Electrical Capacitance Volume Tomography Validation of Column Flotation Process with Gas Injection Variation

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Abstract. Electrical capacitance volume tomography (ECVT) is proposed as a novel real-time monitoring system for column flotation process. This study aims to validate the results between static and dynamic simulations using ECVT and ECVT-Computational Fluids Dynamics (ECVT-CFD) and real-time column flotation monitoring experiment using ECVT system. Bubble fractions from both the simulations and the experiments with the variation of gas velocity (0.02, 0.03 and 0.04 m/s) were compared and measured. The results of the study showed that ECVT experiments successfully simulated the images of bubbles inside the real-time column flotation. Both simulations and experiments showed that the bubble fraction increased with the increasing of gas velocity. Furthermore, the comparison between the simulated bubble fraction and the experimental results gave a validation accuracy range of 86%-98%. It was also indicated that at higher gas velocity (0.04 m/s) the accuracy of the simulations is better than that of lower gas velocity (0.02 and 0.03 m/s).

Keywords: real-time monitoring, bubble fraction, column flotation process, ECVT, gas injection variation, OpenFOAM.

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1. Introduction

Flotation is a separation process of valuable minerals from their impurities by utilizing the difference in the mineral surface properties in term of its hydrophobicity. The surface properties are obtained by manipulating the surface chemistry of minerals by adding some reagents into the process. Among the existing flotation techniques, column flotation is widely used because it has significant advantages compared to the conventional techniques [1]. In general, mineral separation using column flotation machine is conducted by blowing the gas from the bottom of a column to form the gas bubbles which will lift hydrophilic minerals to the surface (froth zone) as a concentrate, while the hydrophile will remain in the pulp as a tailing. The efficiency of the separation process can be determined from its metallurgical performance, which are the mineral grade and the recovery of the process.

The efficiency of the flotation process can be monitored statically and dynamically by observing the characteristics of bubbles generated during the process. Static monitoring evaluates the size, shape and colour of the bubbles, whereas dynamic monitoring perceives the size distribution, number, rate and stability of the bubbles. Until recently the flotation monitoring system is utilizing a high-resolution camera connected to a computer program called machine vision [2–4]. They are able to monitor the characteristics of static and dynamic bubbles on the surface (field x-y). Image analysis of the processes can describe the properties of bubbles so that the flotation performance can be examined [5–6]. They have no ability to monitor the bubble characteristics in the vertical direction (z-axis), because the camera cannot penetrate the flotation cell. However, in industrial applications, this two-dimensional machine vision is still employed as a monitoring system in flotation process [4, 7–8].

Electrical Capacitance Volume Tomography (ECVT) is a tomography technique which produces real-time three-dimensional ECVT image based on the electrical capacitance measurement. ECVT has a great ability to distinguish object based on the permittivity of materials and can also be utilized for multi-phase processes [9]. This technique has been widely used in chemical and petrochemical industries. Owing to its advantages: non-invasive, non-intrusive, non-radiative, low-cost, and high-speed capability, ECVT has been widely applied in the bubble columns and fluidized beds reactors for many purposes [10–14].

For multiphase flow study purposes, a Computational Fluid Dynamics (CFD) simulation is involved so that the flow behaviour and design of a process with the specific parameters can be revealed and optimized after verified by some modalities [15–17]. As a promising tool, ECVT models can be used for verifying the CFD codes of a controlled experiment simulation. Some studies about comparison results of CFD simulation and electrical capacitance tomography (ECT)/ECVT have been reported, among of those are the study of pneumatic transport of granular materials in a inclined conveying pipe, the multiphase flow studies in oil and gas industries, and the drying processes in fluidized bed [15–17].

Column flotation process is influenced by some parameters [18], which are chemical components (collector, frother, activator, depressant, pH), tool components (design of column, agitation, gas injection), and operational components (feeding rate, mineralogy, particle size, pulp density, temperature). The combination of these parameters will determine the optimum condition of column flotation process called metallurgical performance as described previously. It is expected to monitor and observe the flotation process in the three-dimensional axes and real-time condition. ECVT is proposed to be applied in a column flotation process because of its powerful capability in real time and 3D monitoring system.

