

# Region of Interest Guided Stimulation Pattern Selection Strategy for Electrical Impedance Tomography

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**Abstract:** The proposed method identifies an optimal set from all possible combinations of stimulation patterns for a fixed number of electrodes. The reconstructions using the optimal set achieve better localization within the region of interest compared to the common stimulation patterns.

## 1 Introduction

In electrical impedance tomography, stimulation patterns highly influence the attainable spatial resolution and distinguishability in a reconstruction [3]. Typically, these patterns are decided based on personal experiences and simulations. In this paper, we propose a methodology to select stimulation patterns such that we obtain high quality reconstructions in our region of interest (ROI).

## 2 Method

The proposed method has three steps. Given  $N$  electrodes, the number of possible pair-drive stimulation configurations are  $N(N-1)/2$ . We quantify the importance of each pair-drive for reconstructing a point target  $x$  (one mesh-element) in the ROI by calculating the goodness score

$$GS_k = \|\Sigma_k V_k^\top x\|_2 \quad (1)$$

where we use the SVD decomposition of the sensitivity matrix defined for the  $k^{th}$  pair-drive, *i.e.*,  $J_k = U_k \Sigma_k V_k^\top$ . In (1), we measure how well the basis functions represent the point-target, weighted with the singular values.

Now, we build an optimal stimulation set starting with the highest scored pair-drive. We add high scoring pair-drives one by one in the set until the criteria holds:

$$\text{Rank}(\bar{J}) \leq \lfloor s \text{Rank}(J) \rfloor \quad (2)$$

where the sensitivity matrix  $\bar{J}$  is built with pair-drives in the optimal set and  $J$  with all  $N(N-1)/2$  pair-drives. At each step, we eliminate high scoring pair-drives if they do not increase the rank of  $\bar{J}$ . It is likely that high scoring pair-drive sensitivity matrices are linearly dependent to each other.

The tuning parameter  $s \in (0, 1]$  enables us to choose only the most important pair-drives if a high frame rate is desired for data acquisition. For  $s = 1$ , the optimal  $\bar{J}$  is similar to  $J$  in the sense of the number of linearly independent vectors in the sensitivity matrix. The value of  $s$  can also be adjusted by analyzing the singular values of  $J$ .

In the last step, we repeat the first two steps for a number of targets covering the entire ROI. The union of optimal sets, identified for each point-target, is the final optimal set that we use for reconstruction of an object in the ROI.

## 3 Experiments

We define an ellipsoid ROI of semi-axis (0.2, 0.2, 0.5)cm at origin (0, 0.5, 0) inside a cylinder of radius 1cm and height 1.4cm. Total  $N = 2 \times 16$  electrodes are placed at two planes  $z = \pm 0.3$ cm. We take single-ended measurements with respect to a reference electrode for each stimulus pair. We ap-

ply current stimulus at all electrodes but the reference electrode. Hence, the total stimuli are  $S = (N-1)(N-2)/2$ .

We reconstruct a 2D slice at  $z = 0$  of a spherical target with 10% contrast of radius 0.15cm placed at  $y = 0.5$ cm and multiple heights  $h$ ; see Fig. 1 for reconstructions with full stimulus patterns, popular skip-4 square pattern, and three patterns obtained with the proposed approach.

Measurements are with noise of variance  $10^{-8}$ . We calculate regularization parameters  $\lambda$  using the image signal to noise ratio (SNR) measure with  $\text{SNR} = 0.5$  [1].

The reconstructions are highly localized near to the reconstruction plane with both full and the proposed stimuli compared to the skip-4 pattern; as also illustrated through the GREIT resolution measure (lower is better) [2]. But, with the target further away from the reconstruction plane, performance decreases with a reduced number of patterns.

h from center	0.0	0.2	0.4	0.6
full pattern, S = 465 $\lambda = 0.0738$				
resolution	0.366	0.371	0.375	0.392
skip-4 square, S = 30 $\lambda = 0.0769$				
resolution	0.525	0.505	0.479	0.491
$s = 1.0$ , S = 140 $\lambda = 0.0737$				
resolution	0.40	0.397	0.402	0.435
$s = 0.6$ , S = 81 $\lambda = 0.0737$				
resolution	0.424	0.397	0.399	0.417
$s = 0.2$ , S = 29 $\lambda = 0.0727$				
resolution	0.488	0.428	0.434	0.455

**Figure 1:** Reconstructions with five stimulation patterns. Black circle represents the ground-truth target position.

## 4 Conclusion

We proposed a method to obtain stimulation patterns targeting a specific ROI. Our results shows that highly localized reconstructions are possible through optimally selected stimulation patterns. Future work involves improving the method by removing redundancy in the optimal set.

## References

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