In the discrete element method, it is assumed that the particles can overlap each other and penetrate into the wall. Then the particle–particle collision force is calculated by the penetration depth. Similar procedures can be used to calculate the contact force due to the particle-wall collision with the assumptions that the wall has an infinite radius. This model includes Hooke’s elastic model and Newton’s damping model. The elastic model is used for modelling the reversible deformation, while the damping model for irreversible and time-dependent deformation. The torque acting on a particle resulting from the tangential contact force and causing the particle to rotate can be calculated by multiplying the tangential contact force with the particle radius. Fig. 1 shows a brief introduction of the model, and, in the figure, the part A and B are the normal and tangential collision models between particles, part C and D are the normal and tangential collision models between particle and wall.

When evaluating the collisions between parcels, it is too costly to conduct a direct force evaluation that involves all of the parcels. Consider that for \( N \) parcels, the number of pairs that would need to be inspected for every time step would be on the order of \( N^2 \). To address this issue, as shown in Fig. 2, a geometric approach is used: the domain is divided by a suitable Cartesian mesh (where the edge length of the mesh cells is a multiple of the largest parcel diameter), and then the force evaluation is only conducted for parcels that are in neighboring mesh cells, because particles in more remote cells of the collision mesh are a priori known to be out of reach.

What should be mentioned here is that the simulation results will contain a certain error if the particle size is larger than the fluid cell size. Generally, there are two strategies being most commonly used to deal with the problem of particle size. The first method is using relatively large fluid cell size which is 2.5 to 3 times of the particle size. This method is most commonly used in many occasions because it increases the computational velocity. The other strategy takes the inter-phase coupling into consideration to modify calculation resolution. This strategy includes the porous media concept and the two-grid concept. For the porous media concept, it regard the flow field as a porous media. The porous media model can be used for a wide variety of single phase and multiphase problems, including flow through packed beds, filter papers, perforated plates, flow distributors, and tube banks. For the two-grid concept, the physical values of the fluid and solid phases are determined in separated grids, namely the fluid and article grids. The fluid physical variables obtained in a fine fluid grid are transferred to the coarse particle cells to calculate the inter-phase momentum and energy exchange.
Let $K$ be the elastic collision coefficient and $\eta$ ($0 < \eta \leq 1$) be the damper restitution coefficient, thus the formula of the Spring–Dashpot Collision Law are as follows:

$$
\vec{F}^\alpha = (K\delta + \gamma(\vec{v}_{12} \cdot \vec{e}_{12}))\vec{e}_{12}
$$

(5)

$$
\delta = \|X_2 - X_1\| - (r_1 + r_2)
$$

(6)

where $X_1$ and $X_2$ are the locations of particle 1 and 2, $r_1$ and $r_2$ are the radius of 1 and 2, $\delta$ is the penetration depth of particle–particle and particles-wall.

$$
k = \frac{\pi s^2}{3\epsilon d}Dp
$$

(7)

$$
\gamma = -\frac{m_{12} \ln \eta}{t_{coll}}
$$

(8)

$$
t_{coll} = \frac{m_{12}}{K}
$$

(9)

$$
\vec{v}_{12} = \vec{r}_2 - \vec{r}_1
$$

(10)

$$
m_{12} = \frac{m_1 m_2}{m_1 + m_2}
$$

(11)

$$
f_{loss} = \sqrt{\pi \eta^2 + n \eta}
$$

(12)

where $f_{loss}$ is a loss factor, $m_1$ and $m_2$ are the mass of particle 1 and 2; $m_{12}$ is the mass loss in the collision process; $t_{coll}$ is the collision time scale; $\vec{v}_1$ and $\vec{v}_2$ are the velocity of particle 1 and 2; $\vec{v}_{12}$ is the relative velocity; $\gamma$ is the damping coefficient.\(^{15}\)

2.3 Drag model

Momentum exchange between phases is mainly according to the momentum exchange coefficient $K_d$, which is the most different feature between these three models.

