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Application of the compartmental model to the gas-liquid precipitation of CO$_2$-Ca(OH)$_2$ aqueous system in a stirred tank.

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A compartmental model is formulated to assess the influence of fluid dynamics on the gas-liquid precipitation of CO$_2$(g)-Ca(OH)$_2$(aq) system in a stirred tank reactor. The model combines the description of the flow field with several sub-models, namely gas to liquid mass transfer, chemical reaction, precipitation and population balance for both gas bubbles and solid crystals. The modeling predictions, including the average volumetric mass transfer coefficient, the concentration of calcium ions, the pH of the solution and the Sauter mean diameter of the final crystal products are eventually compared with measurements carried out on a pilot-scale stirred tank. The results show that the local volumetric mass transfer rate and the final particle sizes distribution of the crystals are significantly affected by high local turbulence near the impeller. The local information simulated by the compartmental model, such as mass transfer rate, gas hold up and particle size of crystals and bubbles are important for the design and scaling of gas-liquid precipitators, with a computational time which is of several orders of magnitude faster than a full CFD computation.

Keywords: compartmental model; gas-liquid precipitation; mass transfer; population balance

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Introduction

Gas-liquid precipitation is a widely used process in various fields such as CO₂ capture and storage, fine chemicals and gas cleaning process.¹ The ultimate process development objective is to predict and control the mass transfer rate and the size distribution of the crystal products. Such goal can be reached only if a complex model that combines different portions of the physics, namely gas-liquid mass transfer, chemical reaction, precipitation and population balance equations (PBEs) for bubbles and crystals is considered. In addition, the fluid mixing becomes crucial during the scaling up of precipitator: in fact, the non-uniform turbulent intensity locally affects the gas-liquid mass transfer rates by changing the mass transfer coefficient and specific area of the bubbles from point to point in the domain. Moreover, the precipitation kinetics including nucleation and crystal growth are in turn function of the local supersaturation, driven by the local concentration of the different chemical species.² At last, the model should properly consider that the agglomeration and breakage rate of the crystals are significantly influenced by the local values of turbulent energy dissipation in the stirred tank reactor.

For these reasons, the knowledge of the flow fields inside the equipment represents a crucial step in the model formulation. Computational fluid dynamics (CFD) is a useful tool to provide an insight into the flow fields that govern the overall precipitation rate and the crystal properties in the liquid solution.³-⁵ Wei and Garside⁴ for the first time coupled CFD description of the turbulent flow field with precipitation modeling for the description of a liquid phase precipitation. After this pioneering work, the CFD-PBE became a standard tool to predict the performances of industrial-scale crystallizers.⁶-⁸ Regarding the gas-liquid precipitation, Rigopoulos and Jones⁵ coupled a very simplified multiphase CFD model of a bubble column (i.e., assuming a monodisperse bubble distribution, without considering bubble coalescence and breakage) with a reduced PBE model for crystals with only nucleation and growth. These assumptions
may be not reasonable for stirred tank reactors, as the intensive breakage of bubbles and crystals in the vicinity of impeller will increase the gas-liquid mass transfer area and decrease the average crystal size. However, considering the development of a robust and comprehensive model that couples the multiphase flow dynamics with a detailed gas-liquid precipitation model, including mass transfer, chemical reaction, PBEs for bubbles and crystals may require a significant amount of time. It may be useful to reduce the computational costs of the CFD simulation by adopting a coarse-grained description of the fluid dynamics by means of a compartmental model. This latter approach can be beneficial to a full model development in the next step, which is the fine-tuning fitting of the numerous constants contained in the different adopted sub-models. With this approach, CFD is only employed to obtain the hydrodynamic information, such as the flow fields, physical properties and local mixing intensity for each defined compartment. Then the physical sub-models can be solved by considering the interactions between the hydrodynamics and detailed chemical phenomena in the compartmental model.

In previous works, the compartmental model was already applied to the description of the precipitation process. Zauner and Jones used the compartmental model to describe the nucleation rate and the particle size in a liquid phase precipitator. Kagoshima and Mann developed a networks-of-zones fluid mixing model to successfully predict the size distribution of BaSO crystals in the unbaffled stirred vessel. Giery reviewed several simulation methods for CaCO precipitation in different types of reactors. As far as the coarse-grained description of fluid dynamics in gas-liquid systems is concerned, Laakkonen et al. described the local mass transfer rate by using the compartmental model in an agitated vessel. Nauha and Alopaeus combined the fluid dynamics and algal growth model in a bubble column photobioreactor by means of compartmental model. Regarding the modeling of the gas-liquid chemical adsorption process, Rigopoulos and Jones employed a hybrid CFD framework to investigate the chemical absorption of CO bubbles into alkali solution in the bubble column. Although all these aspects
were separately investigated in many previous works, they were never brought in a single simulation tool for the simulation of a gas-liquid precipitation process. The combination of compartmental model and full gas-liquid precipitation model represents the main element of novelty of this work. With this tool, the rates of all the relevant phenomena, including gas-liquid mass transfer, chemical reaction, bubble coalescence and breakage, crystals nucleation, growth, agglomeration and breakage can be properly taken into account, together with a proper description of the turbulence intensity and the relevant mean-flow fields.