In this study, the validation of column flotation process was carried out by using CFD simulation and ECVT experiment. CFD simulation and ECVT experiment studies were performed by varying gas injection debit represented by gas velocities in a laboratory scale column flotation equipment. Other parameters were kept constant. At this stage, the study focused on bubbles behaviors in terms of their morphology and fraction resulted from both simulations using ECVT-CFD and experiments using ECVT. The results were compared and analyzed. Relative accuracy measurement was carried out to analyze the similarity in bubble fraction between simulation and experimental results in this method. The results were expected to give fundamental aspects to the further studies in the future, which will include the determination of metallurgical performance in terms of its flotation recovery.

2. Methods

As seen in Fig. 1, the approach has been done by simulation and experimental studies, respectively. At the beginning step of the simulation part, ECVT simulation obtained the threshold value which will be used to
distinguish the gas and liquid phase from the 3-D ECVT reconstructed image by utilizing the bubble swarm model. After that, an ECVT-CFD simulation of gas injection was performed to provide the volume fraction value of gas bubbles produced. Whilst in the experimental part, the column flotation experiment was conducted within the similar condition to the simulation. The parameter of gas injection variable used in this study was the gas velocity of 0.02, 0.03, and 0.04 m/s. All simulations, including statics and dynamics condition, was performed by finite volume method application using OpenFOAM for windows 16.02 and Python 2.7.9 version which run in a computer with the specification of 32 GB of RAM, Graphic Card GTX, Intel i7 Second Generation processors. 1.8 Terabytes of internal hard drive capacity and Windows 7 64 bit Operating System.

Fig. 1. Flow chart of the study.

2.1. Sensor Design and Sensitivity

32-electrode ECVT sensors (8-electrode per plane) were used both in simulation and experiment (Fig. 2). Diameter and height of ECVT sensor are 50 mm and 100 mm respectively. The shape of the electrode is rectangular with the dimension of 17 mm x 17 mm. The use of shifted plane design of ECVT sensor between each electrode’s plane is based on [19]. The number of measurement produced using 32-electrode is based on the Eq. (1):

$$M = N(N - 1)/2$$

Thus, we got the measurement data of 496, where N is 32. The measurement data also represents the electric field of each pair of electrodes. The distribution of electric field in the sensor can also be used for identifying sensing zone and dead zone as shown in Fig. 3. Sensing zone or sensitive zone is an area where the objects can be easily detected whilst the dead zone is an area where the objects cannot be easily detected due to no variations of its electric field. Figure 3 shows the correlation graph between normalized sensitivity for all 496 capacitance pairs and the axial direction (z-axis) of the sensor. The dead zones are found at the bottom and the top portions of the sensor domain, which are the layer 1 to 5 and 26 to 32, respectively.
Fig. 2. 32-electrode ECVT sensor with 8-electrode per plane.

By adding more variation in the simulation, the distribution curve would be smoother, hence the good convergence shape image would be achieved [19].

Fig. 3. Axial distribution of normalized sensitivity for all 496 capacitance pairs. The dead zones are indicated by the dashed line.

2.2. ECVT Simulation of Bubbles Swarm model (Static Condition)

The simulation used bubbles with 1.5 mm bubble diameter and as the object of simulation, they were approached as a bubble swarm model (see Fig. 4). Physical parameters (relative permittivity value) of simulation were defined 1 for each bubble and 12 for surrounding bubbles. The simulation was conducted in three objects of bubble swarm model (1, 3 and 6 layer(s)) as can be seen in Fig. 4(a) which the distance between each layer was 2.5 mm. The objective of the simulation was to obtain the threshold values of the 3-D ECVT reconstructed images. Furthermore, the threshold values were used for distinguishing the gas and liquid phases from 3-D ECVT reconstructed image both from ECVT-CFD simulation and ECVT experiment.
Fig. 4. The object of simulation - a bubble swarm model. (a) top-view. (b) side-view.

