OPERATION AND DESIGN OF A LIQUID FLUIDIZED BED CLASSIFIER FOR POLYDISPERSE SUSPENSIONS OF EQUAL-DENSITY SOLID PARTICLES THROUGH MODELING AND SIMULATION

A. García 1,2*, and G. López3

1Departamento de Ingeniería Metalúrgica, Facultad de Ingeniería y Ciencias Geológicas, Universidad Católica del Norte, Antofagasta, Chile.
2Centro de Investigación Científico Tecnológico para la Minería, CICITEM, Antofagasta, Chile.
3División El Salvador, Codelco, Chile.

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Abstract
For polydisperse suspensions with equal-density solid particles and continuous particle size distribution, design and operation methodologies of a liquid fluidized bed classifier (LFBC) are introduced, both based on a modified version of the generalized clarifier-thickener (GCT) model presented by Bürger, García, Karlsen, y Towers (2008)Computers and Chemical Engineering 32, 1181-1202. The LFBC is a special case of the GCT characterized by an upwards-directed flow of liquid at the lower end of the unit. Moreover, a versatile way to discretize the particle size variable for the numerical solution of this equation is presented. Numerical examples illustrate the performance of the model and the effectiveness of design and operation methodologies.

Keywords: suspension, fluidization, modeling, simulation, classifier, design, operation.

1 Introduction

Mixtures of disperse solid particles of diverse size and/or density in a fluid, called solid-fluid polydisperse suspensions, are encountered in operations as diverse as mineral processing and food industry, where usually, it is important to group together particles of similar sizes or densities, which is called classification. The classification of particles in solid-liquid systems has been the subject of many theoretical and experimental investigations for several decades.
When a polydisperse suspension is subject to sedimentation, particles of different densities and sizes settle at distinct velocities. Consequently, the final sediment consists of several layers of different composition of particles. This form of segregation is known as differential sedimentation, with faster settling species forming the bottom-most layers, and is commonly used to classify particles in industrial processes. For a system consisting of N different sizes, but equal densities, of particles, N zones of settling suspension are formed, with clear liquid above and a sediment layer at the bottom. The lowest zone, just above the early sediment boundary, contains all particle species at their initial concentration, whereas the region immediately above it is devoid of the largest particles. Each successive zone contains one fewer species than the zone below, with the upper zone containing only the smallest particles.

The author and collaborators (Bürger et al., 2008) present a model for continuous separation and classification of polydisperse suspensions, which extends the model of clarifier-thickener (CT) (Berres et al., 2004; Bürger et al., 2004; Diehl, 2006; Zeidan et al., 2004). The feature is singular sinks describing the continuous discharge of products at several points, whose composition will vary during a transient startup procedure. The well-posedness of the resulting model and the convergence of a numerical scheme for N = 1 are proved by Bürger et al. (2006). They therein formulate a model for a generalized clarifier-thickener (GCT) setup, which may include several sinks, can be operated as a fluidization column, and is allowed to have a varying cross-sectional area. They also define a numerical scheme for its simulation.

Several groups of researchers have conducted experiments with separation devices that are special cases of the GCT setup, and proposed mathematical models for them. Nasr-el-Din et al. (1988; 1990; 1999) study columns for the gravity separation and classification of polydisperse suspensions, that have a feed source at a central depth level and, which are tapped near the top and bottom ends. They also present a mathematical model for the steady-state case only. Experimental results for a similar setup are also presented by Spannenberg et al. (1996). Chen et al. (2002a; 2002b) carry out experiments and develop models of a liquid fluidized-bed classifier for steady state (Chen et al., 2002a) and for the transient case (Chen et al., 2002b). A closely related experimental study is that of Mitsutani et al. (2005).

There are many papers about design and operation for separators and classifiers. On design of thickeners with methods based on Kynch’s theory exist the papers by Talmage and Fitch (1955), Hassett (1958; 1968), Moncrieff (1963/64), Wilhelm and Naide (1981), Lev et al. (1986), Waters and Galvin (1991), Yong et al. (1996) and Chancelier et al. (1997), see also the reviews by Concha and Barrientos (1993) and Schubert (1998); based on computational fluid dynamics (CFD) and numerical simulation there are the articles of Kahane et al. (2002), Garrido et al. (2003), Martin (2004) and Burgos and Concha (2005). On design of hydrocyclones with empirical models, there are the works by Castilho and Medronho (2000) and Kraipech et al. (2006); and with CFD there are the articles of Olson and Van Ommen (2004), Slack et al. (2003) and Delgadillo and Rajamani (2005a; 2005b; 2007).

In this paper, we propose a methodology to design a liquid fluidized bed classifier (LFBC) for suspensions with solid particles of equal-density and continuous particle size distribution, and present a methodology of operation of a LFBC. We also modify the model for continuous separation and classification of polydisperse suspensions proposed by Bürger et al. (2008), by considering a hindered-settling factor whose exponent depends on the size particle and a continuous formula for that exponent, among others. Moreover, a versatile way to discretize the particle size variable for the numerical solution of this equation is introduced. We present numerical examples, in part adopting data from the literature.

## 2 Mathematical model of polydisperse suspension sedimentation

Kinematic models are common approximate descriptions for multiphase flows that are essentially one-dimensional, for example in columns and ducts that are aligned with the driving body force. Usually, in these applications one continuous phase (for solid-liquid suspensions, the fluid), and N disperse phases (solid species) are distinguished. We here consider polydisperse suspensions with a finite number N of solid particle species, where particles of species i have mean diameter $d_i$ and density $\rho_i$, and $d_i \neq d_j$ or $\rho_i \neq \rho_j$ for $i \neq j$.

Kinematic models are based on the specification of the velocity of each species relative to that of the fluid as a function of the local concentrations of all species.
2.1 Model equations

We consider a vessel as shown in Fig. 1. We denote by \( \phi := \phi_1 + \cdots + \phi_N \) the total solids concentration. If \( v_f \) is the fluid phase velocity, and \( S(x) \) is the cross-sectional area of the vessel at depth \( x \) (x-axis has the origin at the level of feeding and growing downward), then the one-dimensional continuity equations for the \( N \) solids phases can be written as

\[
S(x) \frac{\partial \phi_i}{\partial t} + \frac{\partial}{\partial x} \left( S(x) \phi_i v_f \right) = 0, \quad i = 1, \ldots, N, \tag{1}
\]

\[
- S(x) \frac{\partial \phi}{\partial t} + \frac{\partial}{\partial x} \left( S(x)(1 - \phi) v_f \right) = 0. \tag{2}
\]

Introducing the volume flow,

\[
Q(x, t) := S(x)(\phi_1 v_1 + \cdots + \phi_N v_N + (1 - \phi) v_f),
\]

we obtain by adding eqs. (1) and (2) the mixture continuity equation \( \partial Q(x, t)/\partial x = 0 \). Since a constitutive equation will be introduced for the solid-fluid relative velocities or slip velocities \( u_i := v_i - v_f \), \( i = 1, \ldots, N \), we use Eq. (3) and \( \partial Q(x, t)/\partial x = 0 \) to rewrite Eq. (1) as

\[
S(x) \frac{\partial \phi_i}{\partial t} + \frac{\partial}{\partial x} \left( Q(x, t) \phi_i + S(x) \phi_i \left[ u_i - \sum_{j=1}^{N} \phi_j \mu_{ij} \right] \right) = 0, \quad i = 1, \ldots, N. \tag{4}
\]

We define the parameters \( \delta_i := d_i/d_1 \) and \( \bar{\rho}_i := \rho_i - \rho_f \) for \( i = 1, \ldots, N \), and \( \mu := g d_1/(18 \mu_f) \), where \( \rho_f \) and \( \mu_f \) are the density and the viscosity of the fluid, respectively, and \( g \) is the acceleration of gravity, in addition, we specify the phase space of physically relevant concentrations as

\[
D_{\text{max}} := \{ (\phi_1, \ldots, \phi_N) : \phi_i \geq 0, \ldots, \phi_N \geq 0, \phi \leq \phi_{\text{max}} \},
\]

where \( 0 < \phi_{\text{max}} \leq 1 \) is the maximal solids concentration.

