Magnetic induction tomography guided electrical capacitance tomography imaging with grounded conductors

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Abstract – Electrical capacitance tomography (ECT) is an imaging technique commonly used for imaging dielectric permittivity of insulating objects. In applications such industrial process tomography and non-destructive testing (NDT), the objects under test may exhibit variations in both dielectric permittivity and electrical conductivity. In particular, a sample that includes high conductivity, such as metal, can cause a large change in electrical field in ECT. The metal sample in imaging area will cause a large change in the sensitivity map of ECT compared to free space, which will make the ECT image reconstruction inaccurate. This effect is more severe in grounded conductor than floating conductors, so this paper focuses on grounded conductor. In order to update the sensitivity map, one needs to gain information about the conductivity distribution in ECT problem. Magnetic induction tomography (MIT) is sensitive to electrical conductivity and not sensitive to permittivity variations; therefore, it can be used to visualize the conductivity distribution of the target under test. In this paper, a dual-modality MIT and ECT system is proposed to image a medium including conductors and dielectrics. Both simulated and experimental results are presented, which demonstrate the feasibility of the proposed method.

Keywords: Dual-modality ECT/MIT, dielectric imaging with grounded conductors, sensitivity map.

1. Introduction

Electrical capacitance tomography (ECT) is a non-invasive technology that can be used to monitor industrial processes or defect detections in non-destructive testing (NDT). The aim of ECT is to visualize the unknown permittivity distribution via measuring capacitances between pairs of peripheral electrodes around the samples. Recent applications of ECT include monitoring gas-solid flows in pneumatic conveyors and gas-oil in oil pipeline [1, 2]. Typically, a time difference imaging is used in ECT image reconstruction. The change in permittivity is reconstructed by the sensitivity map of reference condition and the difference
between reference measurements and sample measurements [3]. Therefore, the accuracy of a sensitivity map has a significant effect on results of ECT image reconstruction. When there are grounded conductors in ECT sensor area, the electric fields will be significantly different than that of a medium with only dielectric materials. Therefore, the sensitivity map of a typical reference condition, i.e. air-filled sensor, is not accurate [4]. To decrease the effect of increased nonlinearity caused by high conductivity, a precise forward model must be introduced. This modified model, which is close to the real physical scenario, includes the highly conductive content, which is modelled as grounded conductor. The previous study of metal in an ECT sensor has been done by Ville Rimpiläinena et al. [5], where the high-shear mixer has a typical structure with a metal shaft in the centre. For better quality of visualization of the process in the mixer, the known position and size of central shaft was taken into account. Similar to the method used in EIT with metallic samples in [6], the surface of metal is regarded as one electrode of the sensor, and it is considered electrically grounded. Furthermore, the boundary condition on conductor is set at zero voltage. In all above applications location of grounded conductors are known. Here we extended this to the case that the location of conductor is unknown.

To detect and identify the high conductivity in ECT imaging, magnetic induction tomography (MIT) imaging is used. Using the location of grounded metal a precise sensitivity map in ECT is created. MIT aims to visualized the distribution of passive electromagnetic properties in particular permeability $\mu$ or conductivity $\sigma$ [7]. In MIT, the magnetic field induces eddy currents in the conductive object, which produces a change in the magnetic field received by the receiver coils. This paper proposes a combined MIT and ECT system as a dual modality imaging solution for imaging metallic and dielectric samples. MIT is not sensitive to dielectric permittivity variations, which allows a sequential data fusion to be adapted on this dual modality method. The information of electrical conductivity from MIT feeds into the ECT image reconstruction software allowing reconstruction of both conductivity and permittivity. Compared to the dual-modality electrical resistance tomography systems (ERT) and ECT [8, 9], ECT and MIT combination offers a fully contactless imaging solution. ERT requires electrical connection to objects making it unsuitable for many applications. In [10], a two electrode capacitive measurement system is used to image either void (dielectric permittivity change) or rebar (metal). In this paper we propose a dual modality method, in which dielectric permittivity changes and grounded conductor can be monitored simultaneously. This dual modality imaging method provides a wider range of applications for ECT, where the conductive media between an object and a sensor is not known. This paper demonstrates that the ECT imaging capability of dielectric variations will be substantially reduced, in particular when the ground conductors are close to the dielectric
inclusions. The dual modality has potential applications for mixture of both conductive and dielectric samples, such as defect detection for steel reinforced concrete.

