

Damage detection and imaging in solids based on recorded elastodynamic response

Christos G. Panagiotopoulos, Yiannis Petromichelakis and Chrysoula Tsogka

Institute of Applied & Computational Mathematics
Foundation for Research and Technology Hellas, Heraklion, Greece

Department of Mathematics and Applied Mathematics
University of Crete, Heraklion, Greece
e-mail: tsogka@uoc.gr

11th HSTAM Congress 2016, Athens



Table of contents

- 1 Introduction
- 2 Source localization
- 3 Defect localization
- 4 Numerical examples
- 5 Conclusions

Table of contents

1 Introduction

2 Source localization

3 Defect localization

4 Numerical examples

5 Conclusions

Detection and Localization of Damage

- Usually based on response recordings at a number of sensors to monitor structural integrity¹
- Detection : comparison of recordings to a reference (undamaged) state
- Localization : Inverse Problem usually ill-posed
- Solution : *Time-Reversal* (TR) computational tool introduced by Fink et. al.²
- Achieves refocusing of the wave on the source
- Sending back the recorded signals but reversed in time
- Two step approach
 - Forward step
 - Backward step

-
1. GE Stavroulakis, (2000) Inverse and crack identification problems in engineering mechanics
 2. Fink et. al., (2000) Time-reversed acoustics

Time Reversal and applications

- TR is a physical process
- It exploits the time reversibility (based on spatial reciprocity and time reversal invariance) of linear wave equations
- Robust and Simple technique for source localization
- Has been applied in Acoustics³, Elastodynamics⁴, Electromagnetism, Hydrodynamics etc.
- Finds several applications in medicine, telecommunications, underwater acoustics, seismology, engineering structures, etc.
- TR can be used for scatterer localization

example

-
3. L Borcea, G Papanicolaou, C Tsogka and J Berryman, (2002) Imaging and time reversal in random media
 4. D Givoli, (2014) Time Reversal as a Computational Tool in Acoustics and Elastodynamics ↗

In the present work

- Imaging techniques that exploit the fundamental idea of TR
- Description of the numerical implementation for the elastic wave propagation
- Utilization of the Green's function of the Elastodynamic equation to apply imaging techniques
- 2D rectangular bounded domain with elastic behavior
- Investigation of the influence of the boundaries in imaging
- Investigation of the main factors that affect the quality of the image

Table of contents

1 Introduction

2 Source localization

3 Defect localization

4 Numerical examples

5 Conclusions

Forward step

- Simulated Numerically using a mixed finite element formulation⁵
- The domain contains one source at \mathbf{x}_s and N_r receivers at \mathbf{x}_r , $r = 1, \dots, N_r$
- Wave propagation model, Velocity - Stress (first order)

$$\rho \frac{\partial \mathbf{v}}{\partial t} - \operatorname{div} \boldsymbol{\sigma} = \delta(\mathbf{x} - \mathbf{x}_s) f(t) \mathbf{e}_i$$

$$A : \frac{\partial \boldsymbol{\sigma}}{\partial t} - \dot{\boldsymbol{\varepsilon}} = 0 \qquad \qquad \dot{\boldsymbol{\varepsilon}}_{ij} = \frac{1}{2} \left(\frac{\partial v_i}{\partial x_j} + \frac{\partial v_j}{\partial x_i} \right)$$

- Homogeneous Neumann boundary conditions and zero initial conditions
- Excitation function $f(t)$ is a Ricker pulse centered at a known t_0
- The response is being recorded during total time T

5. E Bécache, P Joly and C Tsogka (2002) A new family of mixed finite elements for the linear elastodynamic problem.

Time domain solution - TR Backward step

- Always performed numerically in SHM applications
- The recorded signal is time reversed and retransmitted at \mathbf{x}_r
- Sensors acting as sources introducing right hand side loading terms

$$\rho \frac{\partial \mathbf{v}^{TR}}{\partial t} - \operatorname{div} \boldsymbol{\sigma}^{TR} = \sum_{r=1}^{N_r} \delta(\mathbf{x} - \mathbf{x}_r) v(\mathbf{x}_r, T - t)$$

- Homogeneous Neumann boundary conditions and zero initial conditions
- Refocusing at time $t_{RF} = T - t_0$

Frequency domain solution - Imaging

- Solution of the backward problem

$$F(\mathbf{x}_r, t) = v(\mathbf{x}_r, T-t) \Leftrightarrow \hat{F}(\mathbf{x}_r, \omega) = \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega T}$$