Within the Masliyah-Lockett-Bassoon (MLB) model (Lockett and Bassoon, 1979; Masliyah, 1979), \( u_i \) is specified as

\[
u_i = u_i(\Phi) = \frac{\mu \delta_i}{1 + 0.15 \Re_{\text{i}}^{0.687}} V_i(\phi)(\bar{\rho}_i - \bar{\rho}_f^2) \Phi
\]

\[ i = 1, \ldots, N. \tag{6}
\]

for \( \Re_i < 1000 \), where \( \bar{\rho} := (\bar{\rho}_1, \ldots, \bar{\rho}_N)^T \) and \( V_i(\phi) \) is a hindered settling factor that takes into account the presence of other particles. This function can for example, be chosen as

\[
V_i(\phi) = \begin{cases} (1 - \phi)^{n_i-2} & \text{for } \phi \in [0, \phi_{\text{max}}], \\ 0 & \text{otherwise}, \end{cases} \tag{7}
\]

for \( n_i > 2 \).
according to Richardson and Zaki (1954), where \( n_i \) is a number specified later.

To ensure that the solution assumes values in \( D_{\phi_{\text{max}}} \), we herein choose \( V_i(\phi) \) as

\[
V_i(\phi) = \begin{cases} 
(1 - \phi)^{n_i - 2} & \text{for } \phi \in [0, \phi_q], \\
(1 - \phi)^{n_i - 2} \frac{\phi_{\text{max}} - \phi}{\phi_{\text{max}} - \phi_q} & \text{for } \phi \in [\phi_q, \phi_{\text{max}}], \\
0 & \text{otherwise,}
\end{cases}
\]

where \( n_i > 2 \),

\[
\text{which continuously goes to zero as } \phi \to \phi_{\text{max}} \text{ and, where } 0 < \phi_q < \phi_{\text{max}} \text{ is a parameter.}
\]

\( \text{Re}_i \) is the particle Reynolds number for species \( i \),

\[
\text{Re}_i := |u_i|(1 - \phi)\frac{d\rho_f}{\mu_f} \tag{9}
\]

The pair of equations (6) and (9) defines \( u_i \) implicitly. To avoid this implicit form and to be consistent with previous work, in particular, with the stability analysis of Basson et al. (2009), we approximate \( \text{Re}_i \) by

\[
\tilde{\text{Re}}_i \approx \tilde{\text{Re}}_i := \mu \delta_i \overline{\rho} \beta (1 - \phi)\frac{d\rho_f}{\mu_f} \tag{10}
\]

where \( \beta > 0 \) is a constant parameter that has to be adjusted, and the exponent \( n_i \) is specified below. Then, we utilize

\[
\begin{align*}
\tilde{u}_i &= \mu \delta_i \overline{V}_i(\phi) \overline{\rho} \tilde{\rho}_i - \overline{\rho}^T \Phi, \\
\overline{\delta}_i &= \delta_i(1 + 0.15 \overline{\text{Re}}_i^{0.687}), & \text{for } \overline{\text{Re}}_i < 10^3, \\
& i = 1, \ldots, N.
\end{align*}
\]

For spherical particles, the exponent \( n_i \) depends on the particle Reynolds number at infinite dilution, \( \text{Re}_{\text{oc},i} \), and may be given by

\[
n_i = \frac{5.1 + 0.27 \text{Re}_{\text{oc},i}^{0.9}}{1.0 + 0.1 \text{Re}_{\text{oc},i}^{0.9}} \text{ for all } \text{Re}_{\text{oc},i}, & i = 1, \ldots, N, \tag{12}
\]

\[
\text{Re}_{\text{oc},i} := \frac{\rho_f \overline{V}_{\text{oc},i} d_i / \mu_f}{\text{is the particle Reynolds number based on the particle settling velocity at infinite dilution, } \overline{V}_{\text{oc},i}, \text{ which we calculate as follows (Kunii and Levenspiel, 1991):}
\]

\[
\begin{align*}
\overline{V}_{\text{oc},i} &= \left( \frac{\mu_f \overline{\rho} \tilde{\rho}_i g}{\rho_f^{1/3} (18/\mu_f)^{1/3} + 0.591/(d_i^{1/3})} \right)^{1/3}, \\
d_i^* &= \frac{d_i \left( \frac{\rho_f \overline{\rho} \tilde{\rho}_i g}{\mu_f} \right)^{1/3}}{1 + \frac{\mu_f \overline{\rho} \tilde{\rho}_i g}{\rho_f^{1/3} (18/\mu_f)^{1/3} + 0.591/(d_i^{1/3})}}, & i = 1, \ldots, N.
\end{align*}
\]

Inserting Eq. (11) into Eq. (4) yields the system of conservation laws

\[
S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} \left( Q(x,t) \Phi + S(x) f^M(\Phi) \right) = 0, \tag{14}
\]

where the components of the vector

\[
f^M(\Phi) := (f^M_1(\Phi), \ldots, f^M_N(\Phi))^T \tag{15}
\]

are the MLB flux functions given by

\[
\begin{align*}
& f^M_i(\Phi) := \mu \phi_i \overline{V}_i(\Phi) \overline{\delta}_i \left( \overline{\rho}_i - \sum_{j=1}^N \overline{\rho}_j \phi_j \right) \\
& \quad - \sum_{k=1}^N \overline{V}_k(\Phi) \overline{\delta}_k \phi_k \left( \overline{\rho}_k - \sum_{j=1}^N \overline{\rho}_j \phi_j \right), & i = 1, \ldots, N.
\end{align*}
\]

For the stability analysis of the model for the slip velocity presented here, the reader may refer to work of Basson et al. (2009).

3 The renewed generalized clarifier-thickener (GCT) model

Bürger et al. (2008) consider a vessel with axisymmetric circular interior cross-sectional area and circular cylindrical outer pipes as shown in Fig. 1. This unit can be operated continuously in two modes, the clarifier-thickener (CT) mode and the fluidization column (FC) mode. In the CT mode, the feed flow is divided into upwards- and downwards-directed bulk flows, and the upper and lower ends of the unit are identified as overflow and underflow levels, respectively, whereas in the FC mode, there is an additional counter-gravity bulk inflow of liquid from \( x = x_g \).

We herein subdivide the unit into four different zones: the overflow zone \( (x < x_f) \), the clarification zone \( (x_L < x < 0) \), the settling zone (in CT mode) or fluidization zone (in FC mode) \( (0 < x < x_g) \), and the underflow zone (in CT mode) or water inflow zone (in FC mode) \( (x > x_g) \). The vessel is continuously fed at depth \( x = 0 \), the feed level, with fresh feed suspension, and it has discharge outlets for products at different depths located above and below the feed point.

3.1 Suspension bulk flows

The suspension is fed at the volume rate \( Q_F(t) \geq 0 \) and, \( Q_O(t) \) and \( Q_U(t) \) are the volume bulk flows at overflow and underflow, respectively, where \( Q_U(t) > 0 \).
and \( Q_U(t) \leq 0 \) in the CT and FC modes, respectively, and \( Q_G(t) \leq 0 \).