2.3.1 Wen-Yu drag model. Wen-Yu drag model is based on a variety of experiments done with various liquid-solid systems.\(^{19}\) And this model is actually the supplement and expansion of Richardson–Zak model, the momentum exchange coefficient of which is:

$$
K_d = \frac{3}{4} C_D \frac{a_1 \rho_1 |\vec{v}_1 - \vec{v}_2|}{d_s} \alpha_1^{-2.65}
$$

(13)

where $C_D$ is drag coefficient:

$$
C_D = \frac{24}{\alpha_1 \Re_a} \left[1 + 0.15(\alpha_1 \Re_a)^{0.687}\right]
$$

(14)

$\Re_a$ is the Reynolds number:

$$
\Re_a = \frac{\rho_d |\vec{v}_1 - \vec{v}_2|}{\mu_1}
$$

(15)

2.3.2 Gidaspw drag model. Gidaspw drag model combines Wen-Yu model and Ergun model, which takes discrete phase proportion into consideration. This model is suitable for particulate systems where viscous forces are dominant, for example dilute flows.\(^{20}\) Let $\alpha_1$ be the porosity of continuous phase. When $\alpha_1$ is greater than 0.8, Wen-Yu is chosen, on the contrary, when $\alpha_1$ is under 0.8 Ergun model is selected. Therefore, the formula of Gidaspw model is as follows:

$$
\alpha_1 > 0.8 K_d = \frac{3}{4} C_D \frac{a_1 \rho_1 |\vec{v}_1 - \vec{v}_2|}{d_s} \alpha_1^{-2.65}
$$

(16)

$$
\alpha_1 < 0.8 K_d = \frac{1.75 \alpha_1 \rho_1 |\vec{v}_1 - \vec{v}_2|}{d_s} + \frac{150 \alpha_1 (1 - \alpha_1) \mu_1}{a_1 d_s^2}
$$

(17)

where $C_D$ is the drag coefficient:

$$
C_D = \frac{24}{\alpha_1 \Re_a} \left[1 + 0.15(\alpha_1 \Re_a)^{0.687}\right]
$$

(18)

As can be seen from these formulas, when $\alpha_1$ is under 0.8, the drag model is unrelated to the Reynolds number. However, when $\alpha_1$ is greater than 0.8, the value of $C_D$ follows the change of Reynolds number.

2.3.3 Syamal-O’Brien drag model. The Syamal-O’Brien drag model is based on the notion that for single- and multi-particle systems the Archimedes number remains the same for terminal setting velocity.\(^{21}\)

The momentum exchange coefficient between phases is:

$$
K_d = \frac{a_1 \rho_f f}{\tau_s}
$$

(19)

$$
\tau_s = \frac{\rho_d d_s^2}{18 \mu_1}
$$

(20)

where $\tau_s$ is particle relaxation time, $d_s$ is particle diameter, $\mu_1$ is viscosity of fluid phase.

$f$ is a function of the drag coefficient $C_D$ and Reynolds number $\Re_a$. The model is based on the terminal velocity $V_{ts}$ of the particle phase in the fluid field, volume fraction $\alpha$ and the corresponding particle phase Reynolds number $\Re_{a}$. The formula is:

$$
f = \frac{C_D \Re_{a} \alpha_1}{24 V_{ts}}
$$

(21)

where

$$
C_D = \left(0.63 + \frac{4.8}{\sqrt{\Re_a / V_{ts}}}\right)^2
$$

(22)

So the momentum exchange coefficient of Syamal-O’Brien drag model is:

$$
K_d = \frac{3}{4} C_D \frac{a_1 \rho_1 |\vec{v}_1 - \vec{v}_2|}{d_s} \alpha_1^{-2.65}
$$

(23)

where particle terminal velocity and Reynolds number are:

$$
V_{ts} = 0.5 \left(A - 0.06 \Re_a + \sqrt{(0.06 \Re_a)^2 + 0.12 \Re_a (2B - A) + A^2}\right)
$$

(24)
where $A = \alpha_{l}^{4.14}$ and $B = 0.8\alpha_{l}^{1.28}$ ($\alpha_{l} \leq 0.85$), $B = \alpha_{l}^{2.65}$ ($\alpha_{l} > 0.85$).

