# Reconstruction of Sparse Scatterers through an Approximated Compressive Sensing Strategy

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## Abstract

This report proposes a numerical validation of the Bayesian Compressive Sampling-based technique for imaging dielectric cylinders within the conditions of "sparsity" and "weakness" of the scatterers. Various shake of the scatterers and different values of dielectric permittivity have been considered.

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## 1 TEST CASE: Two Square Cylinders

**GOAL:** show the performances of BCS when dealing with a sparse scatterer

- Number of Views: V
- Number of Measurements: M
- Number of Cells for the Inversion: N
- Number of Cells for the Direct solver: D
- Side of the investigation domain: L

### **Test Case Description**

#### Direct solver:

- Square domain divided in  $\sqrt{D} \times \sqrt{D}$  cells
- Domain side:  $L = 3\lambda$
- D = 1296 (discretization for the direct solver:  $< \lambda/10$ )

## Investigation domain:

- Square domain divided in  $\sqrt{N} \times \sqrt{N}$  cells
- $L = 3\lambda$
- $2ka = 2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2} = 6\pi\sqrt{2} = 26.65$
- $\#DOF = \frac{(2ka)^2}{2} = \frac{(2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2})^2}{2} = 4\pi^2 \left(\frac{L}{\lambda}\right)^2 = 4\pi^2 \times 9 \approx 355.3$
- N scelto in modo da essere vicino a #DOF: N = 324 (18 × 18)

## Measurement domain:

- Measurement points taken on a circle of radius  $\rho = 3\lambda$
- Full-aspect measurements
- $M \approx 2ka \rightarrow M = 27$

### Sources:

- Plane waves
- $V \approx 2ka \rightarrow V = 27$
- Amplitude A = 1
- Frequency: 300 MHz ( $\lambda = 1$ )

### **Object:**

- Two square cylinders of side  $\frac{\lambda}{6} = 0.1667$
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$  (one square),  $\varepsilon_r = 1.9$  (one square)
- $\sigma = 0 [S/m]$

#### **BCS** parameters:

- Initial estimate of the noise:  $n_0 = 1.0 \times 10^{-3}$
- Convergenze parameter:  $\tau = 1.0 \times 10^{-8}$



Figure 17. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

Ricostruzioni molto buone per tutti i valori di SNR.



Figure 18. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

Ricostruzioni in generale buone per tutti i valori di SNR.



Figure 19. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

Ricostruzioni in generale buone per tutti i valori di SNR.



Figure 20. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

Ricostruzioni in generale buone per tutti i valori di SNR.

## **RESULTS:** Error Figures



**Figure 21.** Behaviour of error figures as a function of  $\varepsilon_r$ , for different SNR values: (a) total error  $\xi_{tot}$ , (b) internal error  $\xi_{int}$ , (c) external error  $\xi_{ext}$ .

## 2 TEST CASE: Three Square Cylinders

**GOAL:** show the performances of BCS when dealing with a sparse scatterer

- Number of Views: V
- Number of Measurements: M
- Number of Cells for the Inversion: N
- Number of Cells for the Direct solver: D
- Side of the investigation domain: L

### **Test Case Description**

#### Direct solver:

- Square domain divided in  $\sqrt{D} \times \sqrt{D}$  cells
- Domain side:  $L = 3\lambda$
- D = 1296 (discretization for the direct solver:  $< \lambda/10$ )

## Investigation domain:

- Square domain divided in  $\sqrt{N} \times \sqrt{N}$  cells
- $L = 3\lambda$
- $2ka = 2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2} = 6\pi\sqrt{2} = 26.65$
- $\#DOF = \frac{(2ka)^2}{2} = \frac{(2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2})^2}{2} = 4\pi^2 \left(\frac{L}{\lambda}\right)^2 = 4\pi^2 \times 9 \approx 355.3$
- N scelto in modo da essere vicino a #DOF: N = 324 (18 × 18)

## Measurement domain:

- Measurement points taken on a circle of radius  $\rho = 3\lambda$
- Full-aspect measurements
- $M \approx 2ka \rightarrow M = 27$

### Sources:

- Plane waves
- $V \approx 2ka \rightarrow V = 27$
- Amplitude A = 1
- Frequency: 300 MHz ( $\lambda = 1$ )

#### **Object:**

- Three square cylinders of side  $\frac{\lambda}{6} = 0.1667$
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$  (two squares),  $\varepsilon_r = 1.9$  (one square)
- $\sigma = 0 [S/m]$

## **BCS** parameters:

- Initial estimate of the noise:  $n_0 = 1.0 \times 10^{-3}$
- Convergenze parameter:  $\tau = 1.0 \times 10^{-8}$



Figure 22. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**



Figure 23. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**



Figure 24. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**



Figure 25. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

## **RESULTS:** Error Figures



**Figure 26.** Behaviour of error figures as a function of  $\varepsilon_r$ , for different SNR values: (a) total error  $\xi_{tot}$ , (b) internal error  $\xi_{int}$ , (c) external error  $\xi_{ext}$ .

## 3 TEST CASE: Four Square Cylinders

**GOAL:** show the performances of BCS when dealing with a sparse scatterer

- Number of Views: V
- Number of Measurements: M
- Number of Cells for the Inversion: N
- Number of Cells for the Direct solver: D
- Side of the investigation domain: L

### **Test Case Description**

#### Direct solver:

- Square domain divided in  $\sqrt{D} \times \sqrt{D}$  cells
- Domain side:  $L = 3\lambda$
- D = 1296 (discretization for the direct solver:  $< \lambda/10$ )

## Investigation domain:

- Square domain divided in  $\sqrt{N} \times \sqrt{N}$  cells
- $L = 3\lambda$
- $2ka = 2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2} = 6\pi\sqrt{2} = 26.65$
- $\#DOF = \frac{(2ka)^2}{2} = \frac{(2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2})^2}{2} = 4\pi^2 \left(\frac{L}{\lambda}\right)^2 = 4\pi^2 \times 9 \approx 355.3$
- N scelto in modo da essere vicino a #DOF: N = 324 (18 × 18)

## Measurement domain:

- Measurement points taken on a circle of radius  $\rho = 3\lambda$
- Full-aspect measurements
- $M \approx 2ka \rightarrow M = 27$

### Sources:

- Plane waves
- $V \approx 2ka \rightarrow V = 27$
- Amplitude A = 1
- Frequency: 300 MHz ( $\lambda = 1$ )

### **Object:**

- Four square cylinders of side  $\frac{\lambda}{6} = 0.1667$
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$  (two squares),  $\varepsilon_r = 1.9$  (two square)
- $\sigma = 0 [S/m]$

## **BCS** parameters:

- Initial estimate of the noise:  $n_0 = 1.0 \times 10^{-3}$
- Convergenze parameter:  $\tau = 1.0 \times 10^{-8}$



Figure 27. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**



Figure 28. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**



Figure 29. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**



Figure 30. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

## **RESULTS:** Error Figures



**Figure 31.** Behaviour of error figures as a function of  $\varepsilon_r$ , for different SNR values: (a) total error  $\xi_{tot}$ , (b) internal error  $\xi_{int}$ , (c) external error  $\xi_{ext}$ .

## **Observations:**

## 4 TEST CASE: Cross-Shaped Cylinder

**GOAL:** show the performances of BCS when dealing with a sparse scatterer

- Number of Views: V
- Number of Measurements: M
- Number of Cells for the Inversion: N
- Number of Cells for the Direct solver: D
- Side of the investigation domain: L

#### **Test Case Description**

#### Direct solver:

- Square domain divided in  $\sqrt{D} \times \sqrt{D}$  cells
- Domain side:  $L = 3\lambda$
- D = 1296 (discretization for the direct solver:  $< \lambda/10$ )

## Investigation domain:

- Square domain divided in  $\sqrt{N} \times \sqrt{N}$  cells
- $L = 3\lambda$
- $2ka = 2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2} = 6\pi\sqrt{2} = 26.65$
- $\#DOF = \frac{(2ka)^2}{2} = \frac{(2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2})^2}{2} = 4\pi^2 \left(\frac{L}{\lambda}\right)^2 = 4\pi^2 \times 9 \approx 355.3$
- N scelto in modo da essere vicino a #DOF: N = 324 (18 × 18)

## Measurement domain:

- Measurement points taken on a circle of radius  $\rho = 3\lambda$
- Full-aspect measurements
- $M \approx 2ka \rightarrow M = 27$

### Sources:

- Plane waves
- $V \approx 2ka \rightarrow V = 27$
- Amplitude A = 1
- Frequency: 300 MHz ( $\lambda = 1$ )

#### **Object:**

- Cross-shaped cylinder
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$
- $\sigma = 0 [S/m]$

## **BCS** parameters:

- Initial estimate of the noise:  $n_0 = 1.0 \times 10^{-3}$
- Convergenze parameter:  $\tau = 1.0 \times 10^{-8}$



Figure 32. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

Ricostruzioni abbastanza buone solo per $\varepsilon_r=1.5.$ 



Figure 33. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].



Figure 34. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].



Figure 35. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **RESULTS:** Error Figures



**Figure 36.** Behaviour of error figures as a function of  $\varepsilon_r$ , for different SNR values: (a) total error  $\xi_{tot}$ , (b) internal error  $\xi_{int}$ , (c) external error  $\xi_{ext}$ .

