



UNIVERSITY  
OF MANITOBA



# SPATIAL PRIOR FOR QUANTITATIVE BREAST CANCER MICROWAVE IMAGING: A COMPARISON BETWEEN NON-ITERATIVE EIGENFUNCTION-BASED INVERSION AND SAMPLING METHODS

*Martina T. Bevacqua<sup>1</sup>, Nasim Abdollahi<sup>2</sup>, Ian Jeffrey<sup>2</sup>,  
Tommaso Isernia<sup>1</sup>, Joe LoVetri<sup>2</sup>*

<sup>1</sup>DIIES, Università Mediterranea di Reggio Calabria, Italy

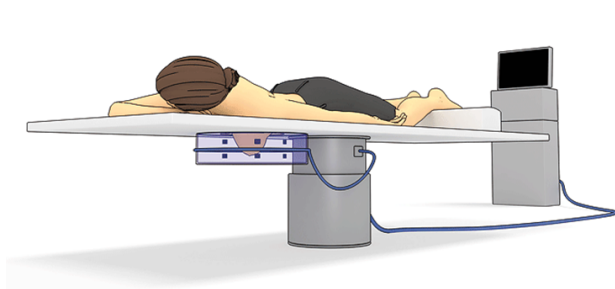
<sup>2</sup>Electrical and Computer Engineering Dept. University of Manitoba, Winnipeg, Canada

- 1. Introduction to Biomedical Microwave Imaging**
- 2. Inverse Scattering Problem**
- 3. Qualitative Methods for generating Spatial Priors**
- 4. Incorporation of Spatial Priors within Contrast Source Inversion**
- 5. 2D Example via Discontinuous Galerkin Method CSI**

- 1. Introduction to Biomedical Microwave Imaging**
2. Inverse Scattering Problem
3. Qualitative Methods for generating Spatial Priors
4. Incorporation of Spatial Priors within Contrast Source Inversion
5. 2D Example via Discontinuous Galerkin Method CSI

# Biomedical Microwave Imaging

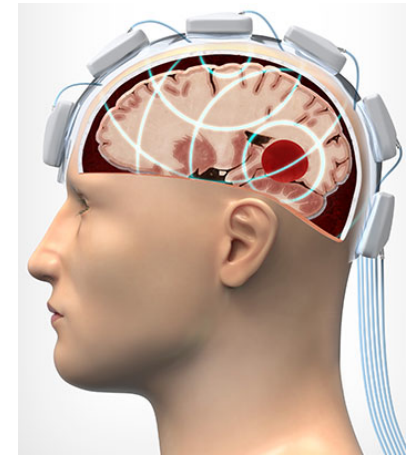
Microwave Imaging is very attractive as a cooperative diagnostic technique.