3. Experimental and numerical methods

As shown in Fig. 3, a spouted fluidized bed experimental rig was set up in the Fluid Machinery Laboratory, Jiangsu University. The quasi-3D fluidized bed is made of plexiglass, with the length, width and height of 15 cm, 2 cm, and 100 cm, respectively. The selected solid particles are consisted of 30 000 average diameter of 2.5 mm glass beads. The initial bed height is about 15 cm, which is measured by the ruler embedded in the right side of the bed. The air is supplied through a centrally placed nozzle at the inlet. Inlet low rate of 300 L min$^{-1}$ at room temperature is carried out in the experiment, which is controlled by a high accuracy mass flow controller. Pressure fluctuations in the bed are measured by three pressure sensors located at the height of 2 cm, 22 cm, and 40 cm, respectively. The particle aggregation and dispersion in the fluidized bed are taken in real time by a high speed camera placed in front of the bed.

As shown in Fig. 4, in order to explore the internal mechanism of the flow of the fluidized bed, a 3D solid model was established in full agreement with the experimental geometry. At the bottom of the fluidized bed piled up 30 000 of 2.5 mm diameter spherical particles, programmatically. At first, we conduct a series of air-blowing and free-fall simulation to achieve the randomness of the initial particle packing, which is called the starting point ($t = 0$). For DEM simulation, since the gas phase equation and the momentum exchange coefficient are both based on the volume of the control body, when...
establishing the mesh model, the diameter of the mesh must be larger than the diameter of the particle. On this basis, through the grid independence analysis, determine the total number of grid as 22,880, and the number of grid nodes in the length, width and height are 22, 4 and 260, respectively. Inlet is set as mass flow inlet and outlet is set as pressure outlet. The motion of fluid phase is modeled by using the standard $k-e$ turbulence model, and Spring-Dashpot model is used to solve the collision process between particles and particles-walls. The setting parameters are listed in Table 1.

4. Results and discussions

4.1 Transient bubble shape comparison

As shown in Fig. 5, (a) is the experimental high-speed imaging results, while (b)–(d) are simulation results of these three drag models, respectively. The latter three pictures represent the volume fraction of particle–phase distribution along Z orientation, which all under the same range from 0 to 0.7. At $t = 0$, all these three cases have the same stacking height, about 15 cm. Obviously, there is a slight difference of initial bed height between numerical simulation and experiment, which is mainly due to that $4.1 \times 10^3$ N m$^{-1}$ is chosen as the linear-elastic collision coefficient between particles according to experience when using DEM model for calculation. If this value is bigger, the deformation during collision is smaller, and vice versa. Through the analysis of enormous simulation and experimental results, it can be found that this value is reasonable because, in the DEM model, the selection of the value is consistent with the allowed mutual penetration depth of the model and the calculation accuracy is also considered. During the previous 75 ms period, these calculation results of three drag model are in good agreement with the experimental results.
agreement with experimental results. With the further development of fluidized state, the bed height becomes higher and the bubble area grows, while particles at the top of the bubble moving towards both sides of the bed so that the upper portion of the bed becomes increasingly sparse, where these three drag force models can be well observed. During 100 ms to 225 ms time period, Syamal-O’brien model slightly overestimates the size of the bubble, in contrast, Gidaspow model slightly underestimated the size of the bubble and Wen-Yu model underestimates the size of the cavity. As bubble size is directly related to the drag force generated by the relative motion between phases, the smaller drag force, the smaller bubble diameter is, and vice versa.

Fig. 6 is for the comparison of high-speed imaging experimental results and numerical simulation results from 250 ms to 475 ms time period. With the increasing of the bubble volume, the velocity of upper particles moving to both sides of the walls increases, leading to the particle layer on the top of the bed becomes much sparser and the gap between the particles is gradually increased. At the time when the internal pressure of the bubble is greater than the surface pressure of the bed, the air escape from the gap between particles to the outlet. During a gas overflow process, the particles on the top of the bubble ejected into the free area till the bubble finally burst, showing an extremely unstable fluidized state. The upper thickness of the bubble of all these three drag model are relatively small, while Syamal-O’brien model cannot well describe the bubble rupture process, representing that the performance of the gas within the bubble cannot get through the top layer of the bed showing an slugging fluidized state. Wen-Yu model is too early to predict the flow of the next fluidized stage process, performing that the bed collapses prematurely and also the bubble area is relatively small. But, Gidaspow model can well predict the vacuoles rupture process, bed collapse process and the bubble shape.