**Modelling approach**

As previously mentioned, a gas-liquid CFD calculation is necessary for creating the coarse-grained compartmental model. Then the flow dynamics is coupled with the sub-models accounting for the physical description of the $\text{CO}_2(g)$-$\text{Ca(OH)}_2(aq)$ precipitation process. The conceptual flowchart of the whole model is shown in Figure 1.

![Figure 1. Flowchart of the compartmental model of gas-liquid precipitation](image)

**Assumptions**
The main model assumptions are:

1) A single bubble size of 4 mm is assumed in the CFD calculation to obtain the initial flow field of gas, since the bubbles ranging between 1-8 mm have approximately the same terminal velocity.\textsuperscript{17}

2) Only buoyancy, gravity and drag forces are taken into account, since they are the most important factors for determining gas holdup and bubble size distribution in stirred tank reactors.\textsuperscript{18}

3) The crystals are assumed to follow the liquid flow because the Stokes number of the crystals (with size less than 100 μm) is generally smaller than 0.1.\textsuperscript{19} The interaction between the crystals and bubbles is neglected.

4) The nucleation and crystal growth rate are expressed as power-law functions of the supersaturation in the bulk solution. The crystal growth rate is taken to be independent of crystal size.

The CFD model and compartmentalization

A gas-liquid flow field is calculated in a stirred tank with Rushton turbine impeller. A sketch of the gas-liquid precipitator is reported in Figure 2(a).\textsuperscript{20} The computational domain consisted of 366644 hexahedral grids. Grid independent study showed no significant variations on the properties of interest after further refinements. To calculate the CFD flow field, a steady-state two-fluid simulation was carried out with the commercial code of Fluent 16.0. The RANS approach, namely the realizable $k$-$\varepsilon$ turbulence model for the mixture with standard wall functions, was adopted to predict the turbulence of the stirred tank, since it represents the only feasible way to simulate pilot-industrial scale turbulent gas-liquid systems.\textsuperscript{21} The degassing boundary condition was used to model the gas flowing out at the top surface of the tank.

The turbulent intensity near the impeller can be of several orders of magnitude higher than other regions in the stirred tank. Values of the mass transfer coefficient ($k_L$), coalescence and breakage rates for the gas bubbles, aggregation and breakage rates for the crystals, depend on the local turbulent energy...
dissipation ($\varepsilon$). Therefore, the compartmentalization strategy uses the spatial distribution of energy dissipation as a reference to subdivide the domain into different areas: feeding area, stirring area, baffle area, suspension area and degassing area. An example of such division is reported in Figure 2(b). All the physical information are averaged while the flow to and from the compartments are calculated with the cell by cell algorithm.\textsuperscript{22} The number/size independence study was carried out by testing the full gas-liquid precipitation model with an increasing number of compartments from 6 to 85. The results showed that a model with 58 compartments (Figure 2(c)) is adequate to perform a number/size independent simulation.

![Figure 2. The schematic diagram of the geometry and compartmentalization of the stirred tank](image)

**Compartmental model**

As previously mentioned, the overall mass balance and population balance of the gas phase and solid phase are coupled with several sub-models to describe mass transfer, reaction and precipitation in the compartmental model. The solution of the PBEs of bubbles is based on the high order moment method of classes (HMMC), which offers a good compromise between computational costs and accuracy.\textsuperscript{23} Therefore, the size range of bubbles (1-20 mm) was discretized into 20 categories in this study. In order to reduce the computational load, quadrature method of moment (QMOM) was introduced to solve the PBE of crystals.\textsuperscript{24} The first six moments of the crystal distributions, from $m_0$ to $m_5$, were tracked, leading
to the calculation of three quadrature nodes. Spatial convective terms of both population balances were calculated by using the averaged information of flow in and out of the compartments.