Frequency domain solution - Imaging

- Solution of the backward problem

$$F(\mathbf{x}_r, t) = v(\mathbf{x}_r, T - t) \Leftrightarrow \hat{F}(\mathbf{x}_r, \omega) = \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega T}$$

Frequency domain solution - Imaging

- Solution of the backward problem

$$F(\mathbf{x}_r, t) = v(\mathbf{x}_r, T-t) \Leftrightarrow \hat{F}(\mathbf{x}_r, \omega) = \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega T}$$

$$v^{TR}(\mathbf{x}, t) = G(\mathbf{x}, \mathbf{x}_r, t) \star_t F(\mathbf{x}_r, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{G}(\mathbf{x}, \mathbf{x}_r, \omega) \hat{F}(\mathbf{x}_r, \omega) d\omega$$

Frequency domain solution - Imaging

- Solution of the backward problem

$$F(\mathbf{x}_r, t) = v(\mathbf{x}_r, T-t) \Leftrightarrow \hat{F}(\mathbf{x}_r, \omega) = \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega T}$$

$$v^{TR}(\mathbf{x}, t) = G(\mathbf{x}, \mathbf{x}_r, t) \star_t F(\mathbf{x}_r, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{G}(\mathbf{x}, \mathbf{x}_r, \omega) \hat{F}(\mathbf{x}_r, \omega) d\omega$$

Frequency domain solution - Imaging

- Solution of the backward problem

$$F(\mathbf{x}_r, t) = v(\mathbf{x}_r, T-t) \Leftrightarrow \hat{F}(\mathbf{x}_r, \omega) = \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega T}$$

$$v^{TR}(\mathbf{x}, t) = G(\mathbf{x}, \mathbf{x}_r, t) \star_t F(\mathbf{x}_r, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{G}(\mathbf{x}, \mathbf{x}_r, \omega) \hat{F}(\mathbf{x}_r, \omega) d\omega$$

- Evaluation of $v^{TR}(\mathbf{x}, t)$ at the refocusing time $T - t_0$

$$v^{TR}(\mathbf{x}, t = T - t_0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{G}(\mathbf{x}, \mathbf{x}_r, \omega) \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega t_0} d\omega$$

Frequency domain solution - Imaging

- Solution of the backward problem

$$F(\mathbf{x}_r, t) = v(\mathbf{x}_r, T-t) \Leftrightarrow \hat{F}(\mathbf{x}_r, \omega) = \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega T}$$

$$v^{TR}(\mathbf{x}, t) = G(\mathbf{x}, \mathbf{x}_r, t) \star_t F(\mathbf{x}_r, t) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{G}(\mathbf{x}, \mathbf{x}_r, \omega) \hat{F}(\mathbf{x}_r, \omega) d\omega$$

- Evaluation of $v^{TR}(\mathbf{x}, t)$ at the refocusing time $T - t_0$

$$v^{TR}(\mathbf{x}, t = T - t_0) = \frac{1}{2\pi} \int_{-\infty}^{\infty} \hat{G}(\mathbf{x}, \mathbf{x}_r, \omega) \overline{\hat{v}(\mathbf{x}_r, \omega)} e^{i\omega t_0} d\omega$$

- Imaging functional - numerical approximation

$$I(\mathbf{x}) = \frac{1}{2\pi} \sum_{i,r} \hat{G}^h(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{\hat{v}(\mathbf{x}_r, \omega_i)} \Delta\omega_i$$

Numerical implementation of Imaging

- More precisely

$$\mathbf{I}(\mathbf{x}) = \begin{bmatrix} \mathbf{I}_x(\mathbf{x}) \\ \mathbf{I}_y(\mathbf{x}) \end{bmatrix}, \quad \hat{\mathbf{v}}(\mathbf{x}_r, \omega_i) = \begin{bmatrix} \overline{\mathbf{v}_x(\mathbf{x}_r, \omega_i)} \\ \overline{\mathbf{v}_y(\mathbf{x}_r, \omega_i)} \end{bmatrix},$$

$$\hat{\mathbf{G}}^h(\mathbf{x}, \mathbf{x}_r, \omega_i) = \begin{bmatrix} \mathbf{G}_{xx}(\mathbf{x}, \mathbf{x}_r, \omega_i) & \mathbf{G}_{xy}(\mathbf{x}, \mathbf{x}_r, \omega_i) \\ \mathbf{G}_{yx}(\mathbf{x}, \mathbf{x}_r, \omega_i) & \mathbf{G}_{yy}(\mathbf{x}, \mathbf{x}_r, \omega_i) \end{bmatrix}$$