Breast cancer imaging



Bone fracture risk monitoring



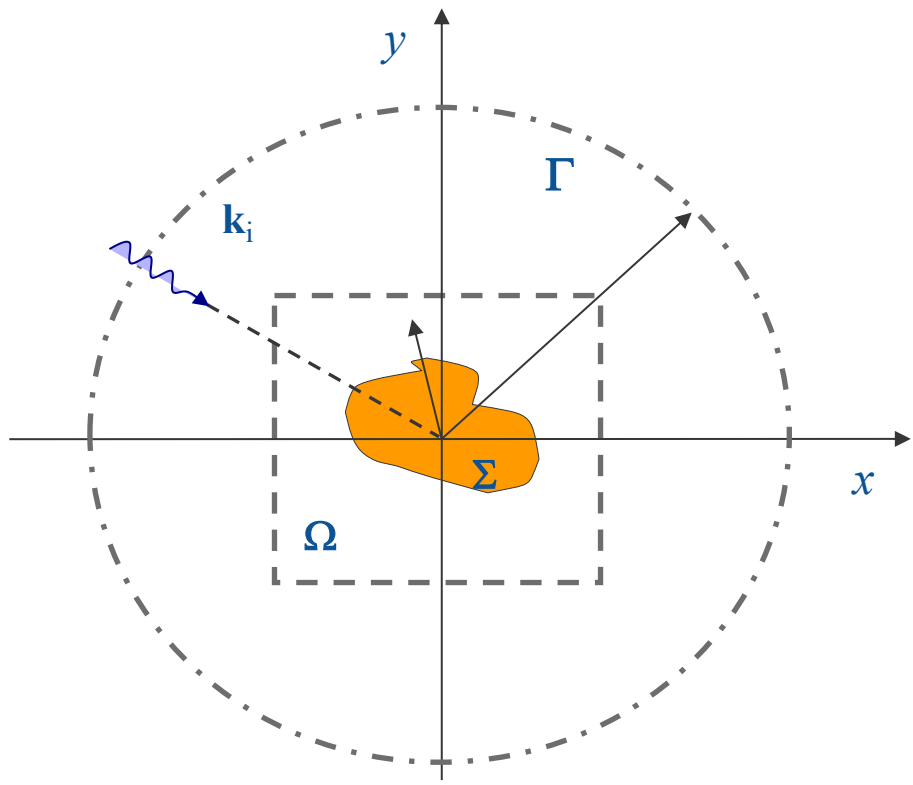
Brain stroke monitoring

- ✓ non-ionizing radiations
- ✓ low-cost and portable devices

- X Resolution
- X Solution of an inverse scattering problem

1. Introduction to Biomedical Microwave Imaging
- 2. Inverse Scattering Problem**
3. Qualitative Methods for generating Spatial Priors
4. Incorporation of Spatial Priors within Contrast Source Inversion
5. 2D Example via Discontinuous Galerkin Method CSI

# Inverse Scattering Problem (2D scalar case)



$$E_i(\mathbf{r}, \mathbf{r}_t), E_s(\mathbf{r}_m, \mathbf{r}_t)$$
$$\mathbf{r} \in \Omega, \mathbf{r}_t, \mathbf{r}_m \in \Gamma$$



$$\chi(\mathbf{r}) = \frac{\varepsilon(\mathbf{r})}{\varepsilon_b} - 1$$

contrast function, which encodes  
target properties  
(e.m. parameters, shape)

# Inverse Scattering Problem (2D scalar case)

$$E_s(\mathbf{r}_m, \mathbf{r}_t) = \int_{\Omega} G_b(\mathbf{r}_m, \mathbf{r}') \chi(\mathbf{r}') E_{tot}(\mathbf{r}', \mathbf{r}_t) d\mathbf{r}' = \mathcal{A}_e[\chi E_{tot}(\mathbf{r}, \mathbf{r}_t)]$$

*'data' equation*

$$W(\mathbf{r}, \mathbf{r}_t) - \chi(\mathbf{r}) E_i(\mathbf{r}, \mathbf{r}_t) = \chi(\mathbf{r}) \int_{\Omega} G_b(\mathbf{r}, \mathbf{r}') W(\mathbf{r}', \mathbf{r}_t) d\mathbf{r}' = \chi(\mathbf{r}) \mathcal{A}_i[\chi E_{tot}(\mathbf{r}, \mathbf{r}_t)]$$

*'state' equation*

**non-linear and ill-posed inverse problem**

1. Introduction to Biomedical Microwave Imaging
2. Inverse Scattering Problem
- 3. Qualitative Methods for generating Spatial Priors**
4. Incorporation of Spatial Priors within Contrast Source Inversion
5. 2D Example via Discontinuous Galerkin Method CSI



# Qualitative methods

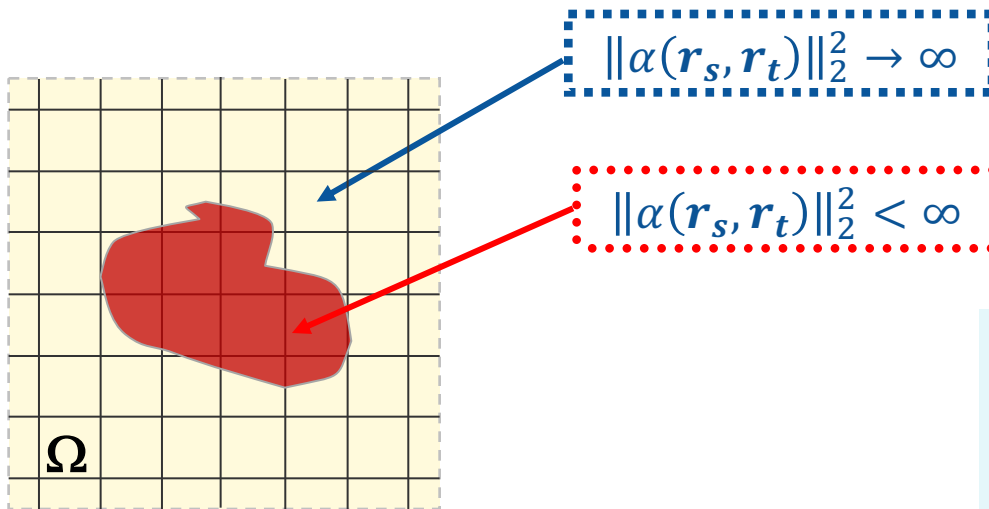
- **LOCATION AND SHAPE RECONSTRUCTION**
  - sufficient in several applications
  - useful prior information
- **A HIGH FLEXIBILITY**
  - dielectric and metallic objects
  - no approximations or prior information
- **COMPUTATIONAL EFFICIENCY**
  - straightforward implementation
  - “quasi-real” time execution