The aim of this paper is to investigate a dual modality imaging for a combination of ground conductors and dielectric inclusions using MIT and ECT. The structure of the paper is organised as follows. In section 2, the ECT and MIT hardware and models are introduced. In section 3, the error tolerance of ECT is will be evaluated by the figure of merits, which is developed by Andy Adler et al.[11]. In section Error! Reference source not found., the experiment of dual-modality MIT/ECT is conducted, and the final results are assessed.

2. ECT and MIT hardware and models

Here we briefly describe ECT and MIT system, sensor and forward modelling and sensitivity maps. A 16 channel MIT system and a 12 channel ECT system is used for sequential data fusion in dual modality ECT and MIT imaging, Figure 1 shows both sensor arrays.

![Figure 1. (a) The 12-electrode ECT sensor; (b) The 16-coil MIT sensor.](image)

2.1 ECT

A typical ECT sensor consists of 6, 8, 12 or 24 electrodes [12], which are evenly mounted outside a pipe-shaped non-conductive wall and separated with radial screens. And an external screen for shielding the noise from outside is installed. All electrically grounded screens reduce the effect of the external capacitance between pairs of electrodes. In some special
cases, where the wall is conductive, the electrodes are built on the inner surface of the wall, like a metal vessel or pipe.

Figure 1(a) is the ECT sensor used in our experiment, which is composed of a plastic pipe, 12 electrodes, radial screen between the electrodes and external shielding. The external diameter of pipe is 150 mm, the size of the electrode is $217 \times 32$ mm$^2$ and the screens between the electrodes are 3 mm wide. The capacitance measurement unit is the PTL 300E ECT system, whose excitation frequency is fixed at 1.25MHz. Twelve channels are connected to the electrodes to measure the inter-capacitance.

Forward problem of ECT is to calculate the capacitance, $C$, for given values of the known permittivity distribution, $\varepsilon(x,y)$, over a given geometric region and excitation signal (the voltage of the excited electrode, $u$). During the measurement, one electrode is selected for excitation and the other electrodes are grounded as detectors, then this protocol will be applied to each electrode in turn. Maxwell’s equations in electro quasi-static are applied to solve the forward problem of ECT [13]. The propagation effects can be neglected, since the dimensions of the sensor are much smaller than the wavelength of the signal wave in low frequency (10Hz-2MHz). The ECT governing equation is

$$\nabla \cdot \varepsilon(x,y) \nabla u(x,y) = 0 \text{ in } \Omega \tag{1}$$

where $u(x,y)$ is the electric potential. And the potential on the excited electrodes is known as

$$u(x,y) = V \text{ on } S_l \tag{2}$$

And potential on sensing electrodes are

$$u(x,y) = 0 \text{ on } S_j \tag{3}$$

where $S_l$ and $S_j$ are the surface of the excited and receiver electrodes respectively, and $V$ is the excitation voltage.

The derivative form of electrical potential is applied in the equation of electric charge, $Q_j$, on the excited electrode, $S_j$:

$$Q_j = - \int_{S_j} \varepsilon(x,y) \frac{\partial u(x,y)}{\partial n} dS \tag{4}$$

where $n$ is the inward normal of $S_l$.

Therefore the capacitance between electrodes $i$ and $j$ can be expressed as a function of permittivity distribution.

$$C_{ij} = \frac{Q_j}{V} = f(\varepsilon(x,y)) \tag{5}$$

Based on finite element method (FEM), the permittivity distribution is divided into $n$ elements. And the forward problem can be solved to calculate the potential distribution, distribution of electric fields and estimated capacitance measurements.
The sensitivity map can be calculated using an efficient formulation based on calculated fields from excitation and sensing electrodes [4].

$$\frac{\partial C_{ij}}{\partial \varepsilon} = - \int_{\Omega} \nabla u_i \nabla u_j dS$$  \hspace{1cm} (6)

where $u_i$ and $u_j$ are potential over region $\Omega$ when electrodes $i$ and electrode $j$ are excitation electrodes respectively.