- The final image \mathbf{I} can be a combination (e.g. SRSS) of \mathbf{I}_x and \mathbf{I}_y , where

$$\begin{bmatrix} \mathbf{I}_x(\mathbf{x}) \\ \mathbf{I}_y(\mathbf{x}) \end{bmatrix} = \frac{\Delta\omega}{2\pi} \sum_{i,r} \begin{bmatrix} \mathbf{G}_{xx}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{\mathbf{v}_x(\mathbf{x}_r, \omega_i)} + \mathbf{G}_{xy}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{\mathbf{v}_y(\mathbf{x}_r, \omega_i)} \\ \mathbf{G}_{yx}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{\mathbf{v}_x(\mathbf{x}_r, \omega_i)} + \mathbf{G}_{yy}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{\mathbf{v}_y(\mathbf{x}_r, \omega_i)} \end{bmatrix}$$

Table of contents

1 Introduction

2 Source localization

3 Defect localization

4 Numerical examples

5 Conclusions

Time domain solution - TR

- Source, receivers and 1 defect - small area around x_d with different wave velocity
- Each time the original pulse passes from the defect it splits into a transmitted and a reflected component
- Assumption : the incident field \mathbf{v}_{inc} is known (response in the healthy domain)
- scattered filed $\mathbf{v}_{scat} = \mathbf{v}_{tot} - \mathbf{v}_{inc}$ to minimize the influence of the source
- The defect acts as a multiple in time source
- \mathbf{v}_{scat} is time reversed and retransmitted
- Not only one refocusing time but the strongest at $t_{RF} = T - t_1 - t_0$

Frequency domain solution - Imaging

- Data at the receiver - Born approximation

$$\hat{v}_{scat}(\mathbf{x}_r, \omega) = k^2 \hat{f}(\omega) \rho \hat{\mathbf{G}}(\mathbf{x}_s, \mathbf{x}_d, \omega) \hat{\mathbf{G}}(\mathbf{x}_d, \mathbf{x}_r, \omega)$$

- It seems natural to define the imaging functional as

$$I(\mathbf{x}) = \sum_{i,r} \hat{\mathbf{G}}^h(\mathbf{x}, \mathbf{x}_s, \omega_i) \hat{\mathbf{G}}^h(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{\hat{v}_{scat}(\mathbf{x}_r, \omega_i)} \Delta\omega_i$$

- Equivalently to source localization, we compute

$$\begin{bmatrix} \mathbf{I}_x(\mathbf{x}) \\ \mathbf{I}_y(\mathbf{x}) \end{bmatrix} = \frac{\Delta\omega}{2\pi} \sum_{i,r} \begin{bmatrix} \mathbf{G}_{xx}(\mathbf{x}, \mathbf{x}_s, \omega_i) & \mathbf{G}_{xy}(\mathbf{x}, \mathbf{x}_r, \omega_i) \\ \mathbf{G}_{yx}(\mathbf{x}, \mathbf{x}_s, \omega_i) & \mathbf{G}_{yy}(\mathbf{x}, \mathbf{x}_s, \omega_i) \end{bmatrix} \times \begin{bmatrix} \mathbf{G}_{xx}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{v_x(\mathbf{x}_r, \omega_i)} + \mathbf{G}_{xy}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{v_y(\mathbf{x}_r, \omega_i)} \\ \mathbf{G}_{yx}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{v_x(\mathbf{x}_r, \omega_i)} + \mathbf{G}_{yy}(\mathbf{x}, \mathbf{x}_r, \omega_i) \overline{v_y(\mathbf{x}_r, \omega_i)} \end{bmatrix}$$