**Mass balance**

The mass balance equation for each chemical specie of the liquid phase, including CO$_2$(l), HCO$_3^-$, CO$_3^{2-}$, OH$^-$ and Ca$^{2+}$ in compartment $i$ is calculated as follow:

$$
\frac{dn_{L,i}}{dt} = c_{L, in} F_{L, in} - c_{L, out} F_{L, out} + \sum_{j=1}^{NB} c_{L,j} F_{L,j} - c_{L,i} \sum_{j=1}^{NB} F_{L,j} + N_{GL,i} A_{GL,i} V_{L,i} + r_i V_{L,i}
$$

Terms on the right hand side are: 1) liquid feed, 2) liquid outlet, 3) internal liquid flow in, 4) internal liquid flow out, 5) mass transfer rate and 6) reaction rate. The mass transfer rate is just calculated for CO$_2$(l). The time-average internal liquid flows were brought from previous CFD calculation and assumed to be constant in the compartmental model. As expected, the liquid flow is mainly driven by the movement of impeller in the stirred tank.

The mass balance equation for each chemical specie of gas phase in compartment $i$ is calculated as follow (CO$_2$(g) is the only specie of gas phase in this case):

$$
\frac{dn_{G,i}}{dt} = c_{G, in} F_{G, in} - c_{G, out} A_{surface} \sum_{p=1}^{NC} v_p U_{p, slip-out} Y_{p, out}
$$

$$
+ \sum_{j=1}^{NB} \left[ c_{G,j} A_{ij} \sum_{p=1}^{NC} v_p U_{p, j} Y_{p, j} \right] - c_{G,i} \sum_{i=1}^{NB} \left[ A_{ij} \phi_i \sum_{p=1}^{NC} v_p U_{p, j} Y_{p, i} \right] - N_{GL,i} a_{GL,i} V_{L,i}
$$

Terms on the right hand side are: 1) Gas feed, 2) Gas outlet, 3) internal gas flow in, 4) internal gas flow out and 5) mass transfer rate. The area between the compartment $i$ and $j$, $A_{ij}$, is the total area of the liquid flow and gas flow. Therefore, the gas hold up in the compartment, $\phi_G$, should be taken into account when the gas flow rates between compartments are calculated. The initial gas velocity, $U_{p, ji}$, is assumed to be the same as in CFD calculation with a single bubble size. The velocities of bubbles with different size
categories are modified by incorporating the different slip velocities. Tomiyama drag force model was implemented to solve the force balance and calculate the slip velocities for the bubbles belonging to each size category.\textsuperscript{25}

Population balance

The discrete version of PBE for the bubbles can be written as:

\[
\frac{\partial Y_{p,i}}{\partial t} = \frac{1}{V_{\text{disp},in}} F_{\text{disp},in} Y_{p,\text{in}} - \frac{1}{V_{\text{disp},out}} U_{n,\text{slip,out}} A_{\text{surface}} Y_{p,\text{out}} + \frac{1}{V_{\text{disp},j}} \sum_{j=1}^{NB} U_{p,j} A_j \phi_j Y_{p,j} - \frac{Y_{p,i}}{V_{\text{disp},j}} \sum_{i=1}^{NB} U_{p,i} A_j \phi_j \\
+ G_{p,i} + B_{p,i}^{cl} + B_{p,i}^{br} - D_{p,i}^{cl} - D_{p,i}^{br}
\]  

Terms on the right hand side are: 1) Gas feed, 2) Gas outlet, 3) internal gas flow in, 4) internal gas flow out, 5) growth due to the mass transfer and varying pressure, 6) birth due to coalescence, 7) birth due to breakage, 8) death due to coalescence and 9) death due to breakage. It is worth mentioning that bubbles with different size rise with different velocities. In the compartmental model, the velocity of the bubble is calculated by summing up the liquid velocity and the bubble slip velocity referring to a single category. The slip velocities of all size categories are obtained by solving the force balances of bubbles with Newton-Raphson iteration. Then the gas flow rate of each size category is revised by updating the gas velocities. In addition, the influence of the hydrostatic pressure and mass transfer on the growth rates of bubbles are taken into account.\textsuperscript{26}

After applying the moment transformation, the PBE for the crystals can be written as:

\[
\frac{\partial m_{k,i}}{\partial t} = \frac{1}{V_{\text{disp},in}} m_{k,in} F_{\text{disp},in} - \frac{1}{V_{\text{disp},out}} m_{k,out} F_{\text{disp},out} + \frac{1}{V_{\text{disp},j}} \sum_{j=1}^{NB} m_{k,j} F_{\text{disp},j} - \frac{1}{V_{\text{disp},i}} m_{k,i} \sum_{j=1}^{NB} F_{\text{disp},ij} \\
+ G_{k,i} + N_{k,i} + B_{k,i}^{ag} + B_{k,i}^{br} - D_{k,i}^{ag} - D_{k,i}^{br}
\]  

The numbers of spin categories and the number of bubble categories are not equal, so the change in the number of bubble categories is equal to the change in the number of crystal categories.
Terms on the right hand side are: 1) crystals feed, 2) products outlet, 3) internal crystals flow in, 4) internal crystals flow out, 5) growth term, 6) nucleation term, 7) birth due to agglomeration, 8) birth due to breakage, 9) death due to agglomeration and 10) death due to breakage. The Wheeler algorithm is used to calculate the quadrature approximation from the set of the transported moments.