Then the perturbation of the permittivity distribution brings the change in the electric charge on electrode $l$ has been given in [4] and [14].

$$\Delta C = \left[ \frac{\partial C_1(\varepsilon)}{\partial \varepsilon_1} \frac{\partial C_2(\varepsilon)}{\partial \varepsilon_2} \cdots \frac{\partial C_n(\varepsilon)}{\partial \varepsilon_n} \right]_{1+n} \ast \begin{bmatrix} \Delta \varepsilon_1 \\ \Delta \varepsilon_2 \\ \vdots \\ \Delta \varepsilon_n \end{bmatrix}_{n+1}$$  \hspace{1cm} (7)

In the case of free space, the sensitivity map of a pair of electrodes is shown in Figure 2. The shallow valley indicates sensitivity distribution of a measurement between a pair of electrodes in opposite position.

![Figure 2. The sensitivity of opposite capacitance measurement in free space](image)

2.2 MIT

A 16 channel low frequency (50 kHz excitation for metallic imaging) MIT system is used to realise the proposed dual-modality imaging technique. The MIT system consists of: (1) a coil array of sixteen air-cored inductive sensors, shown in Figure 1(b); (2) a sixteen channel multiplexer for channel switching; (3) a national instrument (NI-6295) data acquisition card; and (4) a host computer, where the data process and image reconstruction take place. This system was designed to measure targeted object(s) with high conductivity, which corresponds to a negative imaginary part of the magnetic field perturbation, as such, the measurements can be approximated by their amplitudes, the phase shifts are therefore neglected [15]. The system development has been reported in[16], and many applications have been proposed using this system architecture.
To solve the forward problem, finding the magnetic vector potential $A$ is the key. There are many FEM based formulations can be used to solve the A field, such as $(A,A-V)$ and $(A,A)$ formulation. In this study, we adopted a $(A,A)$ formulation using edge based FEM [17, 18].

$$\nabla \times \left( \frac{1}{\mu} \nabla \times A \right) + j \omega \sigma A = J_s$$  \hspace{1cm} (8)

where $\sigma$ is conductivity, $\omega$ is angular frequency of the excitation current, $\mu$ is permeability, and current density $J_s$ can be prescribed by magnetic vector potential from the Biot-Savart Law. If the total current in the excitation coil was $I_0$ the sensitivity of the induced voltage to the conductivity change can be written as:

$$\frac{\partial V_{mn}}{\partial \sigma_k} = -\omega^2 \frac{I_0 A_m A_n dV}{I_0}$$  \hspace{1cm} (9)

where $V_{mn}$ is the measured voltage, $\sigma_k$ is the conductivity of pixel $k$, $\Omega_k$ is the volume of the perturbation (pixel $k$), $A_m$ and $A_n$ are respectively solutions of the forward problem when excitation coil ($m$) is excited by $I_0$ and sensing coil ($n$) is excited with unit current.

3. Analysis of metallic sample in ECT imaging

Our experimental scenario is schematically shown in Figure 3, which is an ECT sensor with grounded conductor. For image reconstruction algorithm in both ECT and MIT we have used standard Tikhonov regularisation method [14]. We have selected an empirical regularisation parameter for MIT and one empirical regularisation parameter for ECT. This way we can assume same level of regularisation in all cases allowing better comparison. The same level of thersholding was used for post processing imaging results for quantitative imaging comparisons.

![Figure 3. Schematic drawings about cross section of the ECT sensors and grounded metal](image-url)
3.1 Reference measurement:
In our experiment the sample under test consists of a metallic bar and a wooden bar. Traditional ECT utilizes the difference between the reference capacitance ($C_r$) and the measured capacitance of the sample ($C_m$) to visualize the dielectric distribution change. In the case of sensing a mixture of a metallic sample and a dielectric sample, without knowing the existence of the metal, the capacitance measurement of the air-fulfilled sensor ($C_{air}$), is chosen as the background data (or reference data). The difference between the measurement of samples ($C_m$) and $C_{air}$ is utilized to solve the inverse problems; however the metal will affect the imaging on the rest of dielectric region. Therefore choosing the capacitance measurement of metal and air, ($C_{metal+air}$) as the reference measurement describes the real condition more precisely.