Table of contents

1 Introduction

2 Source localization

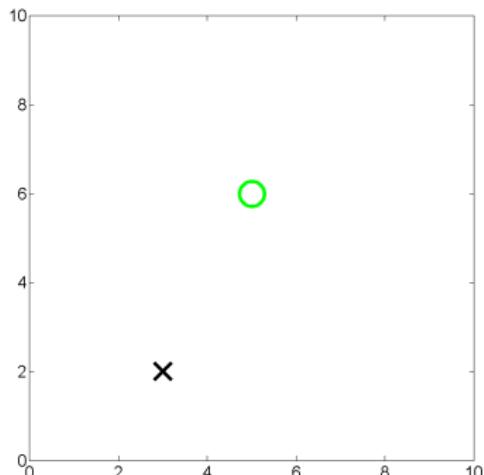
3 Defect localization

4 Numerical examples

5 Conclusions

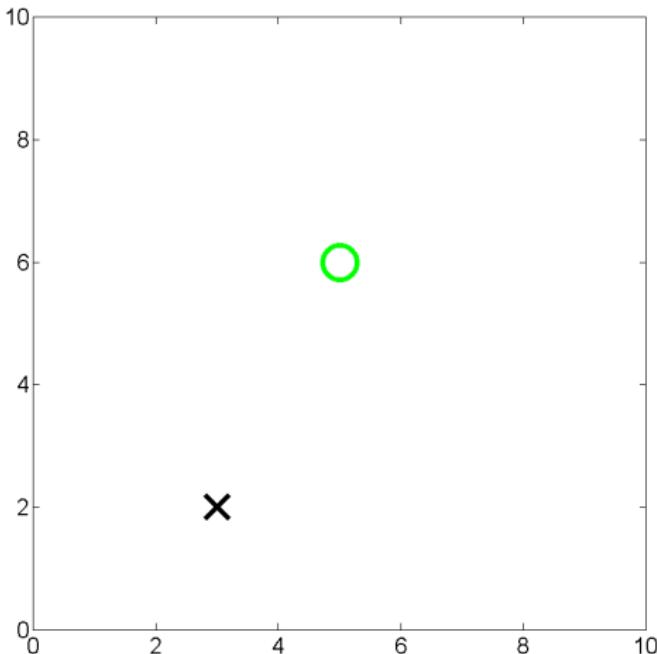
Numerical example - source localization

- **Geometry** : rectangular domain
 $L_x = 10.$ and $L_y = 10.$
- **Mesh (numerical solution)** : 200×200 grid with rectangular elements
- **Mesh (Imaging)** : 50×50 grid with rectangular elements
- **Material** : elastic with Lamé coefficients
 $\lambda = 1.$ and $\mu = 1.$
- **Velocities** : pressure waves $c_p = 1.73$ and shear waves $c_s = 1.$
- **Excitation function** : Ricker pulse at a central frequency 2.



Numerical example - source localization

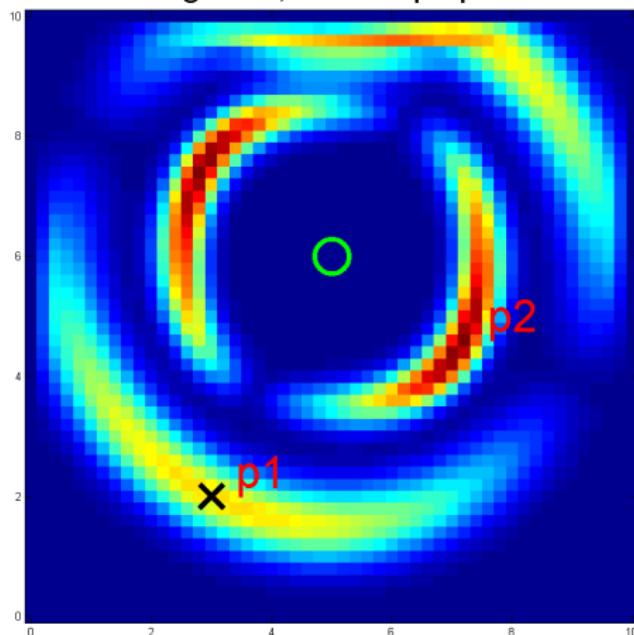
1 source and 1 receiver - increasing total time T



Numerical example - source localization

1 source and 1 receiver - increasing total time T

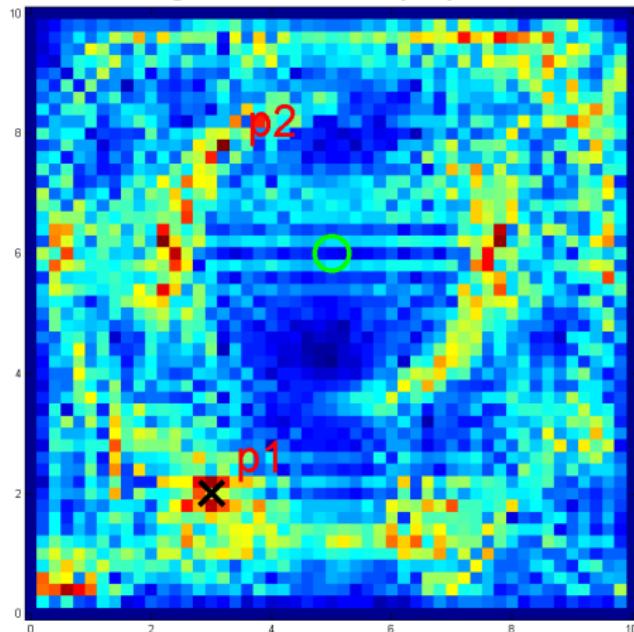
$T = 0.5$ diagonals, SNR = $p_1/p_2 = 0.6534$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