**Table 1. Initial process parameters and physical properties of system**

<table>
<thead>
<tr>
<th>Initial parameters</th>
<th>Values</th>
</tr>
</thead>
<tbody>
<tr>
<td>Initial concentration of Ca(OH)$_2$(aq)</td>
<td>20.32 mol/m$^3$</td>
</tr>
<tr>
<td>Volume of the dispersion</td>
<td>2.64x10$^{-3}$ m$^3$</td>
</tr>
<tr>
<td>Concentration of CO$_2$(g) in the feed gas</td>
<td>100 volume %</td>
</tr>
<tr>
<td>Gas feeding rate</td>
<td>1.67x10$^{-5}$ m$^3$/s</td>
</tr>
<tr>
<td>Impeller speed</td>
<td>58.64 rad/s</td>
</tr>
<tr>
<td>Temperature</td>
<td>25$\degree$C</td>
</tr>
<tr>
<td>Diffusion coefficient of CO$_2$(g)</td>
<td>1.91x10$^{-9}$ m$^2$/s</td>
</tr>
<tr>
<td>Dynamic viscosity of liquid</td>
<td>8.90x10$^{-4}$ Pa·s</td>
</tr>
<tr>
<td>Density of mixture</td>
<td>997.22 kg/m$^3$</td>
</tr>
</tbody>
</table>

**Physical sub-models**

The gas-liquid precipitation of the Ca(OH)$_2$(aq)-CO$_2$(g) system consists of mass transfer of CO$_2$(g), the reaction between the hydroxyl ion and CO$_2$(l), and the precipitation of CaCO$_3$. The initial parameters are listed in Table 1. The gas-liquid mass transfer coefficient in each compartment can be correlated by the following equation, based on the surface renewal theory:\(^{28}\)

$$k_{L,j} = \frac{2}{\sqrt{\pi}} \sqrt{D_{CO_2} \left( \frac{e_j \rho_L}{\mu_L} \right)}$$  \hspace{1cm} (5)
The specific mass transfer area is obtained from the BSD calculated by population balance model. The bubble coalescence and breakage mechanisms are assumed to be caused by turbulent fluctuations, and standard closures for air-water systems in the stirred tank were adopted.\textsuperscript{13} The initial bubble size was assumed the same as in the CFD calculation. The feed bubble size was set to be 3\,mm with a standard deviation of 1\,mm, as estimated by the correlation of Kazakis et al.\textsuperscript{29} for porous gas sparger in the experiment. The reaction model for hydroxyl ion with CO\textsubscript{2}(l) developed by Cents et al.\textsuperscript{30} was adopted. The calcium ions along with the carbonate anions produced by the reaction provide the supersaturation for the precipitation.

In the precipitation model, nucleation, crystal growth, agglomeration of crystals and breakage of agglomerates were taken into account. The power-law functions are commonly used to calculate the nucleation rate as well as the growth rate:

\begin{align}
N & = k_n \left( \sqrt{Ca^{2+}_{} \cdot CO_3^{2-}} - \sqrt{K_{sp,CaCO_3}} \right)^3 \\
G & = k_g \left( \sqrt{Ca^{2+}_{} \cdot CO_3^{2-}} - \sqrt{K_{sp,CaCO_3}} \right)^4
\end{align}

The nucleation constant could be influenced by several parameters, such as energy dissipation, suspension density, temperature, surface tension and viscosity of liquid. However, the detailed description of the nucleation mechanism is out of the scope of this work. Therefore, the precipitation kinetics were taken from the literatures.\textsuperscript{31,32} Outcome of the precipitation process, namely the particle size distribution, can be significantly influenced by the occurrence of agglomeration of crystals and breakage of agglomerates, which are caused by the turbulence fluctuations. The agglomeration rate consists of collision rate and collision efficiency.\textsuperscript{33}
\[ R_{\text{ag}}(L_i, L_j) = \left( \frac{2k_BT}{3\mu} \frac{(L_i + L_j)^2}{L_i L_j} \right) + 1.29 \left( \frac{\varepsilon}{V} \right)^{1/2} (L_i + L_j)^{3/2} \psi' \]  

(8)

The first term inside the brackets accounts for the Brownian collisions, while the second the turbulence-induced collisions. The collision efficiency \( \psi \) was estimated by using the model proposed by Hounslow et al.\textsuperscript{34}

\[ \psi' = \frac{(\Omega(\varepsilon)/\Omega_{50})^{\Gamma}}{1 + (\Omega(\varepsilon)/\Omega_{50})^{\Gamma}} \]  

(9)

The strength of aggregate, \( \Omega(\varepsilon) \), is a function of the energy dissipation, crystal size, apparent yield stress, density of particle, dynamic viscosity of liquid and the growth rate of crystal. \( M_{50} \) and \( \Gamma \) are parameters fitted against experimental data.

The particle breakage is assumed to be caused by two mechanisms: normal stresses acting on the surface of the particle and disrupting stresses acting on a particle trapped between two eddies.\textsuperscript{27} Different mechanisms prevail depending on the ratio between the particle size and the size of smallest turbulence eddy. The breakage is likely to occur by means of the normal stresses when the size of particle is larger than the size of the eddy. For particles of sizes smaller than the turbulent microscale, the breakage is likely to be caused by the shear stress originating from the turbulent velocities of different eddies acting on the opposite sides of the particle. The most often used agglomerate breakage rate model is based on a formulation, where the breakage rate is assumed to be a power law function of kinematic viscosity, turbulence dissipation, and agglomerate size.\textsuperscript{33,35} Due to dimensional considerations, the powers in this model are not independent. Based on dimensional consistency, the breakage law can be re-organized in terms of well-known quantities, namely the ratio between the diameter of the agglomerate (\( L \)) and the Kolmogorov length scale (\( \eta \)), and Kolmogorov time scale (\( \eta_T \)). With these quantities, the breakage rate can be written as follows:
\[ R_{br} = c_{br} \left( \frac{L}{\eta} \right)^\gamma \eta_r^{-1} \]

(10)

where \( c_{br} \) is a dimensionless empirical constant as well as the exponent \( \gamma \). This exponent is the same as reported earlier\(^3\) for power law models for agglomerate size dependency, and the pre-exponential factor is also the same. This formulation of the well-established power law breakage rate is automatically dimensionally correct, and also reveals underlying dependencies between turbulent micro-scales, agglomerate size, and agglomerate breakage rate. By comparing the experimental and modeling results, \( 1.5 \times 10^{-4} \) and 1 were chosen for \( c_{br} \) and \( \gamma \) respectively, which are in line with values given by other researchers in the literature.\(^{36}\) The fragmentation was assumed to be uniform as a first approximation.

**Results and discussion**

**CFD and compartmentalization**

The quality of the compartmentalization can be assessed through the comparison of the flow field calculated with the CFD and that given by the compartmental model. The most important results are the distribution of energy dissipation (\( \varepsilon \)) and gas hold up (\( \varphi_G \)). The results of the compartmental model with only flow field and population balance of bubbles at steady state are shown in Figure 3, together with CFD calculations. The comparison shows that compartmentalization with 58 compartments can preserve the characteristic of the flow fields, since both the local and maximum value of \( \varepsilon \) and \( \varphi_G \) are in the same range of CFD calculation.
Figure 3. Energy dissipation ($\varepsilon$), W/kg and gas hold up ($\varphi_G$): (a) $\varepsilon$ in CFD calculation; (b) $\varepsilon$ in compartmental calculation; (c) $\varphi_G$ in CFD calculation; (d) $\varphi_G$ in compartmental calculation

It is important to mention that the average $\varepsilon$ is 0.26 W/kg in the CFD calculation, however, according to a well-known correlation, the $\varepsilon$ under the concerned operating condition should be near 0.4 W/kg. The reason of this mismatch stems on the $k$-$\varepsilon$ turbulent model, or more in general on RANS models, which are known to underestimate the total turbulent energy dissipation. In this case, a possible workaround is to scale the values of $\varepsilon$ in the compartments to match the value of $\varepsilon$, as different sub-models depend on local value of turbulent dissipation rate. Therefore the torque moment calculated by CFD was introduced to scale the turbulent dissipation values in the compartment model:

$$\varepsilon_L = \frac{2\pi N_s \left| M \right|}{\rho_i V_L}$$

(11)

It is also important to compare the global gas hold-up between compartmental model and CFD: the compartmental model predicted a value of 1.06% in contrast to 0.97% of the CFD calculation. This slight difference could be caused by the calculation of PBEs of bubbles. In fact, the initial bubble diameter in the preliminary CFD simulation is assumed to be 4 mm, while the value of the bubble diameter can change during the compartmental model simulation due to the solution of the PBEs, leading to a size distribution of bubbles with an average diameter of 3.96 mm.
Experimental validation

Measurements of the steady state volume average $k_{L_a}$ on the investigated system were already carried out, while the dynamic concentration of calcium, [Ca$^{2+}$], and the pH of the solution under different operating conditions were performed in this work. [Ca$^{2+}$] of the sample was measured by ion chromatography (ICS-1100 from Thermo Scientific Inc.), while a pH meter (Metrohm 744) was used to monitor in situ the pH of the solution during the gas-liquid precipitation.

The steady state volume average $k_{L_a}$ calculated by the compartmental model under various operating conditions are compared with experimental results in Table 2. As one may note, the agreement is very good under different operating conditions, meaning that the compartmentalization strategy and set of sub-models are adequate to describe the system.

<table>
<thead>
<tr>
<th>$N$, rad/s</th>
<th>$Q$, m$^3$/s</th>
<th>$k_{L_a}$ (Experiment), s$^{-1}$</th>
<th>$k_{L_a}$ (Modeling), s$^{-1}$</th>
</tr>
</thead>
<tbody>
<tr>
<td>58.64</td>
<td>1.67×10$^{-5}$</td>
<td>0.0196</td>
<td>0.0189</td>
</tr>
<tr>
<td>68.06</td>
<td>1.67×10$^{-5}$</td>
<td>0.0304</td>
<td>0.0289</td>
</tr>
<tr>
<td>78.53</td>
<td>1.67×10$^{-5}$</td>
<td>0.0452</td>
<td>0.0427</td>
</tr>
<tr>
<td>58.64</td>
<td>8.35×10$^{-5}$</td>
<td>0.0345</td>
<td>0.0333</td>
</tr>
</tbody>
</table>

As the flow field in a stirred tank is significantly affected by the rotation of impeller, [Ca$^{2+}$] and pH under different impeller rates were simulated and compared with the experimental data. The results of such comparison are reported in Fig 4. As it is possible to observe, the higher impeller speed increases the precipitation rate: during the nucleation period (namely the first few seconds of the process), [Ca$^{2+}$] remains constant because the critical size of the crystal nucleus is assumed to be 0.01μm, which is too
small to have obvious contribution to the amount of CaCO$_3$. After approximately 90s instead, when the concentration reaches the lowest value, [Ca$^{2+}$] tends to increase slightly again since further decreasing of the pH caused by gas feeding may lead to the dissolution of crystals. The dissolution model has not been considered because the crystal dissolution is out of the scope of this work. The influence of impeller rate on the evolution of pH is similar with that on the [Ca$^{2+}$] in Fig 4(b). pH of the solution is determined by the gas-liquid mass transfer rates and the consequent reaction between CO$_2$(l) and OH$^-$. The higher impeller speed leads to the increase of average energy dissipation in the stirred tank, enhancing the mass transfer rates. Moreover, the pH begins to drop dramatically when the precipitation stops. As a key parameter dominating the main species including CO$_2$(l), HCO$_3^-$ and CO$_3^{2-}$ in the CO$_2$(g)-H$_2$O(l) equilibrium system, pH can be used to estimate and monitor the overall precipitation time.

The influence of the impeller rate on the mean Sauter diameter of the crystals is reported in Table 3. As can be clearly seen, $d_{32}$ of crystals decreases when the impeller rate increases from 36.65 rad/s to 58.64 rad/s, as one may expect due to the prevailing of particle breakage over agglomeration at higher impeller rates.

Figure 4. The revolution of [Ca$^{2+}$] and pH. Case1: $N= 36.65$ rad/s; $Q= 1.67 \times 10^{-5}$ m$^3$/s; Case2: $N= 58.64$ rad/s; $Q= 1.67 \times 10^{-5}$ m$^3$/s.
<table>
<thead>
<tr>
<th>$N$, rad/s</th>
<th>$Q$, m$^3$/s</th>
<th>$d_{32}$ (Exp), μm</th>
<th>$d_{32}$ (Mod), μm</th>
</tr>
</thead>
<tbody>
<tr>
<td>36.65</td>
<td>$1.67 \times 10^{-5}$</td>
<td>8.86</td>
<td>7.24</td>
</tr>
<tr>
<td>58.64</td>
<td>$1.67 \times 10^{-5}$</td>
<td>2.24</td>
<td>2.56</td>
</tr>
</tbody>
</table>

**Local information of the stirred tank**

The gas-liquid mass transfer rate is typically the rate limiting step of the overall precipitation process. However, the rate limiting step was reported to be significantly affected by the local flow dynamics.\(^{41}\)

To address this issue, a time scale analysis was performed by calculating the characteristic time scales of gas-liquid mass transfer, micro mixing, nucleation and crystal growth for the operating condition with impeller rate of 58.64 rad/s and gas flow rate of $1.67 \times 10^{-5}$ m$^3$/s. It is important to note that the time scales are not constant during the precipitation process; especially the time scales of crystal nucleation and growth are both varying from 0.001 s to infinite, as supersaturation and growth rates approach zero when the precipitation stops (after approx. 90s as can be seen in Fig 4(a)). Regarding the mass transfer and the mixing time scales instead, they ranged respectively between 1-100 s and 0.0001-0.01 s, showing why the mass transfer rate can be considered as a the rate limiting step of the process. In order to understand the influence of local flow field on the gas-liquid precipitation, the local information of case 2 was presented.

As it is shown in Figure 5(a), the average gas volume fraction, $\overline{\varphi}_G$, decrease quickly during the first few seconds. In a similar case, Darmana et al.\(^{42}\) reported that the $\overline{\varphi}_G$ decreased by 40%-60% in the chemical absorption of CO$_2$ into NaOH solution. This is mainly due to the bubble shrinking caused by the CO$_2$ transferring from the gas to the liquid phase. It is worth noticing that the average gas volume fraction
starts to increase rapidly with time after 90s. The experimental and modeling results of $[Ca^{2+}]$ show that the precipitation stops at 90s in Fig 4(a). Then the accumulation of CO$_2$(l) will reduce the driving force of gas-liquid mass transfer, which leads to the increase of gas volume fraction with time.

![Graph showing the gas volume fraction increase](image1)

**Figure 5. The gas volume fraction, BSD and local $k_{L}a$.** (a) The average $\phi_G$, %; (b) BSD; (c) $k_{L}a$, s$^{-1}$

In order to understand the behavior of the gas-liquid mass transfer during the precipitation, the local bubble size distribution (BSD) and $k_{L}a$ at 30s are presented. The different BSD in different compartments are shown in Figure 5(b). As can be seen, the bubbles in compartment 1 are largest because coalescence phenomenon is prevailing in the feeding area due to the high gas volume fraction. The bubble distribution in compartment 30 is shifted towards small diameters since the breakage and mass transfer are prevailing in the impeller area due to the high turbulent intensity. In the suspension area, the size of bubbles is mainly determined by the combined action between coalescence and mass transfer. Therefore, bubbles
located in compartment 53 near the surface have average size in between these two extremes with a wider size distribution. Figure 5(c) shows instead the spatial profile of $k_La$ in the reactor. As it can be observed, the flow field has a strong influence on the local mass transfer rate: the larger values of $k_La$ are in the feeding area where the local $\phi_G$ is higher. The intensive breakage of bubbles in the impeller area can increase $k_L$ and $a$ significantly as well.

The final crystal size distribution (CSD) is the result of the concurrent actions of several phenomena, as nucleation, growth, agglomeration and breakage. The driving force of the precipitation is known to be the supersaturation, which is the key parameter in the calculation of crystal nucleation, crystal growth and agglomeration. The supersaturation is mainly determined by the chemical absorption of CO₂ into the Ca(OH)₂ solution and the flow field of liquid phase. In order to show the local precipitation rate, the evolution of local supersaturation with time is shown in Figure 6.

![Figure 6. The local supersaturation](image1)

![Figure 7. The local $d_{32}$ of crystals.](image2)

The evolution of supersaturation is faster in feeding region, such as compartment 1, due to the high mass transfer rate. The difference of local supersaturation in other regions of the tank is quite small because the mixing of liquid is uniform due to the high impeller rate. Therefore, the local growth rate and nucleation rate should be similar in agreement with the assumption 4. The final CSD is mainly decided by the breakage and agglomeration in the compartmental model.
The local Sauter mean diameter, $d_{32}$, of the crystals calculated as the ratio of the 3\textsuperscript{rd} and 2\textsuperscript{nd}-order moments is shown in Figure 7, together with the volume-averaged $d_{32}$. The volume-averaged $d_{32}$ is obtained by counting the crystals in all of the compartments. In this case, the average $d_{32}$ is influenced by the competition of nucleation, growth, agglomeration and breakage: the nucleation, growth and agglomeration are prevailing before 25s, corresponding to the highest value of supersaturation reached by the system. Then the breakage of larger agglomerates leads to the decrease of average $d_{32}$, following the decrease of the supersaturation after 25s. It is clearly shown that the stronger turbulence will lead to earlier decrease of $d_{32}$ and small crystals size, since the breakage frequency is a function of energy dissipation and the size of agglomerates. In addition, the supersaturation vanishes at about 90s when the pH drops. Further gas feeding after this point is not beneficial as may cause CO\textsubscript{2} accumulation and consequent crystal dissolution.