3.2 Modification of forward model
If the location and size of grounded conductors are known, we can modify the ECT forward model to account for these grounded conductors. In the case of a metallic bar, the sensitivity map of a pair of electrodes is shown in Figure 4 [4, 5]. Compared with Figure 2, the sensitivity “valley” is split up around the metallic bar, since the metallic bar is defined as a grounded electrode of the sensor. This forward model is proximate to the real condition within the sensor area.

Figure 4. The sensitivity maps of opposite capacitance measurement when the metal bar stands in the centre and near one electrode.
Two images will be reconstructed from different reference measurements and different forward model in Figure 5. In Figure 5(b), the traditional ECT forward model and background data from air-filled sensor indicates the imaging the mixed material under test is not possible. The updated background data and forward model provides a clear distribution of dielectric in Figure 5(c).

4. Dual-modality MIT/ECT

MIT is used to obtain the size and the location of metallic bar in a dual modality solution proposed in this paper. In this dual-modality experiment, separate MIT and ECT sensors measure the samples individually. As mentioned in [7], theoretically MIT is capable to image both conductivity and permittivity, but the in reality MIT has almost no sensitivity to dielectric permittivity. The MIT only focus on visualizing the electrical conductivity distribution and is not affected by dielectric sample. So the information of the metal position from MIT is transferred to the forward model of ECT, and then the space of sensing area excluding the metal is visualized by ECT. The experiment workflow can be explained in Figure 6. The visualized image of MIT may include some errors in location and dimensions. These errors will have an impact on the final results on the image of the dielectric part within the sensing area. Therefore, the error is simulated and corresponding effect is analysed in subsection 4.1.
4.1 Simulation of errors in location and size of metal sample
To identify the size and location of metal, MIT is introduced to image the conductivity distribution. However as a result of the ill-posed and non-linear nature of soft-field tomography and the noise in MIT measurement, the reconstruction image of MIT will have some errors in location and dimensions. Therefore, before combining the ECT with MIT, we simulate errors from MIT images by defining different size and location of metal in ECT forward model. Meanwhile, the capacitance measurements of sample and background kept the same. To analyse the quality of the ECT reconstruction quantitatively, the GREIT figure of merit proposed in [11] is introduced in our test. The figure of merit includes centre of gravity (CoG), resolution and shape deformation (SD). Resolution measures the size of reconstructed targets as a fraction of the medium; SD measures the reconstructed image with threshold setting which does not fit within a circle of an equal area, since reconstruction algorithms typically create circular images for targets.

To obtain each figure of merit, the reconstructed image is transferred to a square gridding of 1000 ×1000 divisions. In this experiment, the iron bar is $R = 2\, \text{cm}$ in radius, and stands at the centre, $(500, 500)$ in coordinate frame. For evaluate the effect of the simulated errors, reference image and figures of merit should be achieved by feeding these practical information of size and location to the forward model of ECT. The reference image is shown in the column of “Distance to centre” = 0 in Table 1, and reference figures of merit are obtained by processing this image.
(1) Simulation of location error
Firstly the real CoG of the wooden sample is obtained. To simulate the error in location, the metallic bar is assumed to approach the real CoG of the wooden sample in a step of 0.1 $R$ in the forward model of ECT. The quality of the image depends on the reconstruction of the dielectric sample. To obtain figures of merits, images are converted to binary images with threshold. In our test, the threshold is set as 80% of the maximum amplitude of reconstructed permittivity matrix. When the radius of the iron bar is $R$, Table 1 indicates the reconstruction image of the wooden sample, as the iron bar is moving towards the CoG of the wooden sample in the forward model of ECT.