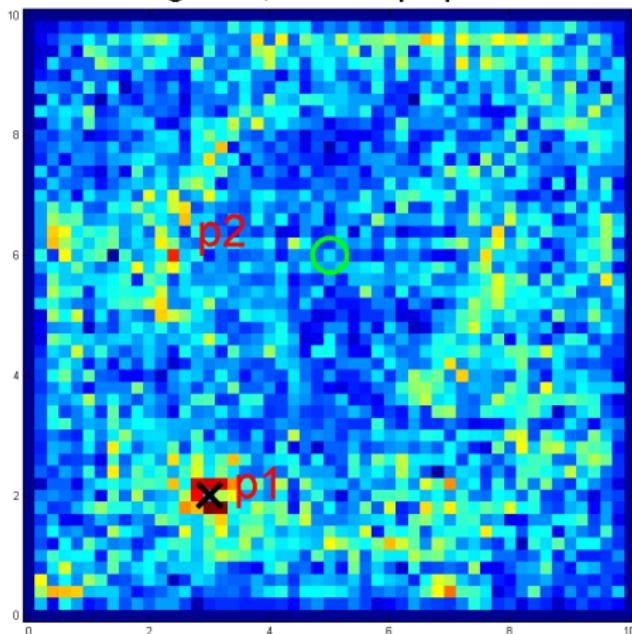
$T = 1$ diagonals, $\text{SNR} = p_1/p_2 = 0.9678$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

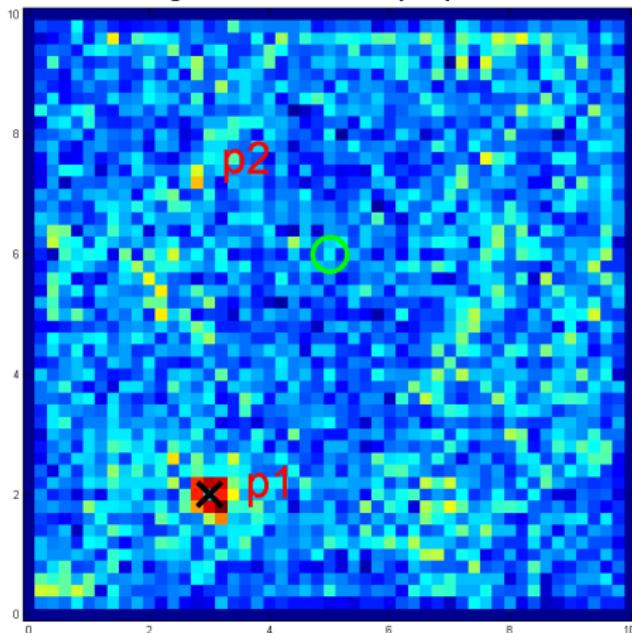
T = 2 diagonals, SNR = $p_1/p_2 = 1.1967$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

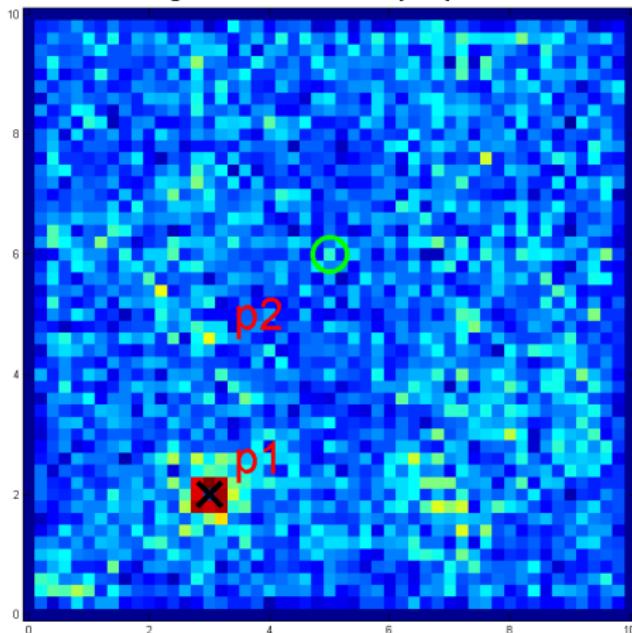
T = 3 diagonals, SNR = $p_1/p_2 = 1.4648$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