2.3. ECVT-CFD Simulation

ECVT-CFD simulation is a combination of CFD and ECVT simulation. As the first step, CFD simulation was conducted to obtain the flow structures, and after that ECVT simulation was carried out by governing the electrical properties of phases previously generated by CFD simulation. Two-phase system approach has been used in the CFD simulation, where gas bubbles regarded as gas phase, and pulp (a mixture of ore, reagents, and water) as a liquid phase. Density and viscosity parameters of gas and liquid phase were adjusted based on the real condition of column flotation experiment, which is 1 kg/m$^3$ and 1 x 10$^4$ Pascal/m$^2$ for gas phase, and 1,280 kg/m$^3$ and 1.83 x 10$^6$ Pascal/m$^2$ for the liquid phase. The diameter and height of CFD simulation were set to 50 mm and 45 mm. Boundaries of CFD simulation were defined as wall, inlet, and outlet. Hence, the flotation column’s wall was considered as a wall, the bottom of the column was set as an inlet, and top of the column was assumed as the outlet (see Fig 5). The wall and outlet boundaries were in a zero gradient condition which means constant. Inlet (bottom of the column) was varied to define the gas injection variation which was set for each CFD simulation to a fixed value (0.02, 0.03, and 0.04 m/s). To obtain the results of CFD simulation as a function of gas injection, the OpenFOAM was used for solving the Navier-Stokes equation which can be seen in Eq. (2).

$$\frac{\partial \alpha}{\partial t} + (\vec{U} \cdot \nabla) \alpha = 0$$

(2)

Where $\alpha$ is the volume fraction which was set as 0 (zero) for gas phase and 1 (one) for the liquid phase. In which, density and kinetic viscosity of $\alpha$ were expressed as the following Eq. (3) and Eq. (4):

$$\rho(X,t) = \rho_{\text{water}} \alpha + \rho_{\text{air}} (1 - \alpha)$$

(3)

$$\mu(X,t) = \mu_{\text{water}} \alpha + \mu_{\text{air}} (1 - \alpha)$$

(4)

CFD simulation results were then used as the objects with permittivity in ECVT simulation. The $\alpha$ values, which were the flow structures from CFD simulation, were converted to electrical properties as relative permittivity through python program. In the ECVT simulation, the relative permittivity of bubbles where $\alpha$ equalled to 0 was 1, while the pulp where $\alpha$ equalled to 1 was 12. The input voltage of each electrode was set to 5 volt, and other boundaries were set as zero charges.
2.4. Global Thresholding

Segmentation involves separating a raw image into regions that correspond to the objects. Thresholding is one of the simpler segmentation methods. Here in the study, global thresholding method was used to distinguish the liquid phase and the gas phase. The common method of global thresholding as in Eq. (5):

\[
g(x, y, z) = \begin{cases} 0, & f(x, y, z) < T \\ 1, & f(x, y, z) \geq T \end{cases}
\]

where \( T \) is the thresholding value, \( f(x, y, z) \) is the 3-D image before thresholding and \( g(x, y, z) \) is the 3-D image after thresholding. In this study, to get the optimum \( T \), we compare the result of thresholding process to the phantom image from simulation data. Otherwise, \( T \) values obtained were averaged and used for thresholding on the ECVT-CFD simulation and experiment images results.

2.5. Relative Accuracy Measurement

Error percentage measures the accuracy of a measurement by comparing a tested value with an accepted value. The error was calculated in the following Eq. (6):

\[
error \ (% ) = \frac{|E-A|}{A} \times 100\%
\]

where \( E \) is a tested value and \( A \) is an accepted value. In this study, simulation results are the tested value, while the experimental results are the accepted value. If the error between simulation and experimental results is less than expected, then the CFD codes are good, otherwise poor. Thus, relative accuracy (\( a \) in \( \% \)) of the simulation results compared with the experimental results can be expressed as follow:

\[
relative \ accuracy, \ a = 100\% - error \ (% )
\]
2.6. Column Flotation Experiment with ECVT System

The experiment was conducted using a column with 50 mm diameter and 150 mm total height, as shown in Fig. 6. An air sparger made from ceramic material was installed at the bottom of the column. ECVT sensors as described previously were set at 5 cm above the sparger. The gas injection variation was set at the same value of CFD simulation, which is 5% solid-water. Feed, mixture reagents which were conditioned for 15 minutes prior the usage, was injected from the top of the column so that the separation occurred. On the other side, ECVT system captured the data signal of flow phenomena and transferred it to a computer for further analysis.