Now let us include discharge openings located at \( x_1^R > \ldots > x_n^R > x_L \) and \( 0 < x_1^R < \ldots < x_m^R < x_R \) associated with the respective discharge rates \( Q^R_L(t) \leq 0, \ldots, Q^R_R(t) \leq 0 \) and \( Q^M_L(t) \leq 0, \ldots, Q^M_R(t) \leq 0 \). We can write the bulk flow as

\[
Q(x, t) = \begin{cases} 
Q_O(t) + \sum_{j=1}^{n_L} Q^L_j(t)H(x - x^L_j) & \text{for } x < 0, \\
Q_U(t) - \sum_{j=1}^{n_R} Q^R_j(t)H(x^R_j - x) & \text{for } x > 0.
\end{cases}
\]

\[
H(\xi) := \begin{cases} 
1, & \text{if } \xi \geq 0, \\
0, & \text{if } \xi < 0,
\end{cases}
\]  
(17)

### 3.2 Solids feed and sink terms

As in Bürgert et al. (2008), we assume that for \( x > x_R \) and \( x < x_L \), the cross sectional area shrinks to a very small value, so that these zones actually correspond to transport pipes in which all solids (if any) move with the velocity of the fluid. Consequently, the slip velocities \( u_1, \ldots, u_N \) are "switched off" outside the vessel interior \((x_L, x_R)\) by the discontinuous function

\[
\gamma_1(x) := \begin{cases} 
S(x) & \text{if } x_L < x < x_R, \\
0 & \text{otherwise}.
\end{cases}
\]  
(18)

The next step is to replace Eq. (14) by the system of equations

\[
S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} \left( Q(x, t) \Phi + \gamma_1(x) f^M(\Phi) \right) = 0, 
\]  
(19)

where \( Q(x, t) \) is given by Eq. (17). Next, we consider that at \( x = 0 \), the unit is fed at a volume rate \( Q_f(t) \geq 0 \) with feed suspension that contains solids of species 1 to \( N \) at the volume fractions \( \phi^F_i(t) \) to \( \phi^F_N(t) \). We assume that

\[
\Phi^F(t) := (\phi^F_1(t), ..., \phi^F_N(t))^T \in D_{\Phi_{\text{feed}}} \text{ for } \forall t > 0.
\]  
(20)

The feed mechanism gives rise to an additional singular source term to Eq. (19), so that we now consider the equation

\[
S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} \left( Q(x, t) \Phi + \gamma_1(x) f^M(\Phi) \right) = \delta(x) Q_f(t) \Phi^F(t),
\]  
(21)

where \( \delta(x) \) is the Dirac delta function centered at \( x = 0 \). Using the Heaviside function we may absorb the right-hand side of Eq. (21) into the flux function. Furthermore, we take into account that the sink terms model the discharge of suspension of unknown concentration. This leads to the equation

\[
S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} \left( Q(x, t) \Phi + \gamma_1(x) f^M(\Phi) \right) - H(x) Q_f(t) \Phi^F(t) = \sum_{m=1}^{n_L} H(x - x^M_m) Q^M_L(t) \Phi(x,t),
\]  
(22)

which can be rewritten as

\[
S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} \left( Q(x, t) \Phi + \gamma_1(x) f^M(\Phi) + K(x, t) \Phi \right) - H(x) Q_f(t) \Phi^F(t) = K(x, t) \frac{\partial \Phi}{\partial x},
\]  
(23)

where we define the piecewise constant (with respect to \( x \)) function

\[
K(x, t) := - \sum_{m=1}^{n_L} H(x - x^M_m) Q^M_L(t) - \sum_{l=1}^{n_R} H(x - x^R_l) Q^R_l(t).
\]  
(24)

### 3.3 Final form of the mathematical model

We assume that the control variables \( Q_f(t), Q_U(t) \) and \( Q_G(t) \) as well as the discharge fluxes controlling the sink terms are constant. Then, in view of Eq. (17), and adding the constant vector \(-Q_f \Phi^F\) into the spatial derivative of the left-hand side of Eq. (22), we can rewrite Eq. (22) as

\[
S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} \left( Q(x, t) \Phi + \gamma_1(x) f^M(\Phi) + K(x, t) \Phi \right) - H(x) Q_f(t) \Phi^F(t) - \sum_{i=1}^{n_L} H(x - x_i^L) Q^L_i(t) - \sum_{l=1}^{n_R} H(x - x_l^R) Q^R_l(t),
\]  
(25)

The feed mechanism gives rise to an additional singular source term to Eq. (19), so that we now consider the equation

\[
S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} \left( Q(x, t) \Phi + \gamma_1(x) f^M(\Phi) \right) = \delta(x) Q_f(t) \Phi^F(t),
\]  
(21)

and the vector \( \gamma(x) := (\gamma_1(x), \gamma_2(x)) \), we obtain

\[
g(x, \Phi) = f(\gamma(x), \Phi) := \gamma_1(x) f^M(\Phi) + \gamma_2(x) (\Phi - \Phi^F).
\]  
(25)
This yields the governing equation
\[ S(x) \frac{\partial \Phi}{\partial t} + \frac{\partial}{\partial x} f(\gamma(x), \Phi) = K(x) \frac{\partial \Phi}{\partial x} \]  \tag{26}

This system is solved together with the initial condition
\[ \Phi(x, 0) = \Phi^0(x) := (\phi^0_1(x), \ldots, \phi^0_N(x))^T \in D_{\phi_{\text{max}}}. \]  \tag{27}

4 The liquid fluidized bed classifier

4.1 Preliminaries

In this section, we determine conditions on the cross-sectional area and volume flow rates of a Liquid Fluidized Bed Classifier (LFBC) (see Fig. 2) under which the MLB mathematical model predicts the existence of different compositions inside the unit and of the overflow, underflow and discharge streams, for given volume flow rates \( Q_F \) and \( Q_U \), and concentration vector \( \Phi^F \). We consider suspensions in which the solid species differ in size only (i.e., \( \rho_1 = \rho_2 = \ldots = \rho_N =: \rho_s \)), then Eq. (16) simplifies to the following equation:
\[ j_i^M(\Phi) := \mu(\rho_s - \rho_f) \phi_i (1 - \phi) \left[ \gamma_i(\Phi) \delta_i - \sum_{k=1}^{N} V_k(\Phi) \delta_k \phi_k \right], \]
\[ i = 1, \ldots, N. \]  \tag{28}

4.2 Design of a LFBC

4.2.1. Criterion 1

We choose as first criterion for design of a LFBC that particles of the largest species (species 1) do not leave the column by the underflow, with the purpose of do not block the pipe for the fluidization liquid. Then, in the zone below the lowest sink, the value of the flux of the largest species must be less than or equal to zero, i.e.
\[ S(x) j_1^M(\Phi) + Q_U \phi_1 \leq 0, \quad x \geq x_{nR}, \]  
from which we obtain
\[ S(x) \leq -\frac{Q_U \phi_1}{j_1^M(\Phi)}, \quad x \geq x_{nR}. \]  \tag{30}

Moreover, we may expect that the largest species is the only present in that zone, so the volume fraction vector has the form
\[ \Phi = \hat{\Phi} := (\hat{\phi}_1, 0, \ldots, 0)^T. \]  \tag{29}

4.2.2. Criterion 2

A second criterion for design of a LFBC is that particles of the smallest species (species \( N \)) do not leave the column by the overflow, with the purpose of obtaining a clean liquid. Then, in the zone over the uppermost sink, the value of the flux of the smallest species must be greater than or equal to zero, i.e.
\[ S(x) j_N^M(\Phi) + Q_O \phi_N \geq 0, \quad x \geq x_{nL}, \]  
from which we obtain
\[ S(x) \geq -\frac{Q_O \phi_N}{j_N^M(\Phi)}, \quad x \geq x_{nL}. \]  \tag{31}

Moreover, we may suppose that the smallest species is the only present in that zone, so the volume fraction vector has the form
\[ \Phi = \hat{\Phi} := (0, \ldots, 0, \hat{\phi}_N)^T. \]  \tag{31}
Therefore, for given $Q_O < 0$ and $\phi_N$ such that $0 < \phi_N \leq \phi_{\text{max}}$, the minimum cross-sectional area of the column in the clarification zone is given by

$$S_{\text{min}}^L := -\frac{Q_O}{f_{MN}(\Phi)}. \quad (32)$$

### 4.3 Operation of a LFBC

We denote $x^+$ and $x^-$ as the right and left limits of $x$, respectively. Furthermore, in this section for a general function $G(x, t)$, because $t = t_0$ is given, we simplify the notation in the following way: $G(x^+) := G(x^+, t_0)$, $G(x^-) := G(x^-, t_0)$.