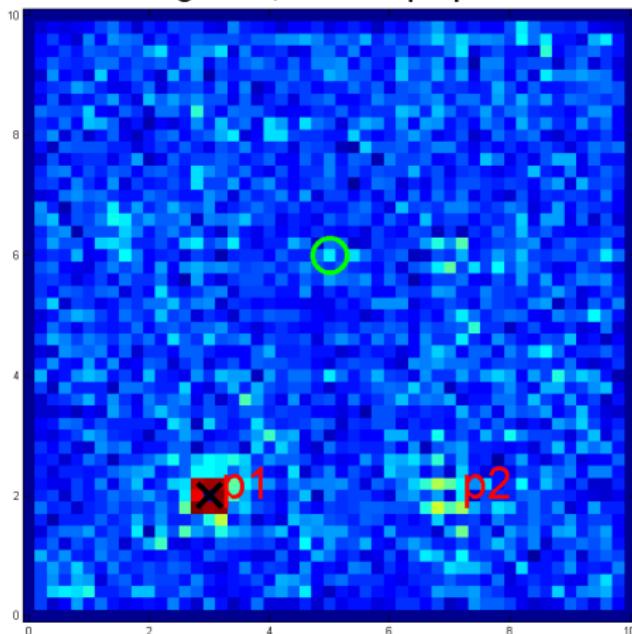
T = 5 diagonals, SNR = $p_1/p_2 = 1.6664$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

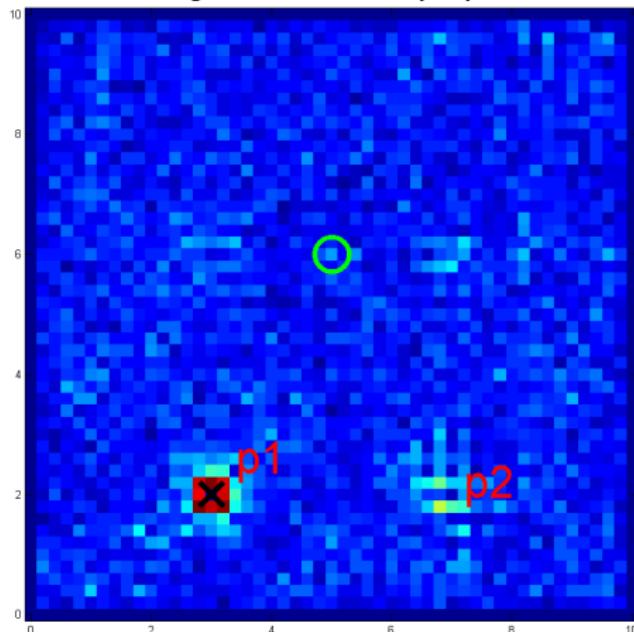
$T = 10$ diagonals, $\text{SNR} = p_1/p_2 = 1.7472$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

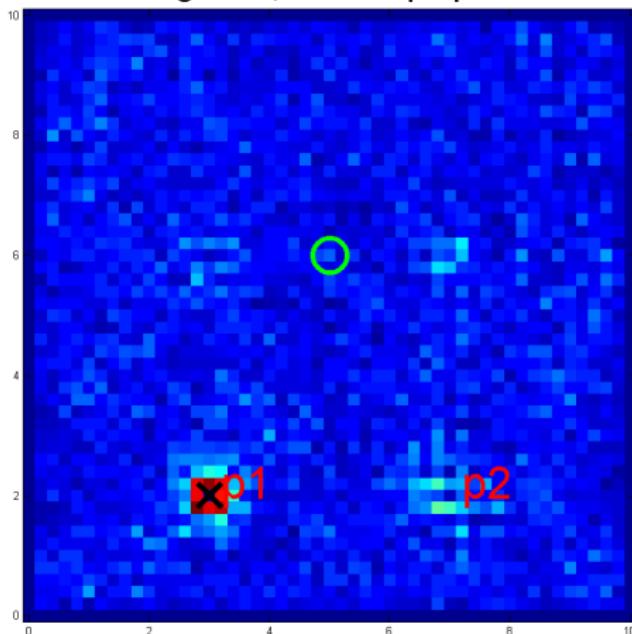
T = 20 diagonals, SNR = $p_1/p_2 = 1.789$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