(2) Simulation of size error
Since images from MIT can be smaller or larger than the real size of the metallic sample, in our test, the radius of bar is assumed to be 0.8$R$ and 1.2$R$ for the simulation of errors in size. Meanwhile, the error in location is also present while the size errors exist. So the simulation of location error is conducted in different radius set. Table 2 and Table 3 indicate the reconstruction images when the radius is 0.8$R$ and 1.2$R$ respectively.

<table>
<thead>
<tr>
<th>Distance to centre</th>
<th>0</th>
<th>0.1 $R$</th>
<th>0.2 $R$</th>
<th>0.3 $R$</th>
<th>0.4 $R$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Reconstruction</td>
<td><img src="image1.png" alt="Image" /></td>
<td><img src="image2.png" alt="Image" /></td>
<td><img src="image3.png" alt="Image" /></td>
<td><img src="image4.png" alt="Image" /></td>
<td><img src="image5.png" alt="Image" /></td>
</tr>
<tr>
<td>Reconstruction with threshold</td>
<td><img src="image6.png" alt="Image" /></td>
<td><img src="image7.png" alt="Image" /></td>
<td><img src="image8.png" alt="Image" /></td>
<td><img src="image9.png" alt="Image" /></td>
<td><img src="image10.png" alt="Image" /></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>0.5 $R$</th>
<th>0.6 $R$</th>
<th>0.7 $R$</th>
<th>0.8 $R$</th>
<th>0.9 $R$</th>
<th>1.0 $R$</th>
</tr>
</thead>
<tbody>
<tr>
<td><img src="image11.png" alt="Image" /></td>
<td><img src="image12.png" alt="Image" /></td>
<td><img src="image13.png" alt="Image" /></td>
<td><img src="image14.png" alt="Image" /></td>
<td><img src="image15.png" alt="Image" /></td>
<td><img src="image16.png" alt="Image" /></td>
</tr>
</tbody>
</table>

Table 1. Reconstruction images when radius of metal is $R$
<table>
<thead>
<tr>
<th>Distance to centre</th>
<th>0</th>
<th>0.1 $R$</th>
<th>0.2 $R$</th>
<th>0.3 $R$</th>
<th>0.4 $R$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Reconstruction</td>
<td><img src="image1" alt="Reconstruction" /></td>
<td><img src="image2" alt="Reconstruction" /></td>
<td><img src="image3" alt="Reconstruction" /></td>
<td><img src="image4" alt="Reconstruction" /></td>
<td><img src="image5" alt="Reconstruction" /></td>
</tr>
<tr>
<td>Reconstruction with threshold</td>
<td><img src="image6" alt="Reconstruction with threshold" /></td>
<td><img src="image7" alt="Reconstruction with threshold" /></td>
<td><img src="image8" alt="Reconstruction with threshold" /></td>
<td><img src="image9" alt="Reconstruction with threshold" /></td>
<td><img src="image10" alt="Reconstruction with threshold" /></td>
</tr>
</tbody>
</table>

Table 2. Reconstruction images when radius of metal is $0.8R$

<table>
<thead>
<tr>
<th>Distance to centre</th>
<th>0.5 $R$</th>
<th>0.6 $R$</th>
<th>0.7 $R$</th>
<th>0.8 $R$</th>
<th>0.9 $R$</th>
<th>1.0 $R$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Reconstruction</td>
<td><img src="image11" alt="Reconstruction" /></td>
<td><img src="image12" alt="Reconstruction" /></td>
<td><img src="image13" alt="Reconstruction" /></td>
<td><img src="image14" alt="Reconstruction" /></td>
<td><img src="image15" alt="Reconstruction" /></td>
<td><img src="image16" alt="Reconstruction" /></td>
</tr>
<tr>
<td>Reconstruction with threshold</td>
<td><img src="image17" alt="Reconstruction with threshold" /></td>
<td><img src="image18" alt="Reconstruction with threshold" /></td>
<td><img src="image19" alt="Reconstruction with threshold" /></td>
<td><img src="image20" alt="Reconstruction with threshold" /></td>
<td><img src="image21" alt="Reconstruction with threshold" /></td>
<td><img src="image22" alt="Reconstruction with threshold" /></td>
</tr>
</tbody>
</table>

Table 3. Reconstruction images when radius of metal is $1.2R$
In this simulation, the iron bar is supposed to be moved with a maximum distance of $R$ from the centre. According to the results shown in Table 1, Table 2 and Table 3, the images and relevant figures of merit are decaying monotonically while the error in location is increasing. The quantitative analysis of the figures of merit is done below.