#### 4.3.1. Volume balance for each species in a node with singular source or sink located at $x = \tilde{x}$

This volume balance will be useful for studying the bulk flows and concentrations around singular sources and sinks. For species $i$, $F_i(x, t)$ represents the flux function in $x$-direction and $F_i^S(t)$ is the singular flux term located at $x = \tilde{x}$, then the volume balance for species $i$ in a control volume with center at $x = \tilde{x}$ and thickness $2\delta$ is the following

$$\int_{\tilde{x}-\delta}^{\tilde{x}+\delta} S(x) \frac{\partial \phi_i}{\partial t} dx = -F_i(\tilde{x} + \delta, t) + F_i(\tilde{x} - \delta, t) + F_i^S(t), \quad i = 1, \ldots, N. \quad (33)$$

Let $\delta \to 0$, then the volume balance for each species at $x = \tilde{x}$ results

$$F_i^S(t) = F_i(\tilde{x}^+, t) - F_i(\tilde{x}^-, t), \quad i = 1, \ldots, N.$$

And as $t = t_0$ is given, we simplify the notation of the above equation as follows:

$$F_i^S = F_i(\tilde{x}^+) - F_i(\tilde{x}^-), \quad i = 1, \ldots, N. \quad (34)$$

Then, for the volume balance at $x = 0$ for each species, we apply the Eq. (34) to yield

$$Q_T \phi_i^F = Q_S \phi_i(0^+) + S(0^+) f_i^M(\Phi(0^+)) - Q_L \phi_i(0^-) - S(0^-) f_i^M(\Phi(0^-)), \quad i = 1, \ldots, N. \quad (35)$$

#### 4.3.2. Condition 1: Separation of species 1, \ldots, $m$ from species $m + 1, \ldots, N$ in the feed point at $x = 0$

This condition means that no particles of species 1 to $m$ in $x < 0$, and no particles of species $m + 1$ to $N$ in $x > 0$, or equivalently particles of species 1 to $m$ move downward in $x < 0$, and particles of species $m + 1$ to $N$ move upward in $x > 0$. Then, the following flux relations are valid

$$Q_T \phi_i(0^+) + S(0^+) f_i^M(\Phi(0^+)) > 0, \quad i = 1, \ldots, m,$$

$$Q_S \phi_i(0^+) + S(0^+) f_i^M(\Phi(0^+)) < 0, \quad i = m + 1, \ldots, N,$$

which we replace in Eq. (35) to produce the following relations

$$-Q_T \phi_i^F + Q_S \phi_i(0^+) + S(0^+) f_i^M(\Phi(0^+)) > 0, \quad i = 1, \ldots, m. \quad (36)$$

$$Q_T \phi_i^F + Q_S \phi_i(0^-) + S(0^-) f_i^M(\Phi(0^-)) < 0, \quad i = m + 1, \ldots, N. \quad (37)$$

Because of the numeration of solid particles species of same density, relations (36) and (37) can be reduced to the following ones

$$-Q_T \phi_i^F + Q_S \phi_i(0^+) + S(0^+) f_i^M(\Phi(0^+)) > 0,$$

$$Q_T \phi_i^F + Q_S \phi_i(0^-) + S(0^-) f_i^M(\Phi(0^-)) < 0,$$

from which we obtain the relation for our Condition 1

$$L_{Q_i}(m + 1) < Q_f < U_{Q_i}(m) \quad (38)$$

with

$$L_{Q_i}(m + 1) := -\frac{Q_T \phi_i^F - S(0^+) f_i^M(\Phi(0^+))}{\phi_i(0^-)}, \quad (39)$$

and

$$U_{Q_i}(m) := \frac{Q_T \phi_i^F - S(0^+) f_i^M(\Phi(0^+))}{\phi_i(0^+)} - Q_f \quad (40)$$

#### 4.3.3. Condition 2: Separation of species $m$ from species 1, \ldots, $m - 1$ in the sink point at $x = x_R^1$

This condition means that all particles of species $m$ in $x > 0$ go through the sink at $x = x_R^1$. The generalization of this case to others sink points is simple.

First, we require that particles of species $m$ move downward, i.e.

$$F_m(x_R^1) := Q_R \phi_m(x_R^1) + S(x_R^1) f_m^M(\Phi(x_R^1)) > 0. \quad (41)$$

Second, we need that all particles of species $m$ in $x > 0$ go through the sink, then

$$F_m(x_R^1) = -Q_R^1 \phi_m^R. \quad (42)$$

Finally, because the water flow for fluidization, the volume fraction of species $m$ in the sink is less or equal than that above the sink level, i.e.

$$\phi_m^R \leq \phi_m(x_R^1). \quad (43)$$
The Condition 2 of separation of species is obtained combining the relations (41), (42) and (43), in the following one

$$0 < F_n(x_{k-1}^j) \leq -Q_R \phi_n(x_R^{k-1}).$$  \hfill (44)

For the scalar case, the above relation (44) can be derived from the jump conditions given by Bürger et al. (2006).

5 Numerical scheme

5.1 Discretization of the interior of the GCT

We discretize the spatial domain into cells $I_j := [x_{j-1/2}, x_{j+1/2}), j \in \{0, \pm 1, \pm 2, \ldots\}$, where $x_k = k\Delta x$ for $k \in \{0, \pm 1/2, \pm 1, \pm 3/2, \ldots\}$. Similarly, the time interval $(0, T)$ is discretized via $t_n = n\Delta t$ for $n \in \{0, \ldots, N\}$, where $N = \lceil T/\Delta t \rceil + 1$, which results in the time strips $P_n := [t_n, t_{n+1})$, $n \in \{0, \ldots, N-1\}$. Here $\Delta x > 0$ and $\Delta t > 0$ denote the spatial and temporal discretization parameters, respectively. We set $\Delta x := L/(J+1)$ where $L$ is the height of the column and $J$ is a natural number, and $\Delta t$ is chosen so that the following stability condition (CFL condition) holds:

$$\frac{\Delta t}{\Delta x S_{min}} \left( \max_x (\rho(J_t(\gamma, \Phi))) + \max_x S_{max} \max_{i \in \{0, \ldots, N\}} |v_i^{rj}| \right) \leq \frac{1}{8},$$

where $\rho(\cdot)$ denotes the spectral radius, $J_t(\gamma, \Phi)$ the $N \times N$ Jacobian of $f(\gamma, \Phi)$, and $S_{min} = \min_{x \in (-\infty, \infty)} S(x)$.

In the numerical scheme, we approximate $\max_x \rho(J_t(\gamma, \Phi))$ by

$$\rho := \max_{x \in (-\infty, \infty)} \left| \gamma^1(x) \right| + S_{max} \max_{i \in \{0, \ldots, N\}} \left| v_i^{rj} \right|,$$  \hfill (45)

where $S_{max} = \max_{x \in (-\infty, \infty)} S(x)$, and $v_i^{rj}$ is given by Eq. (13) with $d$ and $\rho_i$ replaced by $d_i$ and $\rho_i$, respectively.

We denote by $G(x^-)$ the limit of a function $G(\xi)$ for $\xi \to x$, $\xi < x$, and introduce the difference operators $\Delta_{-} V_j := V_j - V_{j-1}$ and $\Delta_{+} V_j := V_{j+1} - V_j$.