4.2 Fluidization characteristics

Fig. 7 shows the comparison of the bed height of all these three different drag models. Obviously, the bed height, along with the time, shows the process of rise and fall, and peaks at around 300 ms. At the stage of nearly completely slump, the initial bed height was significantly higher than the initial bed height, this is because, after the gas entering into the bed, the gas fills in the gaps between the particles, and because of the continuous influx of gas kept at the inlet leading to a lifted state of the
fluidized bed. Within the initial 125 ms, these three kinds of drag model can well describe the changes of the bed height, however, differences began to appear at around 125 ms, and from 125 ms to 400 ms period, all these three kinds of drag model are significantly undervalued the changes of the bed height, which is probably due to the fact that in order to guarantee the initial accumulation state of the bed in simulation is consistent with the experiment as much as possible, there is a subsidence at the middle. This is resulted from, during the simulation process, the underlying particles rise with bubble expanding to compensate for the height difference. Meanwhile, due to the interphase momentum exchange is severe in the expanding section of the bed, as a result, the relatively small force, such as particle non-contact stress, becomes much larger in a short time, while these forces are often ignored in the simulation.

Fig. 8 shows the contrast of the bubble diameter between experimental results and numerical simulation results. The size of the bubble diameter is an important benchmark to measure the performance of the internal flow of a spouting fluidized bed. To quantify the scale size of bubble, the bubble diameter is defined as:

\[ d_{eq} = \sqrt{\frac{4}{\pi} A_b} \]  

(26)

where \( A_b \) represents the area of the bubble which is measured by a professional drawing software.

It is apparent from the figure that the simulation results of the three drag model from 0 to 250 ms time period coincide with the experimental results quite well, and the bubble diameter peaks at 250 ms. With the further development of fluidized state, Wen-Yu models significantly underestimated the changes of the bubble diameter, meanwhile, the model predict the changing time into the next fluidization stage is too earlier. However, the other two models have good prediction about the changes of the bubble diameter.

As shown in Fig. 9, these are the static pressure fluctuations at the monitoring point of \( x = 7.5 \text{ cm}, y = 0 \text{ cm}, z = 2 \text{ cm} \). We could find that all these three models can well predict the pressure fluctuations at that point, showing that the trend of all the pressure fluctuations are consistent and the maximum velocity appears in the vicinity of 150 ms. The bed rises rapidly in this period and interphase momentum exchange is intense. However, with the fluidized state grows further, the pressure
prediction of Syamal-O’Brien model is significantly higher than the other two models and experimental values, Gidaspow model can well describe pressure fluctuations comparatively. Fig. 10 shows the velocity distribution profile at $z = 10$ cm, along the y-axis, of all these three kinds of drag models. It can be seen that the velocity is symmetric distribution along the Y-Z plane. The simulation results of Gidaspow model and Wen-Yu model is similar but significantly lower than the Syamal-O’Brien model. This is probably due to that the calculation of drag force on the Syamal-O’Brien model is based on the terminal velocity of the particles, when the volume fraction is less than 0.85, the calculated terminal velocity of particle phase is relatively low, while the next fluidization stage is ready to happen at this period, so the bed fluctuation is significant, leading to the calculated drag force is much higher.

5. Discussions

Fig. 11 compares the particle velocity predicted by three drag models at $t = 250$ ms. For the above differences in the simulation of a three-dimensional bubbling fluidized among these drag models and experimental results, it can probably be that Syamal-O’Brien model is calculated based on the relative velocity coefficient model and the huge number of experimental data and summed up by Richardson and Zaki, so this model can be used to simulate the relatively dense two-phase flow systems. Meanwhile this model assumes that the Archimedes number of a single particle is the same as that of all the particles in the entire system. The difference between simulation result and experimental result at the initial stage of the fluidization process can be that in this stage forced convection plays a leading role so the Archimedes number of a single particle and particles throughout the flow field is close. But when the bed is in a transport state, free convection also occupies a certain proportion, which means that in this state the Archimedes number of particles is not all the same, so there is some differences between experimental result and simulation result of the bubble ruptured state.