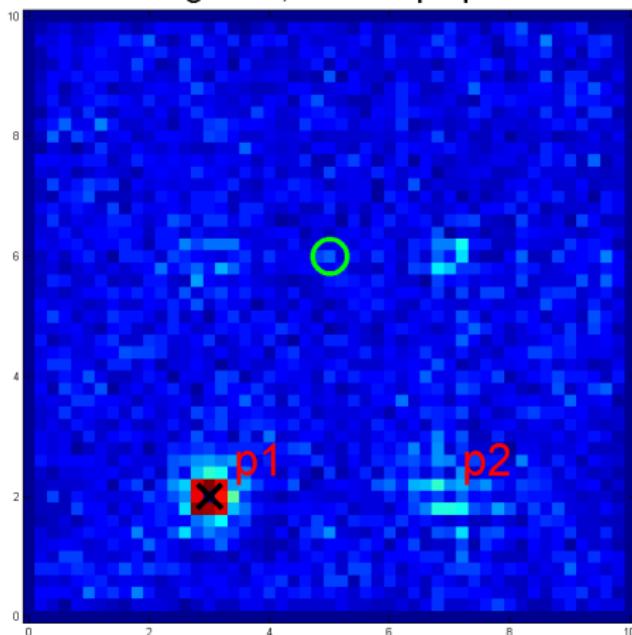
T = 30 diagonals, SNR = $p_1/p_2 = 2.1821$



Numerical example - source localization

1 source and 1 receiver - increasing total time T

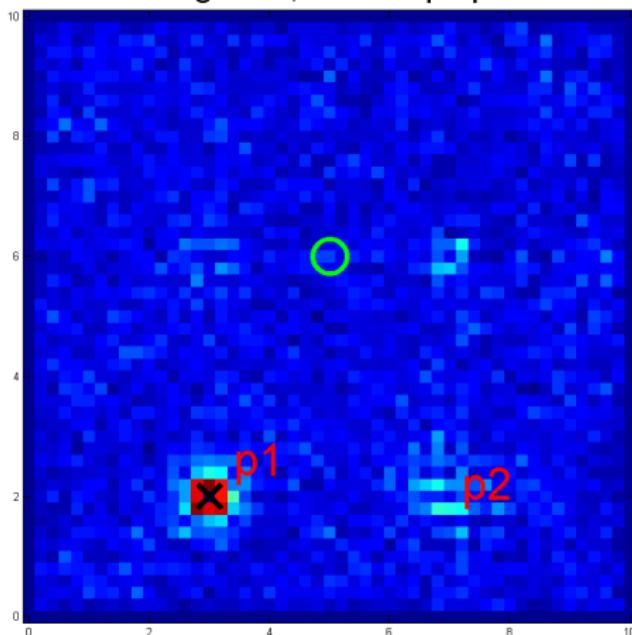
T = 40 diagonals, SNR = $p_1/p_2 = 2.392$



Numerical example - source localization

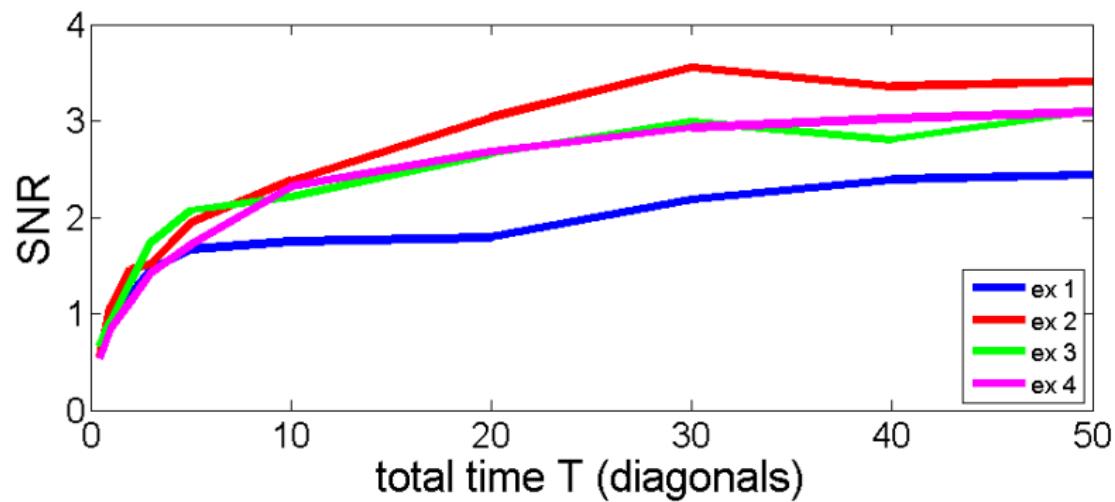
1 source and 1 receiver - increasing total time T