# Linear Sampling Method

In each point  $\mathbf{r}_s \in \Omega$ , LSM solves the far-field equation:

$$\int_{\Gamma} \alpha(\mathbf{r}_s, \mathbf{r}_t) E_s(\mathbf{r}_m, \mathbf{r}_t) d\mathbf{r}_t = G_b(\mathbf{r}_m, \mathbf{r}_s)$$

The energy of the (regularized) solution allows to retrieve the target support.



$$\mathfrak{S}_{LSM}(\mathbf{r}_s) = \|\alpha(\mathbf{r}_s, \mathbf{r}_t)\|_2^2$$

**LSM support indicator**

[\*] I. Catapano, L. Crocco, and T. Isernia, On simple methods for shape reconstruction of unknown scatterers, IEEE TAP, 2007

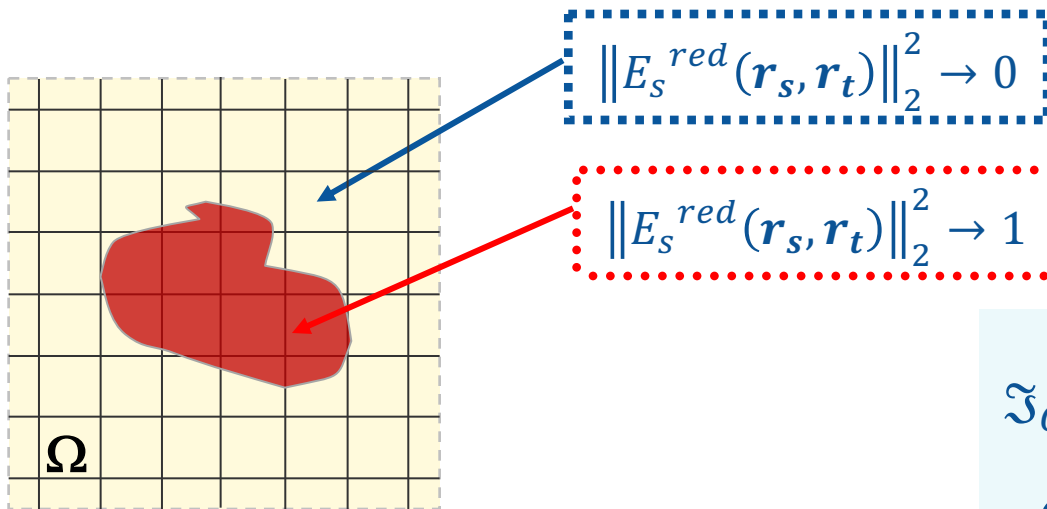
# Orthogonality Sampling Method

OSM tests the orthogonality relationship between the far-field pattern  $E_s^\infty$  and the Green's function  $G_b^\infty$ .

$$E_s^{red}(\mathbf{r}_s, \mathbf{r}_t) = \frac{1}{\gamma} \langle E_s^\infty, G_b^\infty \rangle_{\Gamma(\hat{\mathbf{r}}_m)}$$

$\gamma$  is a constant

The energy of  $E_s^{red}$  allows to retrieve the target support.



$$\mathfrak{S}_{OSM}(\mathbf{r}_s) = \|E_s^{red}(\mathbf{r}_s, \mathbf{r}_t)\|_2^2$$

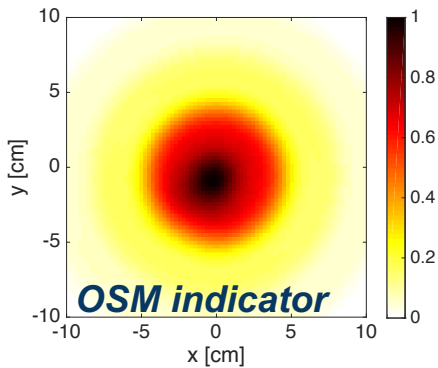
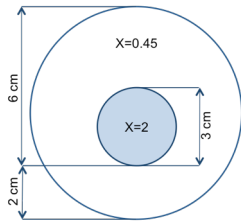
**OSM support indicator**

# Orthogonality Sampling Method

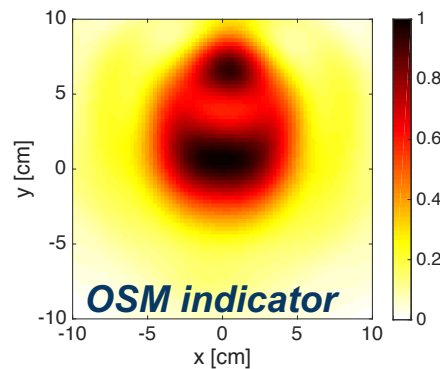
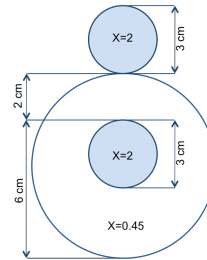
Interestingly, the energy of  $E_s^{red}$  allows to identify discontinuities within the target.