Our scheme is a direct modification of the one described by Kurganov and Tadmor (2000). Let $U^n_j := (U^n_{j-1}, \ldots, U^n_{j+1})^T$ denote our approximation to $\Phi(x_j, t_n)$. Expressed in terms of the forward Euler solver, we consider the one-parameter family of Runge-Kutta schemes

$$U^{(1)}_{j} = U_j^n - \lambda_j \Delta \mathbf h(y_{j+1/2}, U^n_{j+1}, \ldots, U^n_{j+2}) + \lambda_j K_j \Delta \mathbf h(U^n_{j+1}, \ldots, U^n_{j+2}),$$

$$U^{(s+1)}_{j} = (1 - \eta_k) \left( U^{(s)}_{j} - \lambda_j \Delta \mathbf h(y_{j+1/2}, U^n_{j+1}, \ldots, U^n_{j+2}) + \lambda_j K_j \Delta \mathbf h(U^n_{j+1}, \ldots, U^n_{j+2}) \right) + \eta_k U_j^n, \quad k = 1, 2, \ldots, s - 1,$$

$$U^{(s)}_{j} := U_j^n,$$  \hfill (46)

where $y_{j+1/2} := y(x^j_{j+1/2})$, $\lambda_j := \Delta t/(S \Delta x)$ with $S := S(x^j_{j+1/2})$, $K_j := K(x^j_{j+1/2})$, and $U_j^n := U_0^n := 0$.

We employ second-order time differencing $(s = 2)$, for which $\eta_1 = 1/2$; for third-order time differencing $(s = 3)$, the appropriate values are $\eta_1 = 3/4$ and $\eta_2 = 1/3$.

The numerical flux vector $\mathbf h$ appearing in Eq. (46) is given by

$$\mathbf h(y_{j+1/2}, U^n_{j-1}, \ldots, U^n_{j+2}) := \frac{1}{2} \left[ f(y_{j+1/2}, U^n_{j+1/2}(t_n)) + f(y_{j+1/2}, U^n_{j+1/2}(t_n)) \right]$$

$$- \frac{1}{2} \partial_y \frac{\Delta x}{\Delta t} \left[ U^n_{j+1/2}(t_n) - U^n_{j+1/2}(t_n) \right],$$  \hfill (47)

which is expressed in terms of the intermediate values

$$U^n_{j+1/2}(t_n) := U^n_{j+1} - \Delta x \left( \Phi_x \right)^{j+1/2},$$

$$U^n_{j+1/2}(t_n) := U^n_{j} - \Delta x \left( \Phi_x \right)^{j},$$  \hfill (48)

and the local speeds of propagation $\partial_y \Phi_x^{j+1/2}$, which we estimate by

$$\partial_y \Phi_x^{j+1/2} = \gamma^1(x^j_{j+1/2}) \max \left| \left| v_i^{rj} \right| \right|, \quad i \in \{0, \ldots, N\}.$$  \hfill (49)

The numerical derivatives are determined by

$$\left( \Phi_x \right)^{j} := \frac{1}{\Delta x} \text{MM} \left( \partial \Phi^n_{j-1} - \partial \Phi^n_{j} \right),$$

$$\partial \left( U^n_{j+1} - U^n_{j} \right),$$  \hfill (50)

where $\theta \in [1, 2]$ is a parameter and $\text{MM}(\cdot, \cdot)$ is the minmod function:

$$\text{MM}(a, b, c) := \begin{cases} \min \{a, b, c\} & \text{if } a, b, c > 0, \\ \max \{a, b, c\} & \text{if } a, b, c < 0, \\ 0 & \text{otherwise.} \end{cases}$$  \hfill (51)

As stated by Kurganov and Tadmor (2000), in the scalar case ($N = 1$) the value $\theta = 2$ corresponds to the least dissipative limiter, while $\theta = 1$ ensures the non-oscillatory nature of the approximate solution. The
best choice of \( \theta \) depends on the model considered. For systems, the optimal values of \( \theta \) vary between 1.1 and 1.5 (Kurganov and Tadmor, 2000). As a compromise, and following previous works (Berres et al., 2004; Qian et al., 2005), we choose \( \theta = 1.3 \) in all examples.

For the justification of the numerical scheme the reader may refer to work of Bürger et al. (2008).

6 Discretization of a suspension with CPSD

6.1 Reduced size

Definition 1. Let \( d \) be the particle diameter, \( d_{\text{max}} \) be the diameter of the largest particle, and \( k > 0 \) a parameter. We define the Reduced Size as

\[
\xi = \left( \frac{d}{d_{\text{max}}} \right)^k, \quad d \in [0, d_{\text{max}}].
\]

(52)

In function of \( \xi \), the Rosin-Rammler particle size distribution is written as

\[
F_{\text{RR}}(\xi) := 1 - \exp \left( - \left[ \frac{d_{\text{max}}}{l} \xi^{l/k} \right]^m \right), \quad \xi \in [0, 1]
\]

where \( l \) is a characteristic size and \( m \) is a uniformity coefficient.

Definition 2 (Normalized Rosin-Rammler). Since \( F_{\text{RR}}(1) < 1 \), we define the Normalized Rosin-Rammler particle size distribution as

\[
F_{\text{RR}n}(\xi) := F_{\text{RR}}(\xi)/F_{\text{RR}}(1), \quad \xi \in [0, 1]
\]

(53)

6.2 Discretization of a CPSD

We discretize the Reduced Size by defining

\[
\xi_{i+1/2} := (N - i)\Delta \xi,
\]

(54)

for \( i = 0, \ldots, N \), where \( N \in \mathbb{N} \) and \( \Delta \xi := 1/N \).

Definition 3 (Species in a CPSD). We call “Species \( i^n \) for \( i = 1, \ldots, N \), the solid particles with sizes between \( \xi_{i+1/2} \) and \( \xi_{i-1/2} \).

Herein, we assign to species \( i \) the mean reduced size

\[
\xi_i = \frac{1}{2}(\xi_{i+1/2} + \xi_{i-1/2})
\]

If \( \phi^f(t) \) denotes the total solids volume fraction of a feed suspension, the feed volume fraction of the species \( i \) is given by

\[
\phi_i^f(t) = \int_{\xi_{i-1/2}}^{\xi_{i+1/2}} \phi^f(t) F_{\text{RR}n}(\xi) d\xi
\]

\[
= \phi^f(t) \left[ F_{\text{RR}n}(\xi_{i-1/2}) - F_{\text{RR}n}(\xi_{i+1/2}) \right],
\]

\[
t > 0, \quad i = 1, \ldots, N.
\]

Remark 1. Our definition of the Reduced Size is very useful. For example, when Gates-Gaudin-Schumann CPSD, \( F_{\text{GGS}}(d) := (d/d_{\text{max}})^m, \ d \in [0, d_{\text{max}}], m > 0 \) is used and if we choose \( k = 1/m \), then \( \phi_i^f(t) = \phi^f(t)/N \). Moreover, if \( k = 1 \), then \( d_i = d_{\text{max}}\xi_i \), for \( i = 1, \ldots, N \).

Remark 2. Herein we use the arithmetic mean to determine the mean reduced size of each species, but it is possible to improve the calculation of that, considering the particle size distribution inside each species. Of course, while the number of species be greater, the difference between both means will be less.