![Fig. 6. Schematic diagram of column flotation experiment monitored by ECVT system [20].](image)

3. Results and Discussion

All 3-D ECVT reconstructed images were reconstructed using Iterative Linear Back Projection (ILBP) algorithm technique. Since the quality of reconstructed image using ILBP algorithm depends on the constant $\alpha$ (gain factor) and a number of iteration, thus to obtain the optimum parameters, those two parameters were varied on the reconstruction process and evaluated by using correlation coefficient (CC) value comparing phantom and reconstructed images. In this study, $\alpha$ and iteration were limited to 0.1-1 with interval 0.1 and 100 with interval 1, respectively. The simulation was set as in static simulation (1 layer) as seen in Fig. 8(a). The detail about the algorithm technique can be found in [21].

Figure 7 shows the relation between the CC and number of iteration with $\alpha$ from 0.1 to 1 on ILBP system. Based on this study, the optimum parameters for $\alpha$ and iteration are 1 and 100, respectively. Thus, all 3-D ECVT reconstructed images were processed using these parameters.
Fig. 7. Relationship between correlation coefficient and iteration with $\alpha$ 0.1-1 on ILBP system.

3.1. ECVT Simulation of a Bubble Swarm Model (Static Condition)

Figure 8 shows the three-dimensional reconstructed images compared with the phantom of the static condition. Figure 8(a) is the real object of simulated phantom and Fig. 8(b) is the reconstructed images of ECVT seen in 3-D and axial views at slice 16. The result showed that the simulated objects as bubble swarm model of 1, 3, and 6 layer(s) were significantly different qualitatively (Fig. 8). Ideally, as can be seen in Fig. 8, red color depicts the liquid phase, while blue color depicts the gas phase. Due to the ill-posed and ill-conditioned characteristics of ECVT method, the reconstructed images will have some errors. Also, measurement and numerical reconstructed image, as well as the algorithm technique, contributes the errors. As seen in Fig. 8(b), the center region of each reconstructed image is dominated with the high permittivity values (yellow-red colors), whilst air-bubble is distributed homogeneously in phantom object (Fig. 8(a)). Thus, the thresholding method is needed to quantify each fraction (liquid and gas phases) as describe in previous section. Threshold value was determined by estimating the fraction volume of each phase based on the permittivity distribution compared to the phantom. Based on the simulation, the threshold values for each simulation are 0.380, 0.410, and 0.503 for 1, 3, and 6 layer(s) simulation, respectively. Thus, the average $T$ value of 0.431 was used for all ECVT-CFD simulations and ECVT experiments in the current work, where the permittivity values higher and equal than T value were defined as a liquid phase while otherwise they are defined as gas (bubbles) phase.

3.2. Bubbles Imaging

Figure 9 shows the image comparison of CFD simulation, ECVT-CFD simulation and ECVT experiment. In the phantom images, the blue color depicts the bubbles (gas phase). In the ECVT-CFD simulation and ECVT experiment images, the blue color depicts the air bubbles and the red color depicts the liquid phase as well as the colors between them are defined based on the $T$ value obtained in the static simulation.
which, the values lower than \( T \) are defined as gas phase, and the values higher and equal than \( T \) are defined as liquid phase.

\[
\text{(a) 3-D View (phantom), (b) 3-D View (reconstructed image), (c) Top View (Slice-16)}
\]

Fig. 8. Simulation results of static condition (a bubble swarm model), red colour is liquid phase and blue colour is gas phase. (a) Phantom, (b) 3-D ECVT reconstructed images.
As shown in Fig. 9, the number of bubbles generated independently was affected by the gas injection. Qualitatively, the reconstructed images from simulation and experimental results were visible similar. However it seems that the reconstructed images of bubbles visualize a bubble swarm rather than individual bubble. At low gas velocity (0.02 m/s), the shapes of bubbles imaging were blurred in both of simulations and experimental imaging results. At higher gas velocity 0.03 and 0.04 m/s, the bubbles shapes became clearer. The shapes of bubbles indicated the movement from the bottom part to the top of the z axis. In general, the increase of gas injection could increase a number of bubbles and also affected the results of 3-D ECVT reconstructed images. Bubbles agglomeration phenomena were also observed in both, simulation and experimental results.