1) Resolution change

In Figure 7, the minimum resolution is 16.32%, 15.59% and 14.71%, when the radius changes from $0.8R$ to $R$ and finally to $1.2R$. The decrement in resolution is 11.4%, 15.3% and 20.1% from the reference resolution (18.41%).

2) Shift of the CoG

In Figure 8, the reference CoG of the wooden sample is (589,824). When the metallic bar is moving with the radius of $0.8R$, $R$ and $1.2R$, the maximum shift of CoG is 3.8%, 5.1% and 6.3% of the diameter of the sensing area respectively.

3) SD change

In Figure 9, the reference SD is 3708. The maximum SD reaches 7994, 8358 and 8956, while the radius is $0.8R$, $R$ and $1.2R$ respectively; accordingly the deformation rate is 116%, 125% and 142% from reference.

In addition, when the metallic bar is moving away from the dielectric sample from the centre (500,500), it causes very small change on the reconstruction and figure of merit from our experiments. Similarly, both the images and the figures show a better outcome when the
radius is $0.8R$, where reduction in size results in the same effect caused by movement of the iron bar away from the wooden sample.

### 4.2 Results from MIT and analysis

To investigate the location information from MIT, the real location and the analysed location will be compared. The centre point of the region is set as $(0, 0)$ cm, and the metallic bar is moved from the central point to the eddy in the step of 1 cm along the horizontal axis ($x$-axis). Both the reconstruction images and COGs of the metallic bar are obtained in Table 4 and Figure 10. Normally different threshold settings have varying COGs under a small range. In this experiment, the threshold is set to match the size of the reconstructed metallic bar to the real size.

<table>
<thead>
<tr>
<th>Reconstruction image</th>
<th>Real COG coordinate $(x_0, y_0)$ / (cm)</th>
<th>Analysed location $(x_i, y_i)$ / (cm)</th>
<th>Error rate in distance</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>(0,0)</td>
<td>(0.188,0.188)</td>
<td>13.32%</td>
</tr>
<tr>
<td></td>
<td>(1,0)</td>
<td>(0.797,0.100)</td>
<td>11.32%</td>
</tr>
<tr>
<td>(2,0)</td>
<td>(1.558,0.197)</td>
<td>24.20%</td>
<td></td>
</tr>
<tr>
<td>(3,0)</td>
<td>(2.145,0.221)</td>
<td>44.16%</td>
<td></td>
</tr>
<tr>
<td>(4,0)</td>
<td>(2.704,0.199)</td>
<td>65.57%</td>
<td></td>
</tr>
</tbody>
</table>

Table 4. The table of reconstruction images, analysed COG change and error rate

Figure 10. COG of the metallic bar in real and analysed conditions
In Table 4 both the practical and analysed location of COG are listed. The location error rate is defined as the equation below:

\[
\text{location error rate (LER)} = \frac{||(x_i, y_i) - (x_0, y_0)||}{R} \tag{9}
\]

In section 3, the simulation is conducted under the condition that the metallic bar is shifted from \(0.1R\) to \(1R\) in forward model of ECT, so the coefficient before \(R\) represents the simulated LER. In Figure 10, in comparison to the real COG, the COG from MIT is pulled back to centre along x-axis. If the metallic bar is placed further from centre, the LER is increasingly larger. Due to the symmetric structure of MIT sensor, the shifts happened in y-axis is caused by the inevitable inaccuracy in placing the metallic bar along the x-axis during experiments. These analysed data of COG are inputted to the forward model of ECT. Subsequently the reconstruction images and the figures of merits will judge the quality of dual-modality.

### 4.3 Dual modality results

In this section, the forward model of ECT is separately updated with the real COG and analysed COG that is from MIT for comparison. Then ECT will reconstruct the round wooden sample, which is the red part shown in reconstruction image. To evaluate the quality of reconstruction, figure of merits is calculated. The image with threshold is plotted in the same gridding of \(1000 \times 1000\) as the one in section 4.1.