Table 1. Physical parameters for the model fit.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Quantity</th>
</tr>
</thead>
<tbody>
<tr>
<td>( N )</td>
<td>2</td>
</tr>
<tr>
<td>( d_1 \ [\text{m}] )</td>
<td>( 9.0 \times 10^{-4} )</td>
</tr>
<tr>
<td>( d_2 \ [\text{m}] )</td>
<td>( 5.5 \times 10^{-4} )</td>
</tr>
<tr>
<td>( \rho_1 \ [\text{kg/m}^3] )</td>
<td>2470</td>
</tr>
<tr>
<td>( \rho_2 \ [\text{kg/m}^3] )</td>
<td>2470</td>
</tr>
<tr>
<td>( n_1 )</td>
<td>2.96</td>
</tr>
<tr>
<td>( n_2 )</td>
<td>3.24</td>
</tr>
<tr>
<td>( \rho_f \ [\text{kg/m}^3] )</td>
<td>998.2</td>
</tr>
<tr>
<td>( \mu_f \ [\text{kg/m}^3] )</td>
<td>( 1.005 \times 10^{-3} )</td>
</tr>
</tbody>
</table>

7 Numerical examples

7.1 Example 1: Model fit

We here adopt experimental data from the work of Chen et al. (2002a) for the steady-state separation of a bidisperse suspension in a liquid fluidized bed classifier. The vessel, Fig. 2, corresponds to equipment “T-2” of Chen et al. (2002a), and is described by its interior cross-sectional area

\[
S(x) := \begin{cases} 
 4.54 \times 10^{-3} \text{ m}^2 & \text{for } x \leq -0.165 \text{ m}, \\
 0.0287 \text{ m}^2 & \text{for } -0.165 < x < 0.915 \text{ m}, \\
 S_1(x) \text{ m}^2 & \text{for } 0.915 < x < 1.709 \text{ m}, \\
 2.04 \times 10^{-3} \text{ m}^2 & \text{for } x > 1.709 \text{ m}, 
\end{cases}
\]
including a conical segment defined by

\[ S_1(x) := \frac{1}{0.794^2} \left[ 1.709 \sqrt{0.0287} - 0.915 \sqrt{2.04 \times 10^{-3}} \right. \\
+ \left( \sqrt{2.04 \times 10^{-3}} - \sqrt{0.0287} \right) x \right] \]

The solids parameters correspond to glass beads of two sizes. For this suspension, we use Eq. (8) with \( \phi_q = 0.63 \) and \( \phi_{\text{max}} = 0.68 \), and use Eq. (10) with \( \beta = 0.19 \).

The physical and operation parameters are given in tables 1 and 2, respectively.

In this example we record an approximate \( L^1 \) error defined with respect to a reference solution, to evaluate the performance of the scheme. We introduce a \( L^1 \) error, denoted by \( e_1 \), which is defined by

\[ e_1 := \Delta x \sum_{i=M_L}^{M_R} \sum_{j=1}^{m} \sum_{k=1}^{N} | \tilde{U}^n_{k,i} - U^n_{k,i} |, \]

where \( \tilde{U}^n_{k,i} \) and \( U^n_{k,i} \) are the reference solution at \( x = x_i \) and the approximate solution at \( x = x_i \), respectively, both for species \( k \) at \( t = t_n \); \( m \) is the value of the division between \( \Delta x \) of the approximate solution and that of the reference solution; \( M_L \) and \( M_R \) are the indices of the positions between which we calculate the errors of the numerical approximation; and \( \Delta x \) is the spatial discretization parameter of the reference solution. The reference solution was calculated with the discretization parameters \( \Delta x = 3.470 \times 10^{-3} \) m and \( \Delta t = 7.352 \times 10^{-3} \) s.

Table 2. Operation parameters for the model fit.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Quantity</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \phi^F )</td>
<td>0.0676</td>
</tr>
<tr>
<td>( \phi^R )</td>
<td>0.0624</td>
</tr>
<tr>
<td>( Q_F ) [m³/s]</td>
<td>5.960 \times 10^{-3}</td>
</tr>
<tr>
<td>( Q_R ) [m³/s]</td>
<td>-1.444 \times 10^{-3}</td>
</tr>
<tr>
<td>( Q_L ) [m³/s]</td>
<td>-7.404 \times 10^{-3}</td>
</tr>
<tr>
<td>( Q_R^1 ) [m³/s]</td>
<td>-3.668 \times 10^{-4}</td>
</tr>
</tbody>
</table>

Table 3. Physical data and feed volumetric flow for numerical examples 2 to 5.

<table>
<thead>
<tr>
<th>Parameter</th>
<th>Quantity</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \rho_s ) [kg/m³]</td>
<td>2470</td>
</tr>
<tr>
<td>( \rho_f ) [kg/m³]</td>
<td>998.2</td>
</tr>
<tr>
<td>( \mu_f ) [kg/m³]</td>
<td>1.005 \times 10^{-3}</td>
</tr>
<tr>
<td>( Q_F ) [m³/s]</td>
<td>5.0 \times 10^{-4}</td>
</tr>
</tbody>
</table>

Fig. 3 indicates that the model fits reasonably well the experimental data reported in Fig. 3 by Chen et al. (2002a) that have been obtained by sampling. In figs. 4 and 5 we observe that the numerical...
scheme converges to the reference solution.

Data for next examples

In all next examples, the vessels are similar to that of the model fit (Fig. 2), i.e., it has one only sink located under the feed point. On the other hand, the fluid is water at 20 °C, the solid is a chalcopyrite concentrate with continuous particle size distribution with Rosin-Rammler parameters \( d_{\text{max}} = 1.13 \times 10^{-3} \) [m], \( m = 0.7254 \) and \( l = 8.0495 \), and the reduced size parameter \( k = 0.5 \).

The common physical and operational data for the examples are given in Table 3.

For Examples 2, 3 and 4 the set of particles with continuous size distribution is divided in 5 species. The calculated parameters for them are given in Table 4.

For Example 5 the set of particles with continuous size distribution is divided in 10 species. The calculated parameters for it are given in Table 5.

The operational parameters for numerical examples 2 to 5 are given in Table 6.

<table>
<thead>
<tr>
<th>Species</th>
<th>( d_i ) [m]</th>
<th>( \phi_i )</th>
<th>( n_i )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>( 9.15 \times 10^{-4} )</td>
<td>( 5.53 \times 10^{-2} )</td>
<td>3.12</td>
</tr>
<tr>
<td>2</td>
<td>( 5.54 \times 10^{-4} )</td>
<td>( 4.94 \times 10^{-2} )</td>
<td>3.56</td>
</tr>
<tr>
<td>3</td>
<td>( 2.83 \times 10^{-4} )</td>
<td>( 4.24 \times 10^{-2} )</td>
<td>4.40</td>
</tr>
<tr>
<td>4</td>
<td>( 1.02 \times 10^{-4} )</td>
<td>( 3.36 \times 10^{-2} )</td>
<td>5.03</td>
</tr>
<tr>
<td>5</td>
<td>( 1.13 \times 10^{-5} )</td>
<td>( 1.94 \times 10^{-2} )</td>
<td>5.10</td>
</tr>
</tbody>
</table>

Table 5. Calculated parameters for Example 5.

<table>
<thead>
<tr>
<th>Species</th>
<th>( d_i ) [m]</th>
<th>( \phi_i )</th>
<th>( n_i )</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>( 1.02 \times 10^{-3} )</td>
<td>( 2.83 \times 10^{-2} )</td>
<td>2.92</td>
</tr>
<tr>
<td>2</td>
<td>( 8.16 \times 10^{-4} )</td>
<td>( 2.69 \times 10^{-2} )</td>
<td>3.00</td>
</tr>
<tr>
<td>3</td>
<td>( 6.36 \times 10^{-4} )</td>
<td>( 2.55 \times 10^{-2} )</td>
<td>3.14</td>
</tr>
<tr>
<td>4</td>
<td>( 4.77 \times 10^{-4} )</td>
<td>( 2.39 \times 10^{-2} )</td>
<td>3.36</td>
</tr>
<tr>
<td>5</td>
<td>( 3.42 \times 10^{-4} )</td>
<td>( 2.22 \times 10^{-2} )</td>
<td>3.73</td>
</tr>
<tr>
<td>6</td>
<td>( 2.29 \times 10^{-4} )</td>
<td>( 2.02 \times 10^{-2} )</td>
<td>4.25</td>
</tr>
<tr>
<td>7</td>
<td>( 1.38 \times 10^{-4} )</td>
<td>( 1.81 \times 10^{-2} )</td>
<td>4.76</td>
</tr>
<tr>
<td>8</td>
<td>( 7.06 \times 10^{-5} )</td>
<td>( 1.55 \times 10^{-2} )</td>
<td>5.03</td>
</tr>
<tr>
<td>9</td>
<td>( 2.54 \times 10^{-5} )</td>
<td>( 1.23 \times 10^{-2} )</td>
<td>5.10</td>
</tr>
<tr>
<td>10</td>
<td>( 2.82 \times 10^{-6} )</td>
<td>( 7.09 \times 10^{-3} )</td>
<td>5.10</td>
</tr>
</tbody>
</table>

Table 6. Operational data for numerical examples 2 to 5.

<table>
<thead>
<tr>
<th>Example</th>
<th>( Q_e ) [m³/s]</th>
<th>( Q_u ) [m³/s]</th>
</tr>
</thead>
<tbody>
<tr>
<td>2</td>
<td>( -5.0 \times 10^{-4} )</td>
<td>( -1.0 \times 10^{-3} )</td>
</tr>
<tr>
<td>3</td>
<td>( 4.9 \times 10^{-4} )</td>
<td>( -1.0 \times 10^{-5} )</td>
</tr>
<tr>
<td>4</td>
<td>( -3.0 \times 10^{-3} )</td>
<td>( -3.5 \times 10^{-3} )</td>
</tr>
<tr>
<td>5</td>
<td>( -3.0 \times 10^{-3} )</td>
<td>( -3.5 \times 10^{-3} )</td>
</tr>
</tbody>
</table>

7.2 Example 2: Design of a classifier according to Criterion 1

In this example, the criterion for designing a classifier is that the largest particles must not leave the column by the underflow. The vessel is described by

\[
S(x) := \begin{cases} 
4.54 \times 10^{-3} \text{ m}^2 & \text{for } x \leq -0.165 \text{ m}, \\
S^R_{\text{max}} & \text{for } -0.165 \text{ m} < x \leq 0.915 \text{ m}, \\
2.04 \times 10^{-3} \text{ m}^2 & \text{for } x > 1.709 \text{ m}, \\
2.04 \times 10^{-3} \text{ m}^2 & \text{for } x > 1.709 \text{ m}, \\
\end{cases}
\]

including a conical segment defined by

\[
S^2(x) := \frac{1}{0.794^2} \left[ 1.709 \sqrt{S^R_{\text{max}}} - 0.915 \sqrt{2.04 \times 10^{-3}} \\
+ \sqrt{2.04 \times 10^{-3}} \right] \left[ \sqrt{2.04 \times 10^{-3}} \right]^2,
\]

The expected volume fraction of species 1 in the zone below the sink is \( \phi_1 = 0.03 \). Then, according to Eq. (30), the maximum cross-sectional area in the zone below the sink is \( S^R_{\text{max}} = 1.851 \times 10^{-2} \) [m²].

Figs. 6 and 7 show the simulated volume fractions until steady state is reached of species 1 and 3 and, species 5 and total, respectively. Fig. 8 shows the volume fractions of each species and total, versus \( x \) in steady state. Fig. 8 shows that the species 1, which is the largest, not output from the underflow and reaches a volume fraction equal to 0.03, which is the expected concentration. Furthermore, it is seen that 1 is the only species present in the area below the sink.
Fig. 6. Example 2: Simulated volume fractions (a) $\phi_1$ (Species 1), (b) $\phi_3$ (Species 3).

Fig. 7. Example 2: Simulated volume fractions (a) $\phi_5$ (Species 5), (b) $\phi$ (Total solids).

Fig. 8. Example 2: Simulated volume fractions at steady state.
7.3 Example 3: Design of a classifier according to Criterion 2

In this example, the criterion for designing a classifier is that the smallest particles must not leave the column by the overflow. The vessel is described by

\[
S(x) := \begin{cases} 
4.54 \times 10^{-3} \text{ m}^2 & \text{for } x \leq -0.165 \text{ m}, \\
S_\text{L min} & \text{for } -0.165 \text{ m} < x \leq 0.915 \text{ m}, \\
S_3(x) & \text{for } 0.915 \text{ m} < x \leq 1.709 \text{ m}, \\
1.836 \times 10^{-2} \text{ m}^2 & \text{for } x > 1.709 \text{ m}, 
\end{cases}
\]

including a conical segment defined by

\[
S_3(x) := \frac{1}{0.794^2} \left[ 1.709 \sqrt{S^L_{\text{min}}} - 0.915 \sqrt{1.836 \times 10^{-2}} \\
+ \left( \sqrt{1.836 \times 10^{-2}} - \sqrt{S^L_{\text{min}}} \right) x \right]^2.
\]

The expected volume fraction of species N in the zone above the uppermost sink is \( \phi_N = 0.04 \). Then, according to Eq. (32), the minimum cross-sectional area in that zone is \( S^L_{\text{min}} = 0.1208 \text{ [m}^2\text{]} \).

Figs. 9 and 10 show the simulated volume fractions until steady state is reached of species 1 and 3, and species 5 and total, respectively. Fig. 11 shows the volume fractions of all species and total, versus \( x \) in steady state. Fig. 11 shows that the species 5, which is the smallest, not output from the overflow. Furthermore, it is seen that 5 is the only species present in the area above the feeder.

7.4 Example 4: Operation of a classifier enforcing Condition 1

In this example, the condition for operation is that no particles of species 1 to \( m \) in \( x < 0 \), and no particles of species \( m+1 \) to \( N \) in \( x > 0 \). The vessel is described by

\[
S(x) := \begin{cases} 
4.54 \times 10^{-3} \text{ m}^2 & \text{for } x \leq -0.165 \text{ m}, \\
0.0574 \text{ m}^2 & \text{for } -0.165 \text{ m} < x \leq 0.915 \text{ m}, \\
S_1(x) & \text{for } 0.915 \text{ m} < x \leq 1.709 \text{ m}, \\
2.04 \times 10^{-3} \text{ m}^2 & \text{for } x > 1.709 \text{ m}, 
\end{cases}
\]

Fig. 9. Example 3: Simulated volume fractions (a) \( \phi_1 \) (Species 1), (b) \( \phi_3 \) (Species 3).

Fig. 10. Example 3: Simulated volume fractions (a) \( \phi_5 \) (Species 5), (b) \( \phi \) (Total solids).

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Figs. 11 Example 3: Simulated volume fractions at steady state.

Figs. 12 and 13 show the simulated volume fractions of species 2 and 3, and total, respectively, until steady state is reached. Fig. 14 shows the volume fractions of each species and total, versus $x$ in steady state. Fig. 14 shows that species 2 and 3 are separated into the feeder. The species 2 that is larger is directed downward, while the species 3 is smaller than species 2 is directed upwards.

In Table 7, the values of the lower and upper bounds of the relation (38) at time $t = 200$ (s), when the system is in steady state, are given. Table 7 confirms what is observed in Figs. 12 and 14, in the sense that species 2 and 3 are separated in the feed point, as for the species 2 is satisfied the relation (38).

Table 7. Example 4: Values of the lower and upper bounds of the relation (38) in steady state ($t = 200$ (s)). Note that species 2 satisfies the relation (38) because $Q_L = -3.5 \times 10^{-3}$ [m$^3$/s].

<table>
<thead>
<tr>
<th>Species</th>
<th>$L_{Q_L}(i + 1)$</th>
<th>$U_{Q_L}(i)$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>$-3.5023 \times 10^{-3}$</td>
<td>-3.6975</td>
</tr>
<tr>
<td>2</td>
<td>$-3.5013 \times 10^{-3}$</td>
<td>$-3.4997 \times 10^{-3}$</td>
</tr>
<tr>
<td>3</td>
<td>0.1487</td>
<td>$-3.5003 \times 10^{-3}$</td>
</tr>
<tr>
<td>4</td>
<td>28.3176</td>
<td>$-3.5002 \times 10^{-3}$</td>
</tr>
<tr>
<td>5</td>
<td>60.1320</td>
<td>$\infty$</td>
</tr>
</tbody>
</table>

Fig. 12 Example 4: Simulated volume fractions (a) $\phi_2$ (Species 2), (b) $\phi_3$ (Species 3).
7.5 Example 5: Operation of a classifier enforcing Condition 2

Here, the condition for operation is that all particles of species \( m \) in \( x > 0 \) go through the sink at \( x = x_1/R \). The vessel is the same of the Example 4.