## Fresnel experimental data inversions

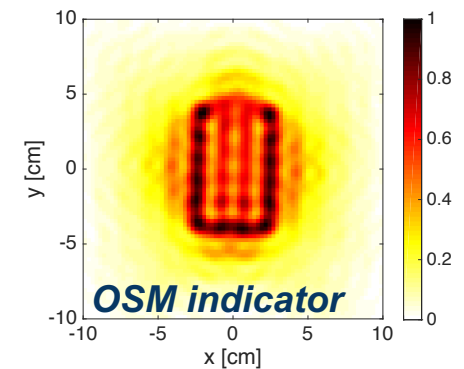
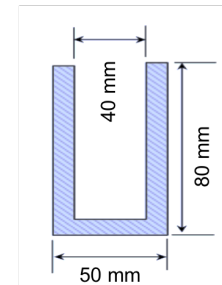
FoamDiellInt, f=5GHz



FoamTwinDiellInt, f=4GHz



U-shaped target, f=12GHz



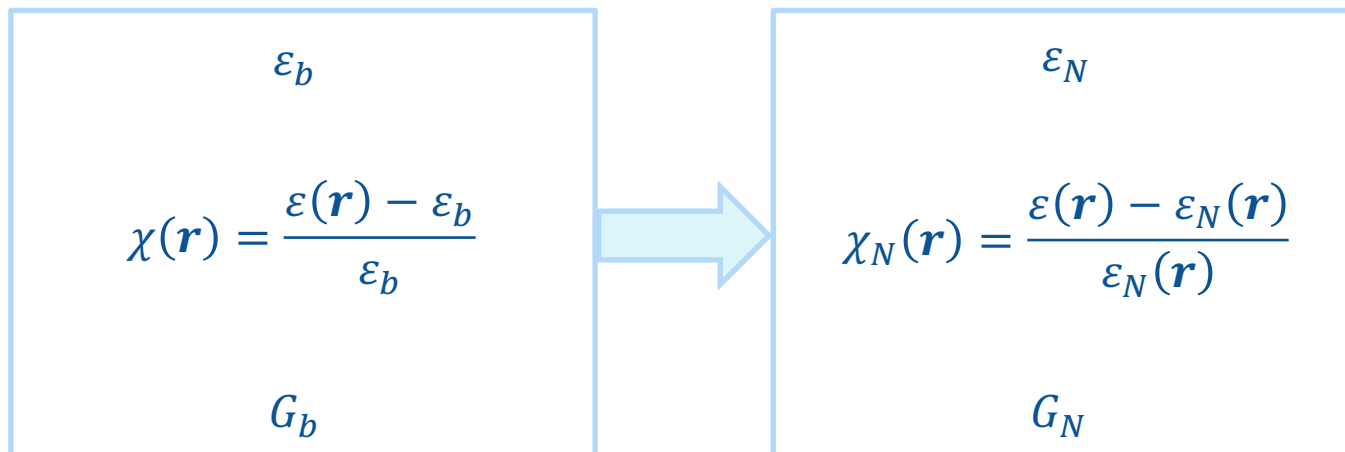
1. Introduction to Biomedical Microwave Imaging
2. Inverse Scattering Problem
3. Qualitative Methods for generating Spatial Priors
- 4. Incorporation of Spatial Priors within Contrast Source Inversion (CSI)**
5. 2D Example via Discontinuous Galerkin Method CSI

# Incorporation of Spatial Priors within CSI

In Contrast Source Inversion (CSI),  $\chi$  and  $W$  are estimated by minimizing:

$$\Phi(W, \chi) = \sum_t \frac{\|\chi E_i(\mathbf{r}_t) + \chi A_i[W(\mathbf{r}_t)] - W(\mathbf{r}_t)\|_2^2}{\|E_i(\mathbf{r}_t)\|_2^2} + \sum_t \frac{\|E_s(\mathbf{r}_t) - A_e[W(\mathbf{r}_t)]\|_2^2}{\|E_s(\mathbf{r}_t)\|_2^2}$$