<table>
<thead>
<tr>
<th>Reconstruction image</th>
<th>Real location</th>
<th>Location from MIT</th>
</tr>
</thead>
<tbody>
<tr>
<td><img src="image1.png" alt="Reconstruction image" /></td>
<td><img src="image2.png" alt="Reconstruction image" /></td>
<td></td>
</tr>
<tr>
<td>Reconstruction image with threshold</td>
<td><img src="image3.png" alt="Reconstruction image with threshold" /></td>
<td><img src="image4.png" alt="Reconstruction image with threshold" /></td>
</tr>
<tr>
<td>COG</td>
<td>(725.3, 734.7)</td>
<td>(730.6, 739.5)</td>
</tr>
<tr>
<td>RES</td>
<td>20.54%</td>
<td>19.85%</td>
</tr>
<tr>
<td>SD</td>
<td>8353</td>
<td>8875</td>
</tr>
</tbody>
</table>

a. Metallic bar stands at (0,0)cm
Table 5. ECT Reconstruction images and the ones with threshold under both real and analysed location in forward model

The image quality is decaying while the metallic bar is moving away from the centre, even though the forward model is defined by the accurate location of the metallic bar. However, in the analysis of result the key point is to show the difference in the quality of images between the two different forward models.
Table 6. The change in figures of merit from the forward model with real location to the analysed one.

<table>
<thead>
<tr>
<th></th>
<th>(0,0)cm</th>
<th>(1,0)cm</th>
<th>(2,0)cm</th>
<th>(3,0)cm</th>
<th>(4,0)cm</th>
</tr>
</thead>
<tbody>
<tr>
<td>ΔCOG (unit length)</td>
<td>7.15</td>
<td>2.94</td>
<td>3.12</td>
<td>14.23</td>
<td>25.54</td>
</tr>
<tr>
<td>ΔRES</td>
<td>-3.36%</td>
<td>0.19%</td>
<td>1.81%</td>
<td>-3.4%</td>
<td>24.79%</td>
</tr>
<tr>
<td>ΔSD</td>
<td>6.25%</td>
<td>-0.60%</td>
<td>-2.60%</td>
<td>-13.57%</td>
<td>84.08%</td>
</tr>
</tbody>
</table>

Since the change in COG in Table 6 is from the gridding of 1000×1000 unit length, the maximum shift of COG is only 25.54/1000=2.554% of the diameter of the imaging region, which is very small. The change in RES and SD stays in a low percentage until the metallic bar is placed at (3, 0) cm. From the images in Table 5 (d) and (e), with the real forward model applied, the reconstruction of round wooden stick has seriously deformed, therefore the factor of SD is not accurate enough. When the reference image is largely deformed, it is not applicable to use SD to describe the image quality and the 84.08% change in SD cannot express the real change in the image as we can see in Table 5(e).

5. Conclusions

The purpose of this paper is to introduce a new method for ECT imaging with ground conductors assisted by the MIT imaging. The dual-modality can detect the existence of highly conductive sample and use the location information to improve the image reconstruction of ECT. The simulations and the experiments are conducted in various imaging scenarios, and the images are assessed by figure of merits: COG, RES and SD. The result from dual-modality tomography is promising and can be applied to test complex samples with metallic and dielectric material. For example, detection cracks in reinforced concrete, where reinforcing steel bars in concrete is highly conductive and electrically grounded. Depending on location of the metallic object, the ECT forward problem needs to be solved and an updated sensitivity map needs to be generated. In dynamical imaging situations where online image reconstruction may be needed further work is needed to speed up the computational time, in particular those of the ECT system. For purpose of this study an ECT system and an MIT system was used, a future hybrid ECT-MIT system can be developing allowing for some sharing hardware and structure, which can help reduce the cost of hybrid system. In our future study we will extend the dual modality ECT, MIT imaging by adapting a complex impedance forward and inverse model for ECT where we can deal with floated conductors as well as low conductivity inclusions such as water contents.
References: