

Reconstruction of sparse scatterers through a Bayesian Compressive Sampling-based Inversion Technique under the Rytov Approximations

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Abstract

This report proposes the numerical validation of a Bayesian Compressive Sensing technique applied to solve an inverse scattering problem within the Rytov approximation. The reconstruction of sparse objects has been investigated, considering single or multiple objects with different shape and different values of dielectric permittivity.

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1 Numerical Validation within Sparsity Conditions

1.1 TEST CASE: Four Square Cylinders $L = 0.16\lambda$

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 $\frac{\lambda}{6} = 0.1667$
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$ (two square), $\varepsilon_r = 1.9$ (two square)

- $\sigma = 0$ [S/m]

BCS parameters:

- Initial estimate of the noise: $n_0 = 8.0 \times 10^{-3}$
- Convergence parameter: $\tau = 1.0 \times 10^{-8}$

RESULTS: Four Square Cylinders $L = 0.16\lambda$

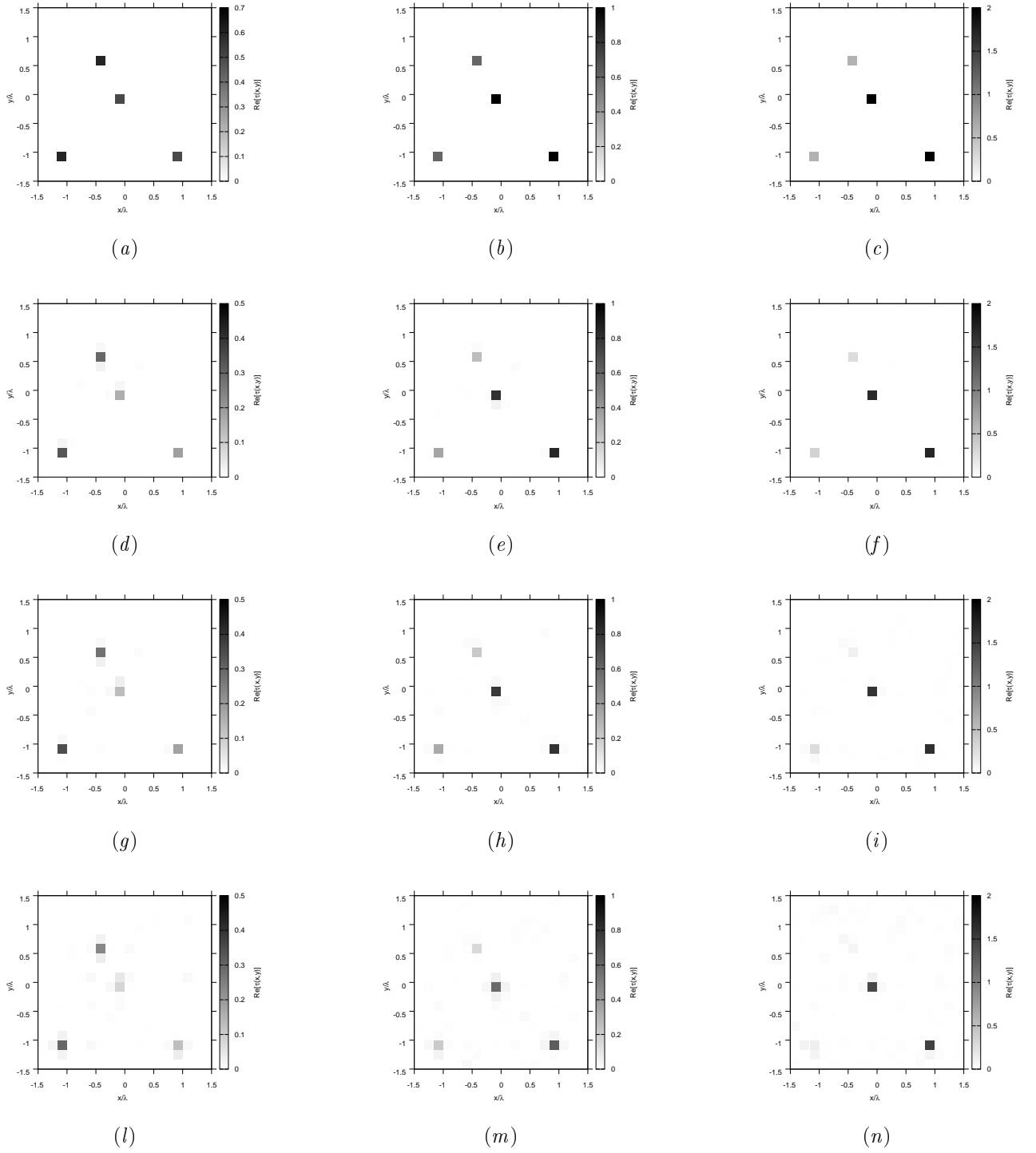
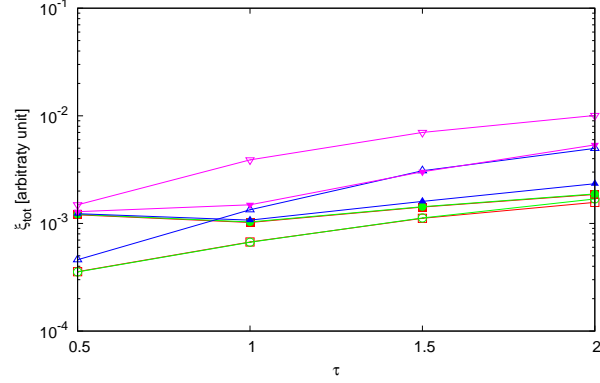
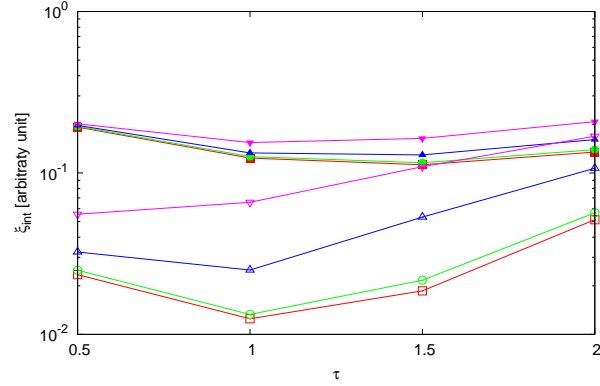


Figure 1. Actual object (a)(b)(c) and BCS reconstructed object with (d)(g)(l) $\epsilon_r = 1.5$, (e)(h)(m) $\epsilon_r = 2.0$, and (f)(i)(n) $\epsilon_r = 3.0$, for (d)(e)(f) Noiseless case, (g)(h)(i) $\text{SNR} = 10$ [dB] and (l)(m)(n) $\text{SNR} = 5$ [dB].

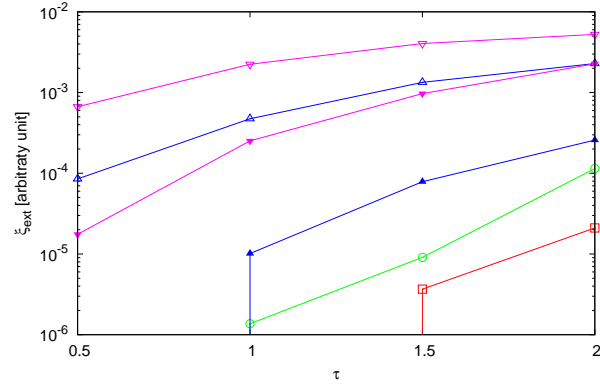
RESULTS: Two Square Cylinders $L = 0.16\lambda$ - Error Figures - Comparison Born/Rytov Approximation



(a)



(b)



(c)

Figure 2. Behaviour of error figures as a function of ε_r , for different SNR values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

1.2 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: $n_0 = 8.0 \times 10^{-3}$
- Convergence parameter: $\tau = 1.0 \times 10^{-8}$

RESULTS: Cross-Shaped Cylinder

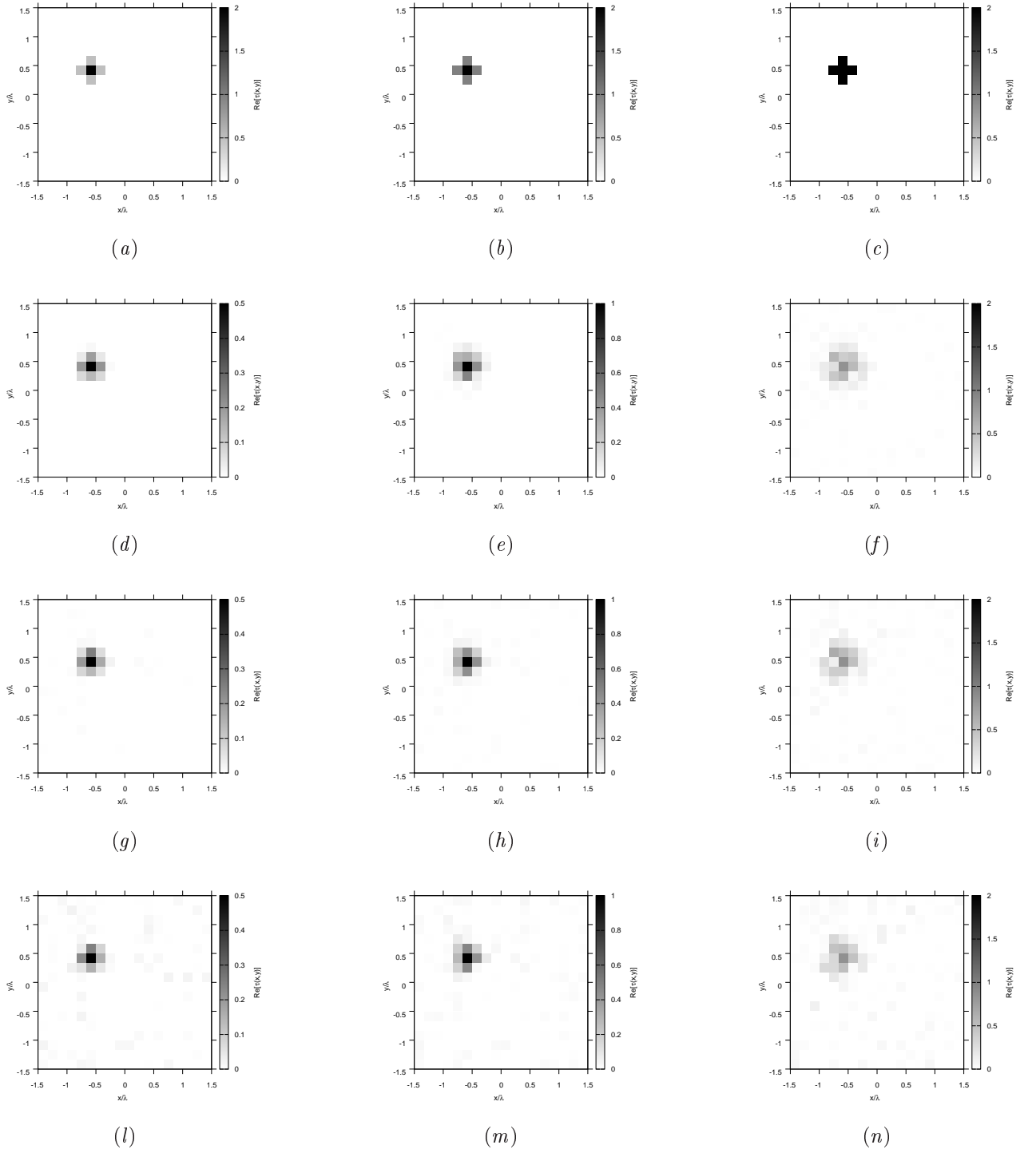
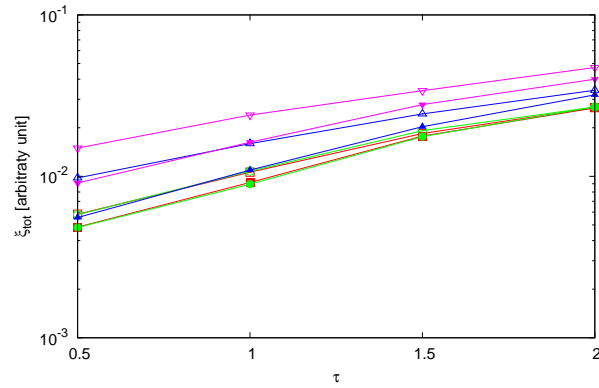


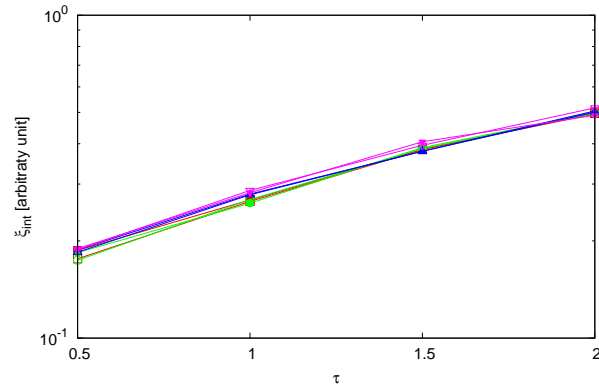
Figure 3. Actual object (a)(b)(c) and BCS reconstructed object with (d)(g)(l) $\epsilon_r = 1.5$, (e)(h)(m) $\epsilon_r = 2.0$, and (f)(i)(n) $\epsilon_r = 3.0$, for (d)(e)(f) Noiseless case, (g)(h)(i) $SNR = 10$ [dB] and (l)(m)(n) $SNR = 5$ [dB].

RESULTS: Cross-Shaped Cylinder - Error Figures - Comparison Born/Rytov Approximation



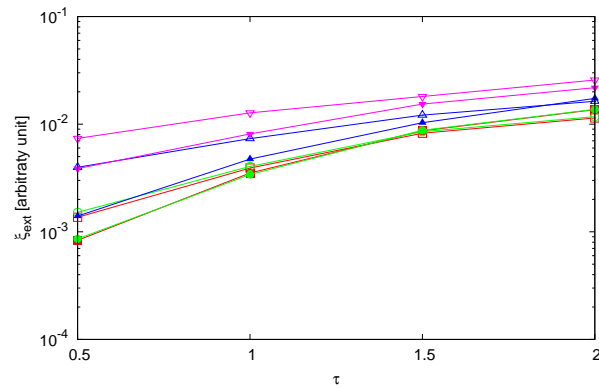
BORN: Noiseless \square SNR = 20dB \circ SNR = 10dB \triangle SNR = 5dB ∇
 RYTOV: Noiseless \blacksquare SNR = 20dB \bullet SNR = 10dB \blacktriangle SNR = 5dB \blacktriangledown

(a)



BORN: Noiseless \square SNR = 20dB \circ SNR = 10dB \triangle SNR = 5dB ∇
 RYTOV: Noiseless \blacksquare SNR = 20dB \bullet SNR = 10dB \blacktriangle SNR = 5dB \blacktriangledown

(b)



BORN: Noiseless \square SNR = 20dB \circ SNR = 10dB \triangle SNR = 5dB ∇
 RYTOV: Noiseless \blacksquare SNR = 20dB \bullet SNR = 10dB \blacktriangle SNR = 5dB \blacktriangledown

(c)

Figure 4. Behaviour of error figures as a function of ε_r , for different SNR values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

1.3 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 $\#DOF$: $N = 324 (18 \times 18)$

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:

- 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: $n_0 = 8.0 \times 10^{-3}$
- Convergence parameter: $\tau = 1.0 \times 10^{-8}$

RESULTS: L-Shaped Cylinder

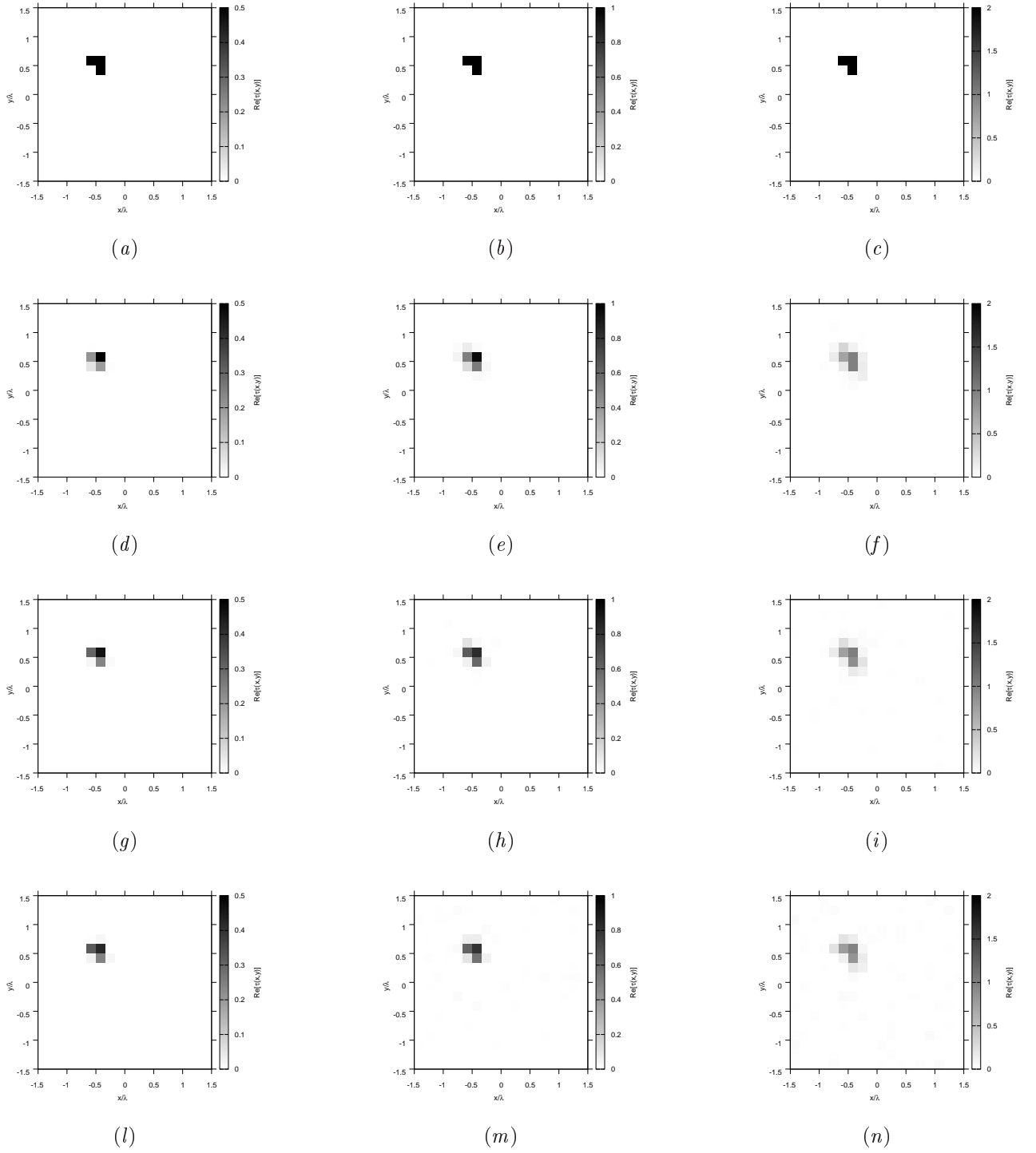
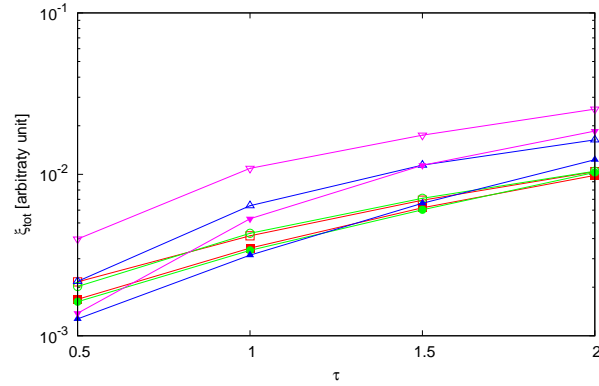


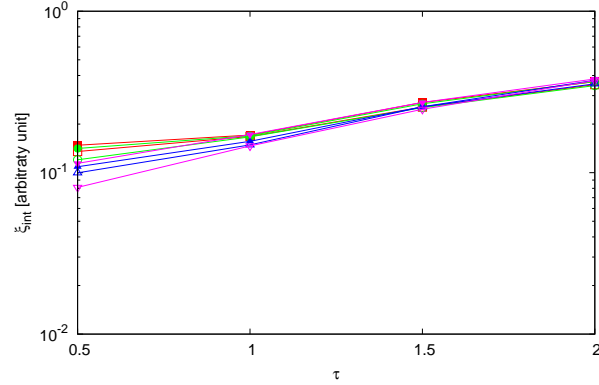
Figure 5. Actual object (a)(b)(c) and BCS reconstructed object with (d)(g)(l) $\epsilon_r = 1.5$, (e)(h)(m) $\epsilon_r = 2.0$, and (f)(i)(n) $\epsilon_r = 3.0$, for (d)(e)(f) Noiseless case, (g)(h)(i) $SNR = 10$ [dB] and (l)(m)(n) $SNR = 5$ [dB].

RESULTS: L-Shaped Cylinder - Error Figures - Comparison Born/Rytov Approximation



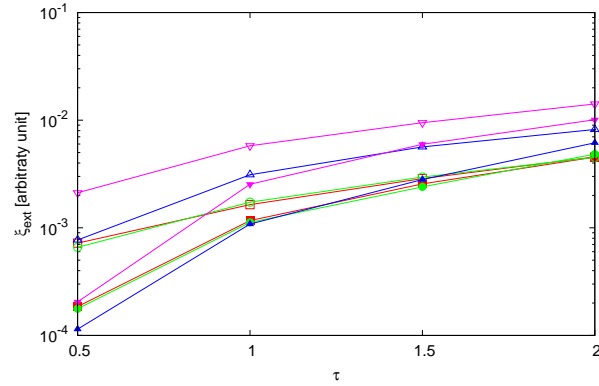
BORN: Noiseless \square SNR = 20dB \circ SNR = 10dB \triangle SNR = 5dB ∇
 RYTOV: Noiseless \blacksquare SNR = 20dB \bullet SNR = 10dB \blacktriangle SNR = 5dB \blacktriangledown

(a)



BORN: Noiseless \square SNR = 20dB \circ SNR = 10dB \triangle SNR = 5dB ∇
 RYTOV: Noiseless \blacksquare SNR = 20dB \bullet SNR = 10dB \blacktriangle SNR = 5dB \blacktriangledown

(b)



BORN: Noiseless \square SNR = 20dB \circ SNR = 10dB \triangle SNR = 5dB ∇
 RYTOV: Noiseless \blacksquare SNR = 20dB \bullet SNR = 10dB \blacktriangle SNR = 5dB \blacktriangledown

(c)

Figure 6. Behaviour of error figures as a function of ε_r , for different SNR values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

1.4 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 $\#DOF$: $N = 324$ (18×18)

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: $n_0 = 8.0 \times 10^{-3}$
- Convergence parameter: $\tau = 1.0 \times 10^{-8}$

RESULTS: Inhomogeneous L-Shaped Cylinder

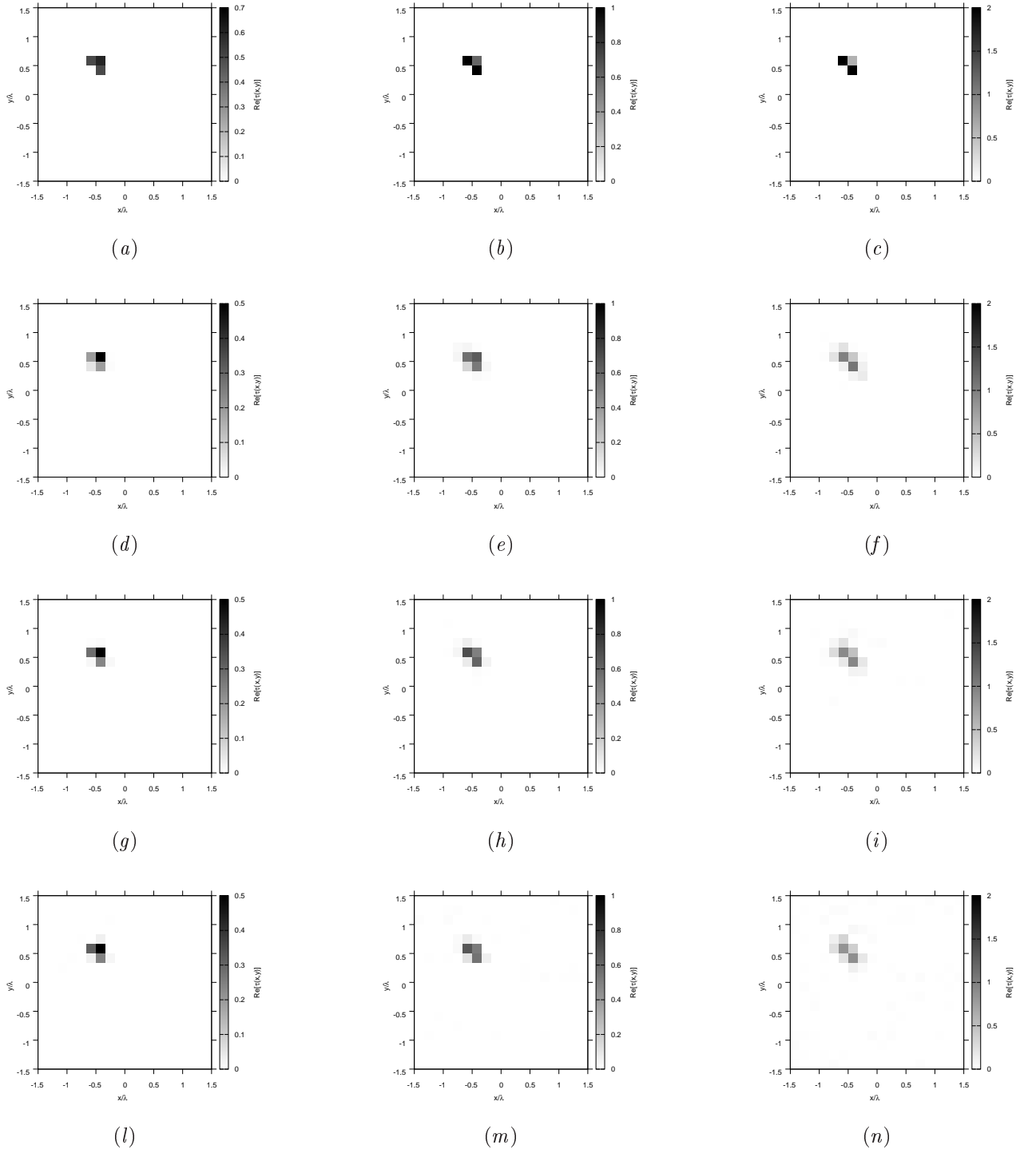
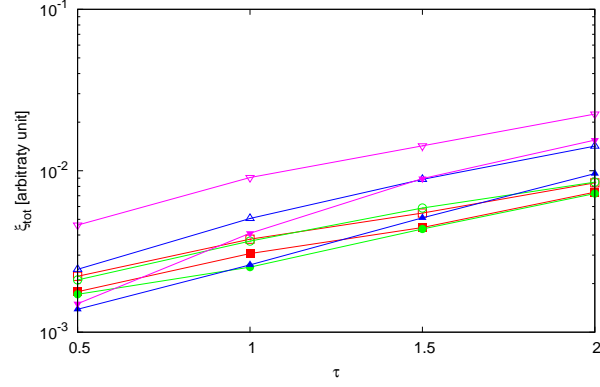
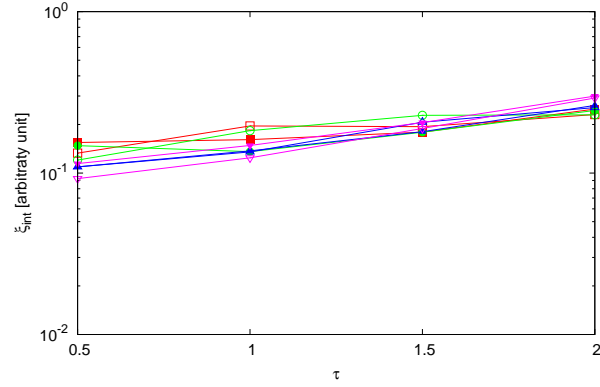


Figure 7. Actual object (a)(b)(c) and BCS reconstructed object with (d)(g)(l) $\epsilon_r = 1.5$, (e)(h)(m) $\epsilon_r = 2.0$, and (f)(i)(n) $\epsilon_r = 3.0$, for (d)(e)(f) Noiseless case, (g)(h)(i) $SNR = 10$ [dB] and (l)(m)(n) $SNR = 5$ [dB].

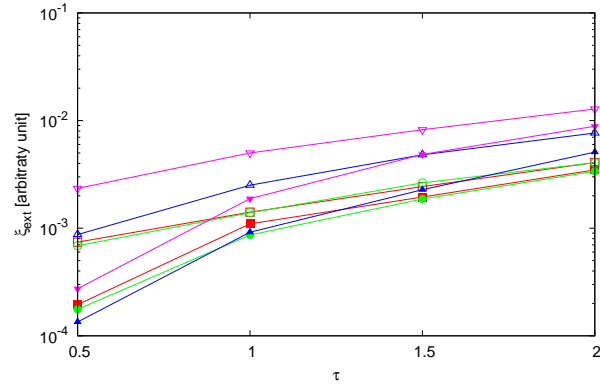
RESULTS: Inhomogeneous L-Shaped Cylinder - Error Figures - Comparison Born/Rytov Approximation



(a)



(b)



(c)

Figure 8. Behaviour of error figures as a function of ε_r , for different SNR values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

1.5 TEST CASE: Two Square Cylinders on the Diagonal

GOAL: evaluate the the performances of *BCS*

- 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
- $L = 3\lambda$
- $N = 324$

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 $\frac{\lambda}{6} = 0.16667$ at a distance $\Delta x, \Delta y$ from each other
- $\varepsilon_r \in \{1.5, 2.0, 2.5, 3.0\}$
- $\sigma = 0$ [S/m]

BCS parameters:

- Initial estimate of the noise: $n_0 = 8.0 \times 10^{-3}$
- Convergence parameter: $\tau = 1.0 \times 10^{-8}$

Resolution Analysis:

- $\Delta x = \Delta y = \{k\lambda/10, k = 0, \dots, 14\}$

RESULTS: Two Square Cylinders on the Diagonal - $\varepsilon_r = 1.5$

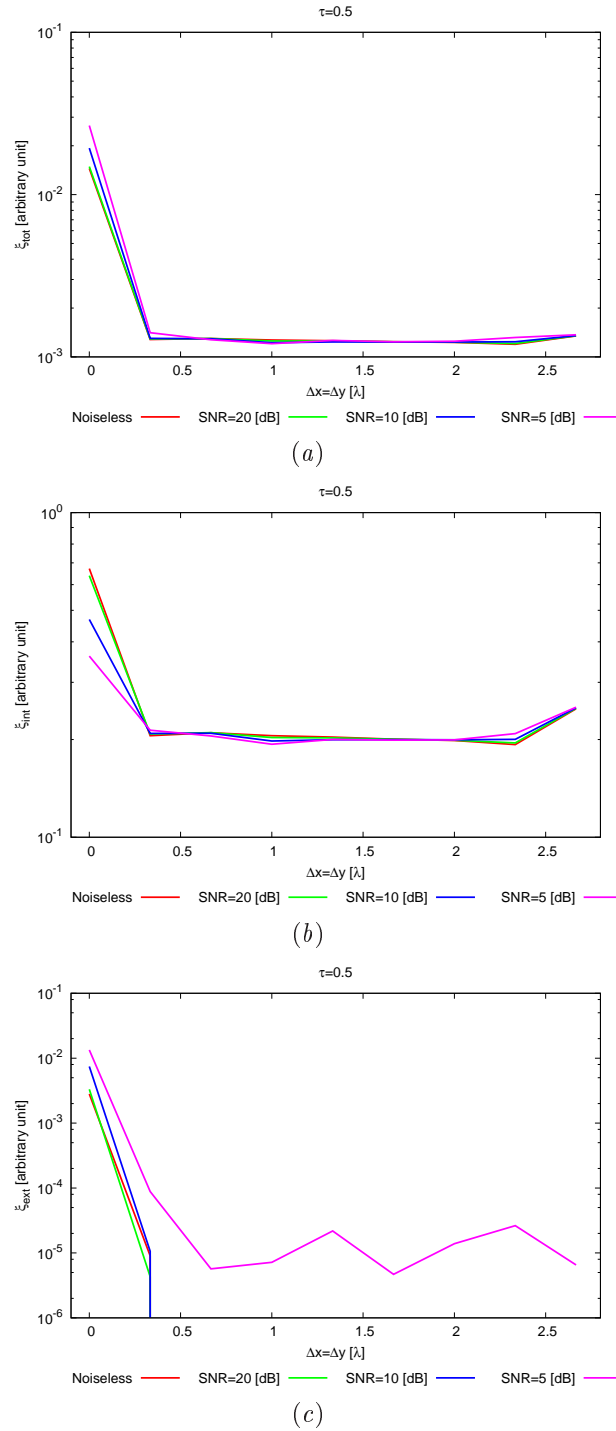


Figure 9. *Resolution analysis* - Behaviour of error figures as a function of $\Delta x = \Delta y$ for different *SNR* values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

RESULTS: Two Square Cylinders on the Diagonal - $\varepsilon_r = 2.0$

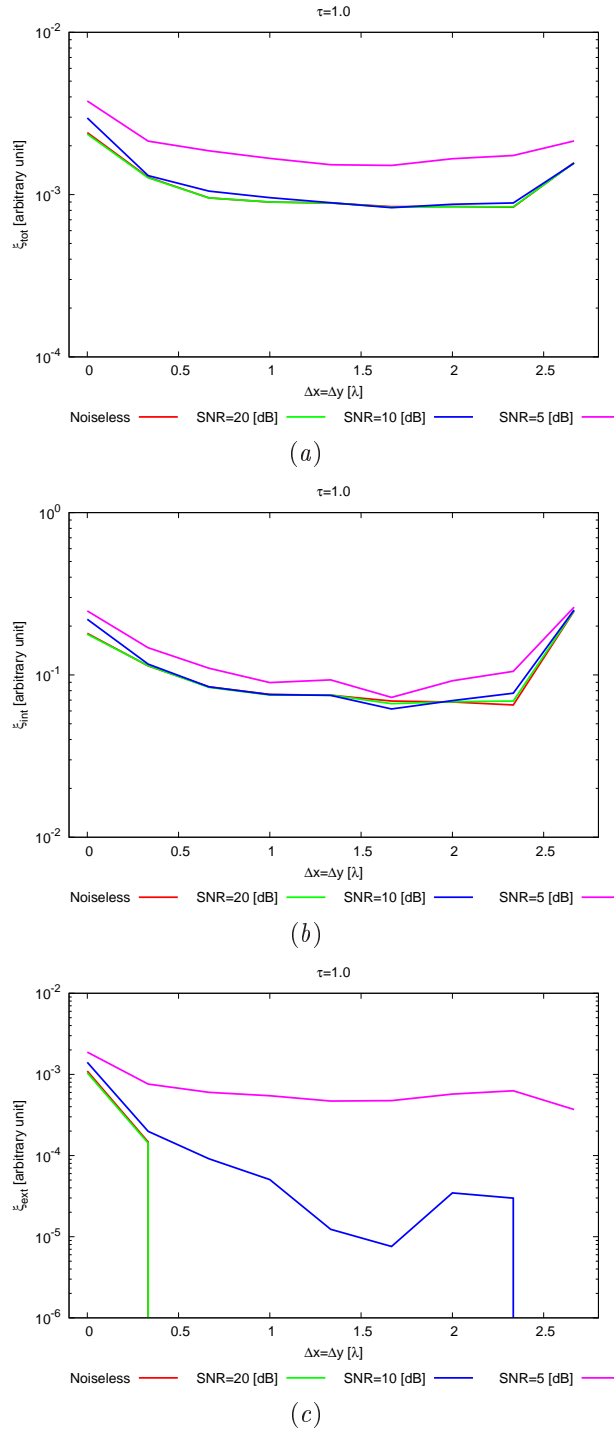


Figure 19. Resolution analysis - Behaviour of error figures as a function of $\Delta x = \Delta y$ for different SNR values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

RESULTS: Two Square Cylinders on the Diagonal - $\varepsilon_r = 2.5$

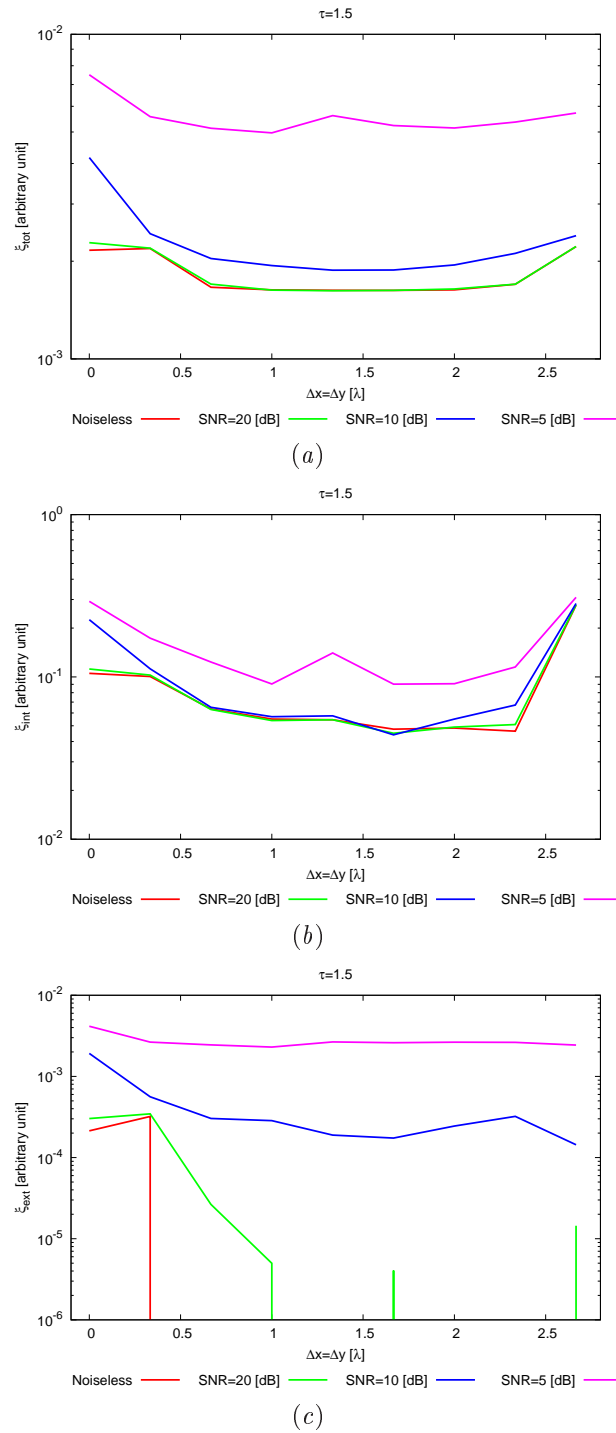


Figure 10. Resolution analysis - Behaviour of error figures as a function of $\Delta x = \Delta y$ for different SNR values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

RESULTS: Two Square Cylinders on the Diagonal - $\varepsilon_r = 3.0$

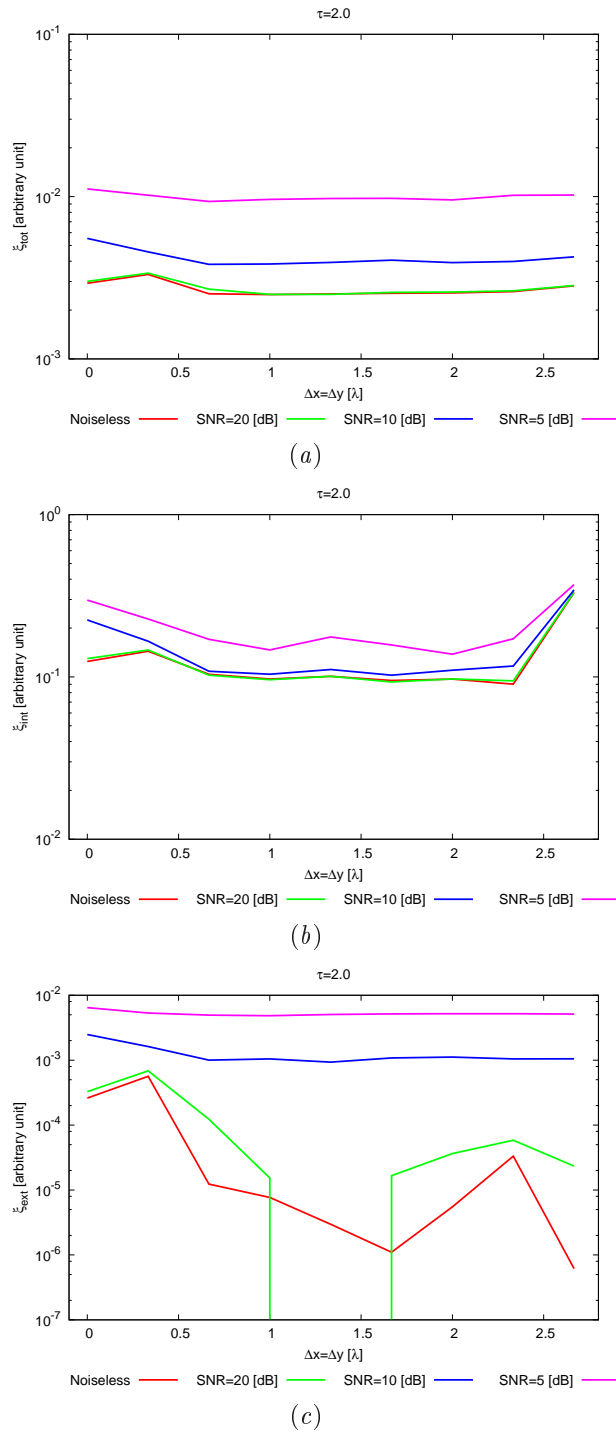


Figure 11. Resolution analysis - Behaviour of error figures as a function of $\Delta x = \Delta y$ for different SNR values: (a) total error ξ_{tot} , (b) internal error ξ_{int} , (c) external error ξ_{ext} .

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