## 5 TEST CASE: L-Shaped Cylinder

**GOAL:** show the performances of BCS when dealing with a sparse scatterer

- Number of Views: V
- Number of Measurements: M
- Number of Cells for the Inversion: N
- Number of Cells for the Direct solver: D
- Side of the investigation domain: L

## Test Case Description

#### Direct solver:

- Square domain divided in  $\sqrt{D} \times \sqrt{D}$  cells
- Domain side:  $L = 3\lambda$
- D = 1296 (discretization for the direct solver:  $< \lambda/10$ )

## Investigation domain:

- Square domain divided in  $\sqrt{N} \times \sqrt{N}$  cells
- $L = 3\lambda$
- $2ka = 2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2} = 6\pi\sqrt{2} = 26.65$
- $\#DOF = \frac{(2ka)^2}{2} = \frac{(2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2})^2}{2} = 4\pi^2 \left(\frac{L}{\lambda}\right)^2 = 4\pi^2 \times 9 \approx 355.3$
- N scelto in modo da essere vicino a #DOF: N = 324 (18 × 18)

## Measurement domain:

- Measurement points taken on a circle of radius  $\rho = 3\lambda$
- Full-aspect measurements
- $M \approx 2ka \rightarrow M = 27$

### Sources:

- Plane waves
- $V \approx 2ka \rightarrow V = 27$
- Amplitude A = 1
- Frequency: 300 MHz ( $\lambda = 1$ )

### **Object:**

- L-shaped cylinder
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$
- $\sigma = 0 [S/m]$

## **BCS** parameters:

- Initial estimate of the noise:  $n_0 = 1.0 \times 10^{-3}$
- Convergenze parameter:  $\tau = 1.0 \times 10^{-8}$



Figure 37. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

Ricostruzioni abbastanza buone solo per $\varepsilon_r=1.5.$ 



Figure 38. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].



Figure 39. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].



Figure 40. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **RESULTS:** Error Figures



**Figure 41.** Behaviour of error figures as a function of  $\varepsilon_r$ , for different SNR values: (a) total error  $\xi_{tot}$ , (b) internal error  $\xi_{int}$ , (c) external error  $\xi_{ext}$ .

## 6 TEST CASE: Inhomogeneous L-Shaped Cylinder

**GOAL:** show the performances of BCS when dealing with a sparse scatterer

- Number of Views: V
- Number of Measurements: M
- Number of Cells for the Inversion: N
- Number of Cells for the Direct solver: D
- Side of the investigation domain: L

### **Test Case Description**

#### Direct solver:

- Square domain divided in  $\sqrt{D} \times \sqrt{D}$  cells
- Domain side:  $L = 3\lambda$
- D = 1296 (discretization for the direct solver:  $< \lambda/10$ )

## Investigation domain:

- Square domain divided in  $\sqrt{N} \times \sqrt{N}$  cells
- $L = 3\lambda$
- $2ka = 2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2} = 6\pi\sqrt{2} = 26.65$
- $\#DOF = \frac{(2ka)^2}{2} = \frac{(2 \times \frac{2\pi}{\lambda} \times \frac{L\sqrt{2}}{2})^2}{2} = 4\pi^2 \left(\frac{L}{\lambda}\right)^2 = 4\pi^2 \times 9 \approx 355.3$
- N scelto in modo da essere vicino a #DOF: N = 324 (18  $\times$  18)

## Measurement domain:

- Measurement points taken on a circle of radius  $\rho = 3\lambda$
- Full-aspect measurements
- $M \approx 2ka \rightarrow M = 27$

### Sources:

- Plane waves
- $V \approx 2ka \rightarrow V = 27$
- Amplitude A = 1
- Frequency: 300 MHz ( $\lambda = 1$ )

### **Object:**

- Inhomogeneous L-shaped cylinder
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$
- $\sigma = 0 [S/m]$

#### **BCS** parameters:

- Initial estimate of the noise:  $n_0 = 1.0 \times 10^{-3}$
- Convergenze parameter:  $\tau = 1.0 \times 10^{-8}$



Figure 42. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **Observations:**

Ricostruzioni abbastanza buone solo per $\varepsilon_r=1.5.$ 



Figure 43. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].



Figure 44. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].



Figure 45. Actual object (a) and BCS reconstructed object for (b) Noiseless case, (c) SNR = 20 [dB], (d) SNR = 10 [dB], (e) SNR = 5 [dB].

## **RESULTS:** Error Figures



**Figure 46.** Behaviour of error figures as a function of  $\varepsilon_r$ , for different SNR values: (a) total error  $\xi_{tot}$ , (b) internal error  $\xi_{int}$ , (c) external error  $\xi_{ext}$ .

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