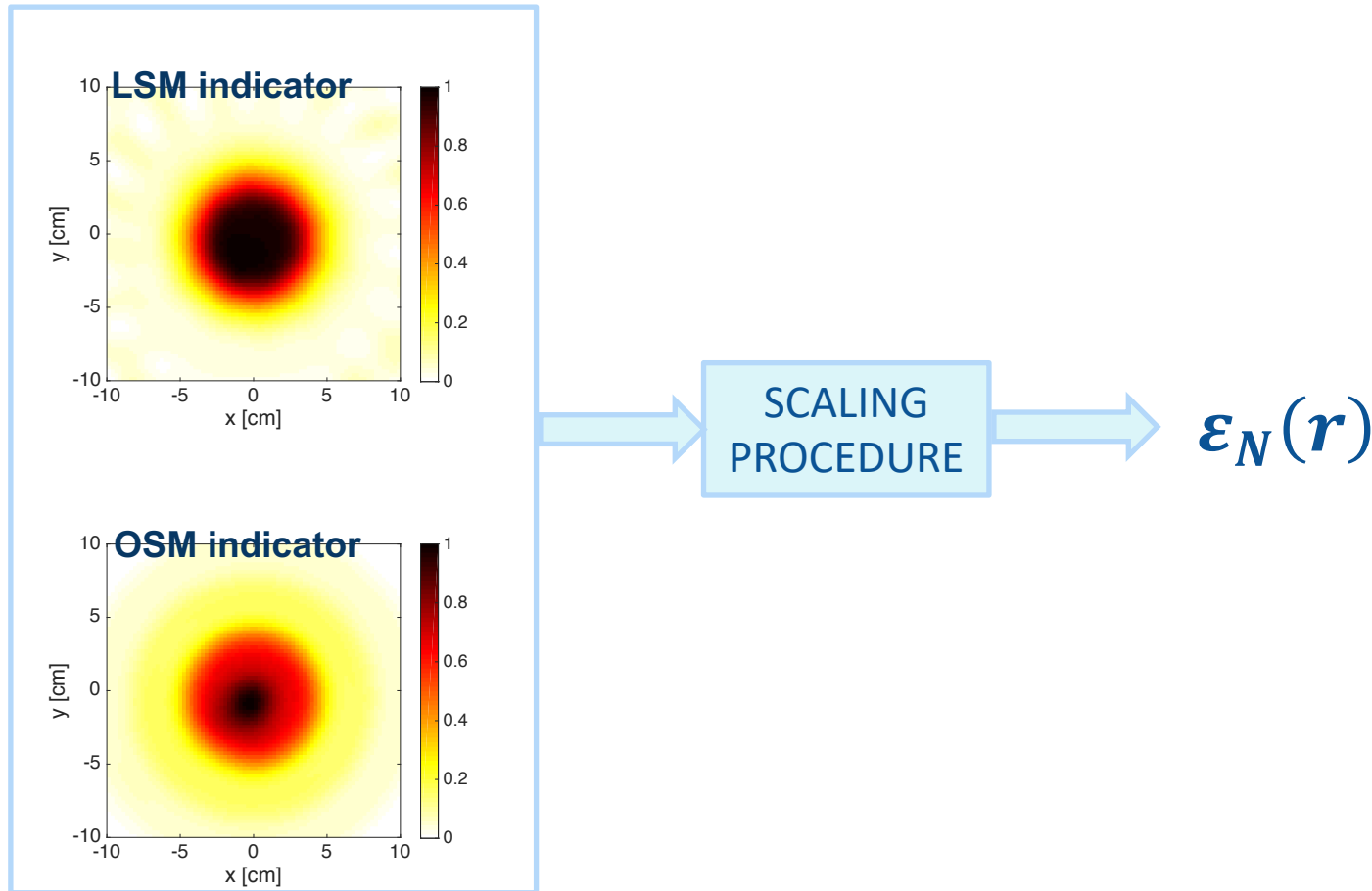
To integrate the prior information in CSI\*:



where  $G_N$  corresponds to the inhomogeneous background  $\varepsilon_N$ .

# LSM and OSM for generating Spatial Prior Information

Different spatial priors  $\epsilon_N(\mathbf{r})$  can be generated by scaling the support indicators according to the expected value of permittivity and conductivity of the breast tissues.

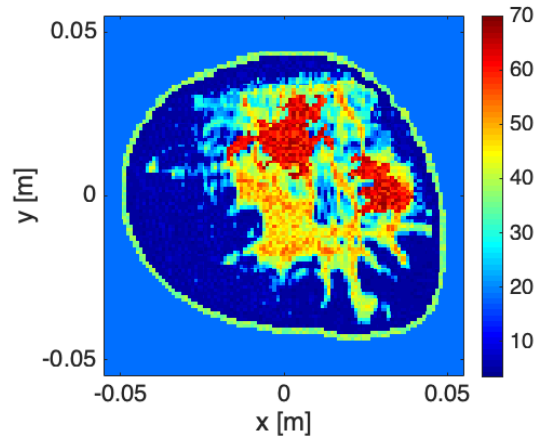


1. Introduction to Biomedical Microwave Imaging
2. Inverse Scattering Problem
3. Qualitative Methods for generating Spatial Priors
4. Incorporation of Spatial Priors within Contrast Source Inversion
5. **2D Example via Discontinuous Galerkin Method CSI**

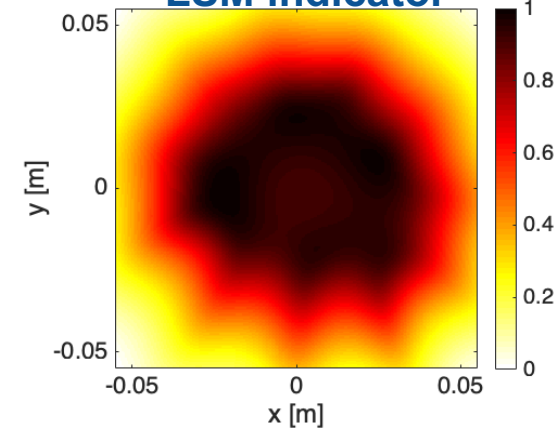


# 2D numerical example

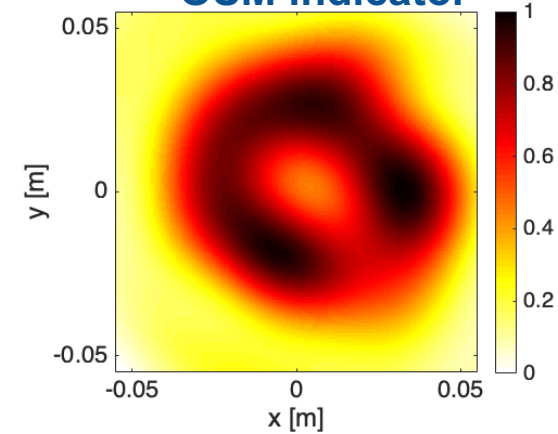
**Permittivity of the adopted breast model\***



**LSM indicator**



**OSM indicator**

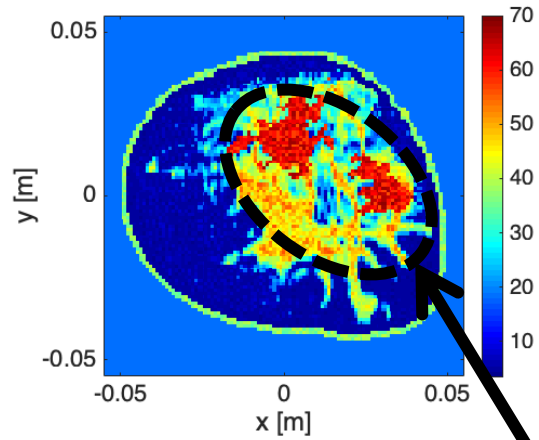


**INFO:**  $freq = 1\text{ GHz}$ ,  $R = 13\text{ cm}$ ,  $T = M = 24$ ,  $SNR = 30\text{ dB}$ ,  $\epsilon_b = 18$ ,  $\sigma_b = 0$

[\*] Anthropomorphic breast model repository for research and development of microwave breast imaging technologies. Scientific data, 2018.

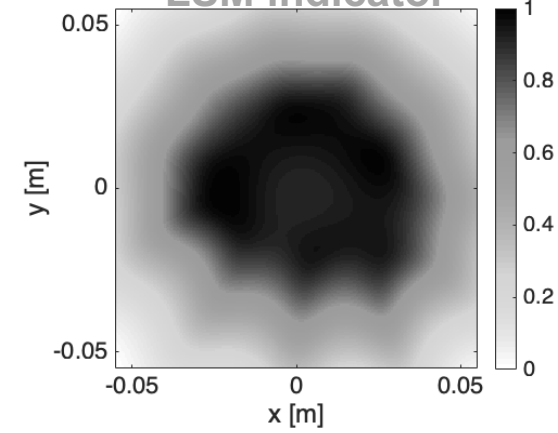
# 2D numerical example

**Permittivity of the adopted breast model\***

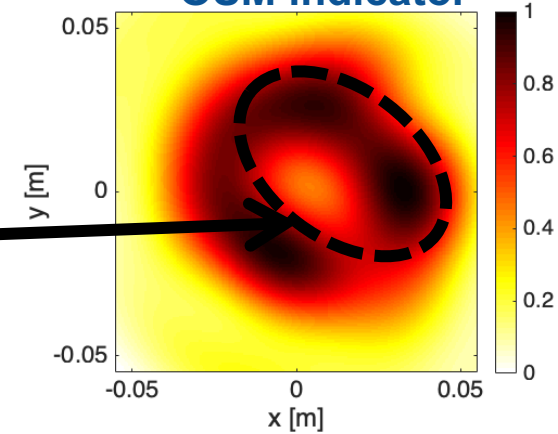


**Denser parts of the breast**

**LSM indicator**



**OSM indicator**

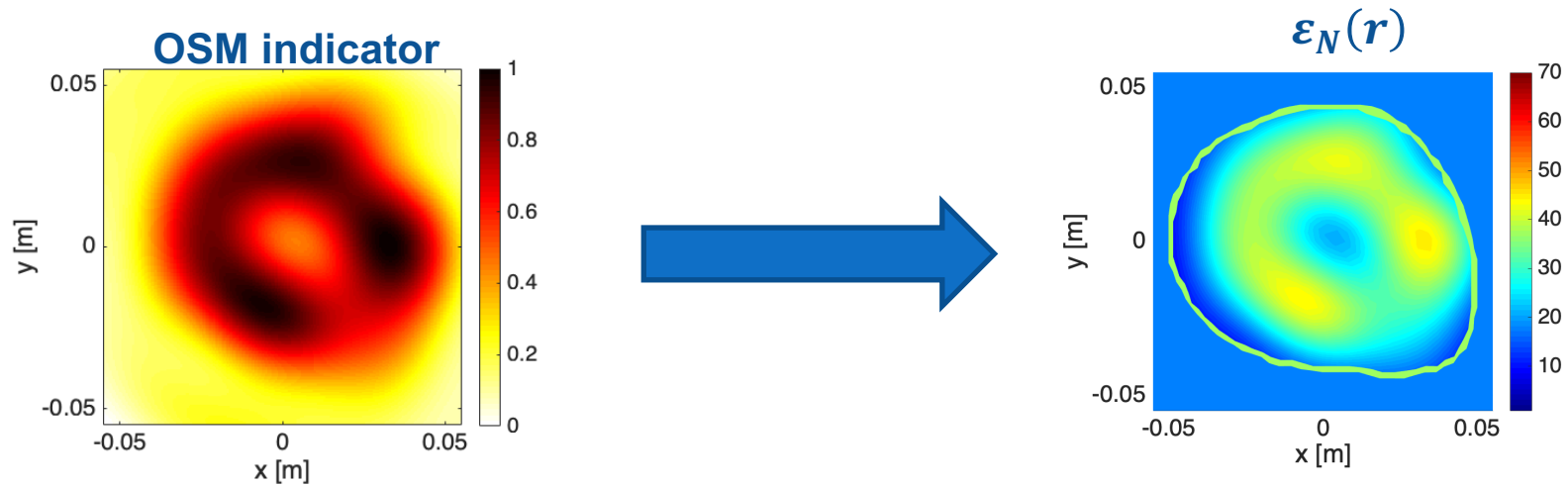


**INFO:**  $freq = 1\text{ GHz}$ ,  $R = 13\text{ cm}$ ,  $T = M = 24$ ,  $SNR = 30\text{ dB}$ ,  $\epsilon_b = 18$ ,  $\sigma_b = 0$

[\*] Anthropomorphic breast model repository for research and development of microwave breast imaging technologies. Scientific data, 2018.

# 2D numerical example

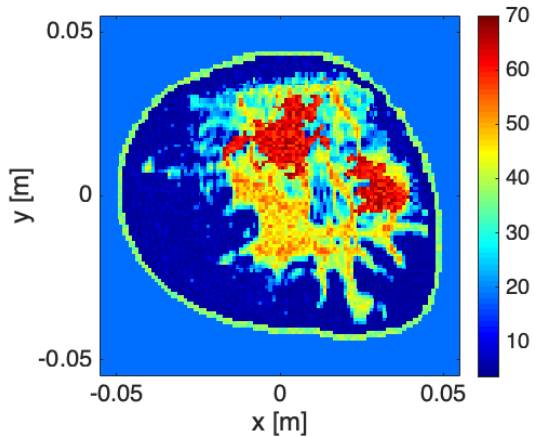
The normalized OSM maps is modulated in amplitude according to the expected values of permittivity and conductivity of the breast tissues.



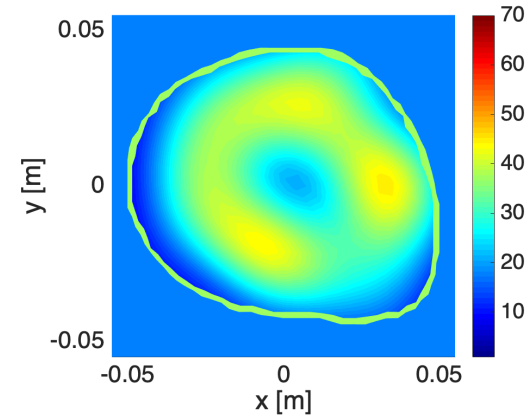
**INFO:**  $freq = 1\text{ GHz}$ ,  $R = 13\text{ cm}$ ,  $T = M = 24$ ,  $SNR = 30\text{ dB}$ ,  $\epsilon_b = 18$ ,  $\sigma_b = 0$