In the view of above discussion, the combination of flow field and complex sub-models is necessary in the simulation of gas-liquid precipitation, as neglecting the interactions between the flow field and bubbles/particles dynamics may cause large errors during the design and scaling up of these processes. The compartment model represents a cheap alternative to full CFD simulations, since it is capable of coupling the flow dynamics and the interconnected sub-models, allowing also to evaluate the effect of operating parameters that do not modify the fluid field such as the purity of CO\textsubscript{2}(g) and the initial concentration of Ca\textsuperscript{2+}, without any additional CFD calculations.

**Conclusion**

This paper introduces the compartmental model to estimate the influence of the fluid dynamics on the gas-liquid precipitation in a pilot-scale stirred tank. This approach inherently combines the local fluid field with complex sub-models including mass transfer, chemical reaction, precipitation and population balance for bubbles and crystals. Compartmentalization is performed from a steady state two-fluid CFD
calculation by considering the spatial distribution of turbulent dissipation. The physical sub-models are integrated together with the flow field description in the compartmental model.

The flow dynamics has a strong influence on the local mass transfer coefficient and the particle size distribution of crystals. Due to the low computational costs compared to a full CFD simulation, the compartmental model allows time-consuming operations, such as the fine-tuning fitting of the sub-model parameters. With further experimental investigations, several model parameters in the sub-models can be fine-tuned and used to scale-up the process. By appropriate division of the fluid domain, the compartmental model can also be used to successfully model gas-liquid precipitation process in different geometries.

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Notation

\[
\begin{align*}
\alpha & = \text{exponent of nucleation} \\
\beta & = \text{exponent of growth} \\
\gamma & = \text{exponent for particle size in power-law breakage kernel} \\
\varepsilon & = \text{turbulent energy dissipation} \\
\mu & = \text{dynamics viscosity} \\
\rho & = \text{density} \\
\varphi & = \text{gas hold up in each compartment} \\
\psi & = \text{collision efficiency} \\
\Omega & = \text{strength of aggregate}
\end{align*}
\]
\[ \eta = \text{Kolmogorov length scale} \]

\[ \eta_T = \text{Kolmogorov time scale} \]

\[ c = \text{concentration of component} \]

\[ aq = \text{aqueous solution} \]

\[ B = \text{birth function} \]

\[ D = \text{death function} \]

\[ D_{\text{CO}_2} = \text{diffusion coefficient of carbon dioxide in the liquid} \]

\[ F_{ij} = \text{liquid flow rate from compartment } i \text{ to compartment } j \]

\[ g = \text{gas phase} \]

\[ G = \text{growth term of bubbles and crystals} \]

\[ K_{\text{sp}} = \text{solubility product} \]

\[ k_n = \text{nucleation rate constant} \]

\[ k_g = \text{growth rate constant} \]

\[ l = \text{liquid phase} \]

\[ m_k = \text{the } k\text{-th moment of crystals} \]

\[ M = \text{torque of the impeller} \]

\[ n = \text{amount of component} \]

\[ N = \text{nucleation term of crystals} \]

\[ N_r = \text{impeller speed} \]

\[ NB = \text{total number of compartments} \]

\[ NC = \text{total number of bubble size categories} \]

\[ N_{\text{GL}} = \text{flux of gas-liquid mass transfer} \]

\[ r = \text{reaction rate} \]

\[ R = \text{agglomeration and breakage rate} \]

\[ S = \text{supersaturation} \]
\[ U_{ij} = \] gas velocity from compartment \( i \) to compartment \( j \)

\[ V = \] volume of compartment

\[ v = \] volume of single bubble of each category

\[ Y = \] bubble number density

subscripts

\[ \text{ag} = \] agglomeration

\[ \text{br} = \] breakage

\[ \text{cl} = \] coalescence

\[ \text{disp} = \] dispersion

\[ G = \] gas phass

\[ i, j = \] index of compartment

\[ in = \] index of the feeding compartment

\[ k = \] index of the moment

\[ L = \] liquid phase

\[ \text{out} = \] index of the flow out compartment

\[ p = \] index of bubble size category

\[ \text{slip} = \] slip velocity

Literature Cited