Figs. 15 and 16 show the simulated volume fractions of species 2 and 3, and 5 and total, respectively. Fig. 17 shows the volume fractions of each species and total, versus \( x \) near steady state. Fig. 17 shows that in steady state the species 3 does not lower the level of the sink located at \( x_1/R \) and leaves the unit for it. Species 1 and 2 which are larger, lower-level sink at \( x_1/R \).

![Fig. 13. Example 4: Simulated total volume fraction of solids.](image1)

![Fig. 14. Example 4: Simulated volume fractions in steady state (\( t = 200 \) s).](image2)

![Fig. 15. Example 5: Simulated volume fractions (a) \( \phi_2 \) (Species 2), (b) \( \phi_3 \) (Species 3).](image3)
In Table 8, the values of the species flux and upper bound of the relation (44) at time $t_{ss} = 100$ (s), when the system is near steady state, are given. Table 8 shows that not only species 3 and 4 do not cross the level of the sink, as shown in Fig. 17, but so does the species 5, as for these three species satisfies the relation (44).

![Fig. 16. Example 5: Simulated volume fractions (a) $\phi_5$ (Species 5), (b) $\phi$ (Total solids).](image1)

![Fig. 17. Example 5: Simulated volume fractions in steady state.](image2)
Table 8. Example 5: Values of the species flux and upper bound of the relation (44) near steady state ($t_{ss} = 100$ (s)). Note that species 3, 4 and 5 satisfy this relation.

<table>
<thead>
<tr>
<th>Species</th>
<th>$F_i(x_R^{1-}, t_{ss})$</th>
<th>$-Q_R^{1-} \Phi_i(x_R^{1-})$</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>$1.4151 \times 10^{-5}$</td>
<td>$1.0490 \times 10^{-5}$</td>
</tr>
<tr>
<td>2</td>
<td>$1.3457 \times 10^{-5}$</td>
<td>$1.1475 \times 10^{-5}$</td>
</tr>
<tr>
<td>3</td>
<td>$1.2711 \times 10^{-5}$</td>
<td>$1.4055 \times 10^{-5}$</td>
</tr>
<tr>
<td>4</td>
<td>$1.1865 \times 10^{-5}$</td>
<td>$2.2408 \times 10^{-5}$</td>
</tr>
<tr>
<td>5</td>
<td>$5.816 \times 10^{-15}$</td>
<td>$7.468 \times 10^{-14}$</td>
</tr>
<tr>
<td>6</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>7</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>8</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>9</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>10</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>

Conclusions

The contribution of this work is summarized as follows:

- Progress in the model of the generalized clarifier-thickener, presented by Bürger et al. (2008), primarily through the adoption of a hindered settling function for each kind of solid particles.

- Proposition of a method to discretize the variables particle size and volume fraction of species, of a suspension with continuous particle size distribution.

- Presentation of a methodology for designing a liquid fluidized bed classifier (LFBC), in the sense of calculating cross-sectional areas as operational constraints of the equipment, i.e. the non-blocking with solid particles of the pipe that feeds water for fluidization and the collection of clear water by the upper duct.

- Development of a methodology of operation of a LFBC, in the sense of handling the control variables such as volumetric flow at the entrances and exits of the unit, to obtain the desired products.

In the work of Bürger et al. (2008), for all species of particles one hindered settling function $V(\phi)$ are considered, specifically the same exponent in this function, which is calculated as the arithmetic average of the exponents calculated for each species of particles. This assumption we believe is improvable, as in steady state, in the equipment zones of different composition of particles are produced according to their size and density, for example in the case of a suspension of particles of the same density, the lower zone of the equipment is occupied by larger particles, and the upper zone, by smaller particles, so we believe that every species must have its own hindered settling function. Other changes to the model presented by Bürger et al. (2008) are the elimination of the discontinuity in $Re_i = 0.1$ for the formula of the solid-fluid relative velocity $u_i$, the change in the formula for calculating the exponent of the hindered settling function, from the formula of Richardson and Zaki (1954), which is discontinuous, to the formula of Garside and Al-Dibouni (1977), which is continuous, and the relocation of the adjustable parameter in the formula for $Re_i$, so as to increase their range of validity. This work could be useful not only for the design and operation of a LFBC, but also for all equipment whose operation can be modeled with the equations presented here, such as sedimentation of non-flocculated suspensions.

Nomenclature

$d_i$ diameter of species $i$, m
$d_{max}$ diameter of the largest particle species, m
$D_\phi$ phase space of physically relevant concentrations defined by Eq. (5)
$f_i^M(\Phi)$ MLB flux function for species $i$ defined by Eqs. (16) or (28), m/s
$g$ acceleration of gravity, m/s$^2$
$g(x, \Phi)$ vector of flux functions defined by Eq. (25), m$^3$/s
$h$ numerical flux vector defined by Eq. (47), m$^3$/s
$h$ numerical flux vector defined by Eq. (47), m$^3$/s
$H$ Heaviside function defined in Eq. (17)
$J$ number of discretization intervals of space
$K(x)$ piecewise constant function defined by Eq. (23), m$^3$/s
$L$ classifier height, m
$l$ characteristic size of the distribution function of Rosin-Rammler, m
$L_{QL}$ lower limit of condition 1 for operation of a LFBC defined by Eq. (39), m$^3$/s
$m$ exponent of the distribution function of Rosin-Rammler
$MM$ minmod function defined in Eq. (51)

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Greek symbols

$\alpha$ approximation of the maximum spectral radius of the Jacobian of $f(y, \Phi)$ defined by Eq. (45), m/s

$\beta$ parameter adjustment model solid-fluid relative velocity

$\gamma^1$ factor in the vector of flux functions $g(x, \Phi)$ defined by Eq. (18), m$^2$

$\gamma^2$ factor in the vector of flux functions $g(x, \Phi)$ defined by Eq. (24), m$^2$/s

$\gamma_{j+1/2}$ variable defined as $\gamma_{j+1/2} := \gamma(x_{j+1/2})$

$\delta_i$ parameter defined as $d_i/d_1$

$\eta_1, \eta_2$ coefficients in the numerical scheme

$\theta$ weight factor in the minmod function

$\lambda_j$ variable defined as $\lambda_j := \Delta t/(S_j \Delta x)$, s/m$^3$

$\mu_f$ fluid viscosity, Pa·s

$\phi$ total solid volume fraction

$\phi_i$ volume fraction of species $i$

$\Phi$ vector of volume fractions of species

$\Phi_{max}$ maximum solid volume fraction

$\Phi_{g}$ parameter of the hindered settling function

$\phi^i_j$ volume fraction of species $i$ in the feed

$\Phi^F$ vector of volume fractions of species in the feed

$\phi_{tot}$ total volume fraction of solids in the feed

$\phi^{i}_0$ volume fraction of species $i$ at the initial time

$\Phi^{o}$ vector of volume fractions of species at the initial time

$\Phi^1$ vector of volume fractions for criterion 1 of LFBC design defined by Eq. (29)

$\Phi^2$ vector of volume fractions for criterion 2 of LFBC design defined by Eq. (31)

$\rho_f$ fluid density, kg/m$^3$

$\rho_s$ solid density, kg/m$^3$

$\rho_i$ density of species $i$, kg/m$^3$

$\bar{\rho}_i$ relative density of species $i$ defined as $\rho_i - \rho_f$, kg/m$^3$

References