# 2D numerical example

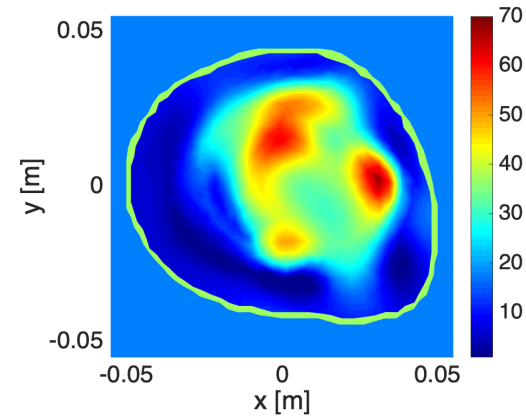
**Permittivity of the adopted breast model\***



$\epsilon_N(\mathbf{r})$



**Retrieved permittivity**

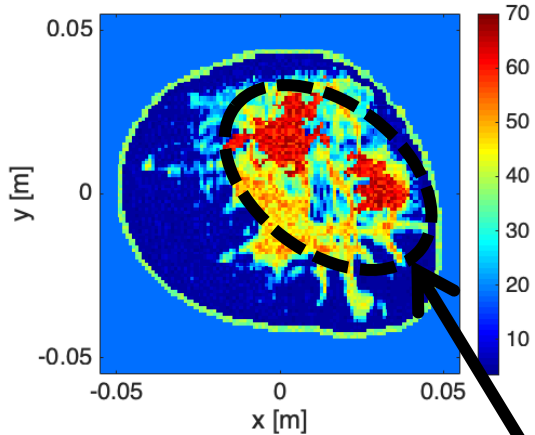


**INFO:**  $freq = 1\text{ GHz}$ ,  $R = 13\text{ cm}$ ,  $T = M = 24$ ,  $SNR = 30\text{ dB}$ ,  $\epsilon_b = 18$ ,  $\sigma_b = 0$

[\*] Anthropomorphic breast model repository for research and development of microwave breast imaging technologies. Scientific data, 2018.

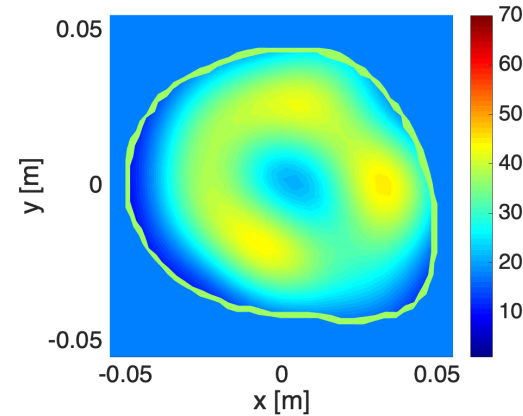
# 2D numerical example

**Permittivity of the adopted breast model\***

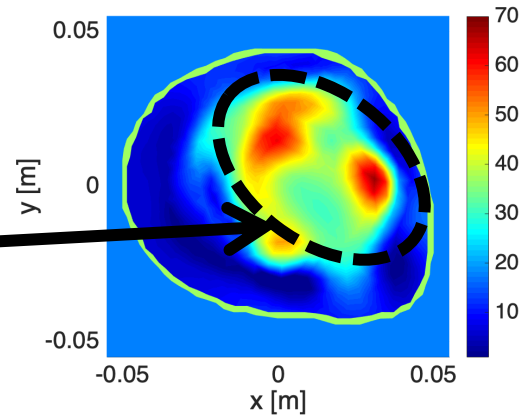


**Denser parts of the breast**

$\epsilon_N(r)$



**Retrieved permittivity**



**INFO:**  $freq = 1 \text{ GHz}$ ,  $R = 13 \text{ cm}$ ,  $T = M = 24$ ,  $SNR = 30\text{dB}$ ,  $\epsilon_b = 18$ ,  $\sigma_b = 0$

[\*] Anthropomorphic breast model repository for research and development of microwave breast imaging technologies. Scientific data, 2018.

- Spatial priors to improve both the accuracy and resolution of microwave imaging reconstructions.
- A combination of linear/orthogonal sampling method maps to generate spatial priors in a simple and fast way.
- Incorporation with the CSI as inhomogeneous background.
- Preliminary 2D inversion against a heterogeneously dense breast model.
- Future works: comparisons with respect to the quantitative NIEI eigenfunction prior.\*



UNIVERSITY  
OF MANITOBA



# SPATIAL PRIOR FOR QUANTITATIVE BREAST CANCER MICROWAVE IMAGING: A COMPARISON BETWEEN NON-ITERATIVE EIGENFUNCTION-BASED INVERSION AND SAMPLING METHODS

*Martina T. Bevacqua<sup>1</sup>, Nasim Abdollahi<sup>2</sup>, Ian Jeffrey<sup>2</sup>,  
Tommaso Isernia<sup>1</sup>, Joe LoVetri<sup>2</sup>*

<sup>1</sup>DIIES, Università Mediterranea di Reggio Calabria, Italy

<sup>2</sup>Electrical and Computer Engineering Dept. University of Manitoba, Winnipeg, Canada