Gidaspow model integrates Wen-Yu model and Ergun model based on the principle that internal forces between particles are negligible, which means that viscous forces dominate the two-phase flow system, so this model is suitable for relatively dilute gas-solid two phase systems. For the difference between simulation results and experimental results it can be that, in this bubbling fluidized bed, a particle diameter is relatively large, about 2.5 mm, thus, compared to the viscous force, the impact force between particles is also the main force. Still, when the gas-phase volume fraction is greater than 0.8, the Wen-Yu model is chosen, and the gas-phase volume fraction is less than 0.8, the Ergun model is chosen. While the bubble under approaches to rupture, the gas phase volume fraction, which is in the vicinity of 0.8, changes significantly, thereby computing model switched more frequently, resulting in some errors with data transmission method and the accuracy is not good enough. In addition, for typical applications, the computational cost of tracking all of the particles individually is prohibitive. Instead, the approach for the discrete element method is similar to that of the DPM, in that like particles are divided into parcels, and then the position of each parcel is determined by tracking a single representative particle. During the fluidization process, the particles are not in the presence of a single particle, but gathering together in the form of particle groups. Considering the number of the particles used in this simulation, the simplification that leading to the underestimate of the agglomeration force is relatively large, so using this model to calculate the overall system makes the particle distribution more uniform and leads to a certain deviations. At the same time, when calculating the interaction forces between particles, there are many physical parameters (such as elastic modulus, coefficient of restitution etc.) of the particle phase is unknown, but is often determined by the experience, which will inevitably result in some errors.

The key difference between these three drag models is the drag coefficient $C_D$. For Wen-Yu and Gidaspow drag model, the former model is the complement and extension of the latter model, and Gidaspow drag model takes low porosity of continuous phase into consideration, so it is more accurate. For Gidaspow drag model and Syamal-O’Brien drag model, it is the different drag coefficient that results in the differences between the two simulation results. Syamal-O’Brien drag model

![3D snaps of particle velocity magnitude (m s$^{-1}$) predicted by three drag models at $t = 250$ ms.](image)

Fig. 11
considers particle terminal setting velocity which based on the assumption that the Archimedes number is the same for the velocity, but it cannot well predict the bubble rupture stage because at this stage free convection plays a significant role which means that the Archimedes number is not the same. Overall, it is more accurate to use the Gidaspow drag model than the Syamal-O’brien drag model to simulate the gas–solid flow in this spouted fluidized bed.

6. Conclusions

This paper using different drag models to carry out the exploration of the application CFD/DEM. Syamal-O’brien model, Gidaspow model and Wen-Yu model were selected for the simulation of gas–solid flow in a spouted fluidized bed. A high-speed imaging experiment rig was established to compare and validate the numerical predicted results. It could be found that these three models can well predict the movement trend of the particle phase in the fluidized bed, but the simulated bed height were significantly underestimated compared to the experiment result. However, in terms of the changes of the bubble diameter, Gidaspow model and Syamal-O’brien model are good predictors of these changes. In general, the simulation results of Gidaspow model are in good agreement with experimental results. As for the small differences, it is mainly because that when the fluidized bed is in a transportation state, free convection plays a certain role, so the Archimedes number of all the particles in the bed is different, however the model assumes that the Archimedes number of a single particle is the same as that of the whole.

However, since many physical parameters are not clear when using the DEM model to calculate the particle phase, there are some differences between experimental results and simulation results in the calculation of particle collision and bubble ruptured state. But it can be found from these data, the error is within 2.5% for the simulation using Gidaspow drag model, which can be considered that Gidaspow drag model is appropriate apply in the simulation of the dense gas–solid fluidized bed. Meanwhile, there remains to be much work to seek or establish an improved drag model to predict the bubble rupture process and the latter process with more precision.

Acknowledgements

This work was supported by the National Natural Science Foundation of China (Grant No. 51609106), Natural Science Foundation of Jiangsu Province (Grant No. BK20150508), Jiangsu Province Universities Natural Sciences Foundation (Grant No. 15KJB570001), and the funding for the Construction of Dominant Disciplines in Colleges and Universities and Universities in Jiangsu (PAPD).

References