![Fig. 9. The comparisons of ECVT-CFD simulation and ECVT experiment with the phantom from CFD simulation (units in mm).](image)

In CFD simulation results or phantom, the liquid phase is indicated by grey colour, and gas phase is indicated by dark-blue colour. While, both in ECVT-CFD simulation and ECVT experiment results, the liquid phase is indicated by red colour and gas phase is indicated by other colours (usually blue to orange colour). When gas velocities are varied, the ECVT images results are clearly different qualitatively, in which the region of red colour (liquid phase) decreased as the gas injection increased. This observation agrees with
column flotation experiment, the bubble motion and bubble rise velocity were influenced by gas injection which was blown by compressor [22, 23]. In general when gas injection increased, the bubble moved spirally and the shape of the rising bubble could be sphere or oval [24–26]. At this stage, those phenomenon were also observed, however more detail investigation may be conducted to elucidate the bubbles or bubble swarm movements. To quantify and compare the bubble fraction of each study (simulation and experiment), the average threshold value of 0.431 obtained in section 3.1 was used. The volume elements (voxels) [19] less than threshold value were defined as gas phase while the higher voxels were defined as a liquid phase for both ECVT-CFD simulation and ECVT experimental studies.

3.3. Bubble Fraction and Relative Accuracy

Figure 10 shows a relation between bubble fraction and gas injection or gas velocities. Generally, the bubble fraction increased when gas injection or gas velocity is increased. For phantom (CFD results), the bubble fraction increases from 10.3% to 17.2%, whilst for simulation study (ECVT-CFD results), the bubble fraction increased from 9.85% to 16.8%, and for experimental study (ECVT results), the bubble fraction increased from 8.64% to 16.6%.

The differences of bubble fraction percentage between simulation and experiment showed in Fig. 10 may be caused by the differences in some parameters which are not governed in detail by CFD simulation. For example, density and viscosity of the simulation are calculated based on the number of theory, while the density and viscosity of experimental conditions are calculated based on actual weight and volume of the component substances, which included the reagents itself, some water and galena inside. Hence, simulated conditions for pulp might differ with the experimental conditions. Another reason is surface tension. In the OpenFOAM simulation, the surface tension between bubble-pulp used the maximum limit of the sigma value, consequently, the bubbles generated were not easily broken, while in the experiment most of the bubbles will always be ruptured and it affected the results.

As for column pressure, in the simulation, column pressure used zero gradient condition, which means the pressure generated will be distributed evenly on the column; therefore the pressure on the entire surface of the column has the same value. In the experimental conditions, column pressure was not uniform. The pressure at the centre of the column was higher than at the edge, and also from the bottom to the top of the column. These might cause bubble fraction discrepancies between simulation and experiment.

Table 1 shows relative accuracy of ECVT-CFD simulation compared with the ECVT experiment. Basically, the relative accuracy of simulations and experimental results were more than 86%. At the higher gas velocity of 0.04 m/s, the relative accuracy increased up to 98.7 %, which indicated a relatively good correlation between simulations and experiments. However, further improvement in optimum accuracy is still needed.

4. Conclusion and Future Direction

Investigation of column flotation process with gas injection variation using CFD simulation and ECVT measurement was conducted in this study. By using the threshold value obtained from the simulation of a bubble swarm model (static condition), the bubble fraction generated from CFD simulation (phantom), ECVT-CFD simulation, and ECVT experiment were compared and validated. Based on this study, bubbles images in the column experiments were successfully reconstructed by using ECVT. Qualitatively, their morphologies showed similarities with simulation results. For all methods used in this study, bubble fraction increased as gas injection increased. At higher gas velocity, the simulation result showed relatively better relative accuracy. It may be caused by the conditional discrepancies of parameters used in CFD simulation. Thus, laid for the future direction of the study, the CFD simulation needs to consider some aspects for approaching the real condition of the experiment so that the relative accuracy between simulation and experiment studies will increase. Further related study of real time-3D of column flotation monitoring system by using ECVT which will be related to their metallurgical performance, is still undergoing.
Fig. 10. The relationship between bubble fraction and velocity of gas obtained from simulation and experimental results.

Table 1. Relative accuracy of ECVT-CFD simulation compared with ECVT experiment.

| Gas velocity (m/s) | Relative accuracy (%) |
|-------------------|-----------------------|
| 0.02              | 86.0                  |
| 0.03              | 86.3                  |
| 0.04              | 98.7                  |

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