T = 50 diagonals, SNR = $p_1/p_2 = 2.44$



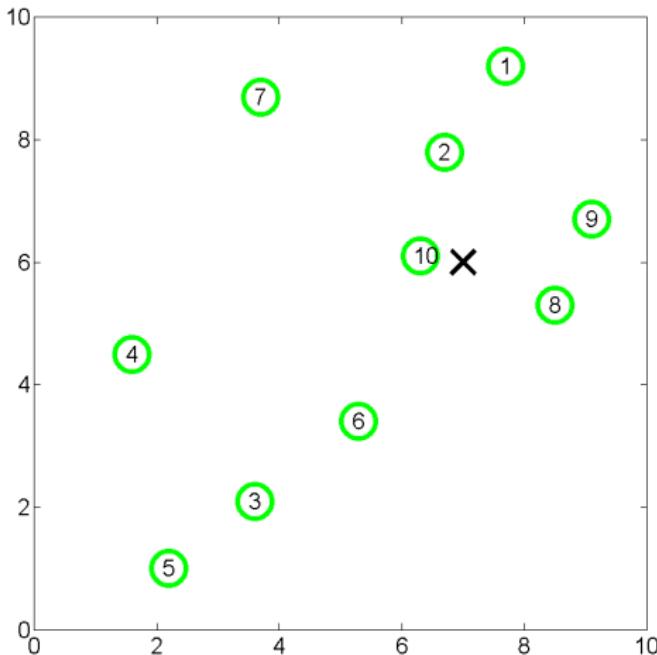
Numerical example - source localization

1 source and 1 receiver - increasing total time T



Numerical example - source localization

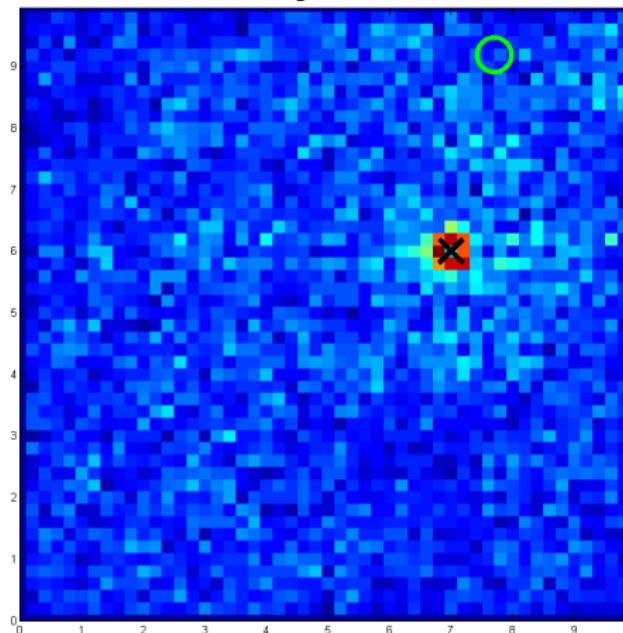
1 source and increasing number of receivers



Numerical example - source localization

1 source and increasing number of receivers

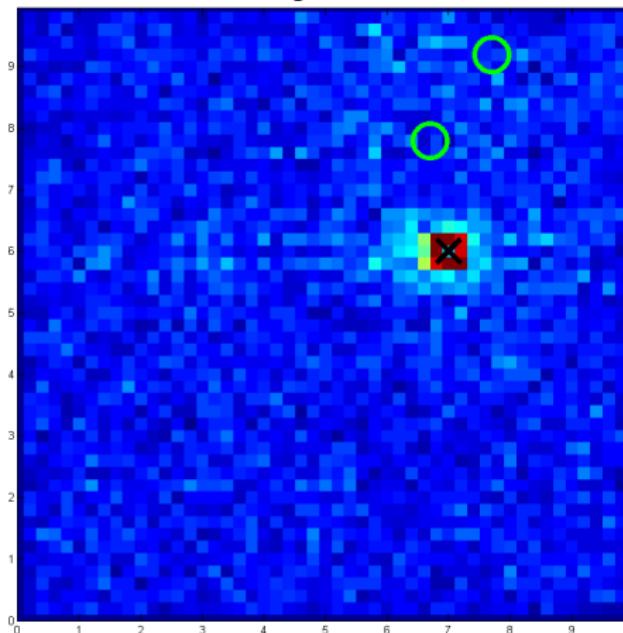
$N_r = 1$, $T = 10$ diagonals, $SNR = 2.3351$



Numerical example - source localization

1 source and increasing number of receivers

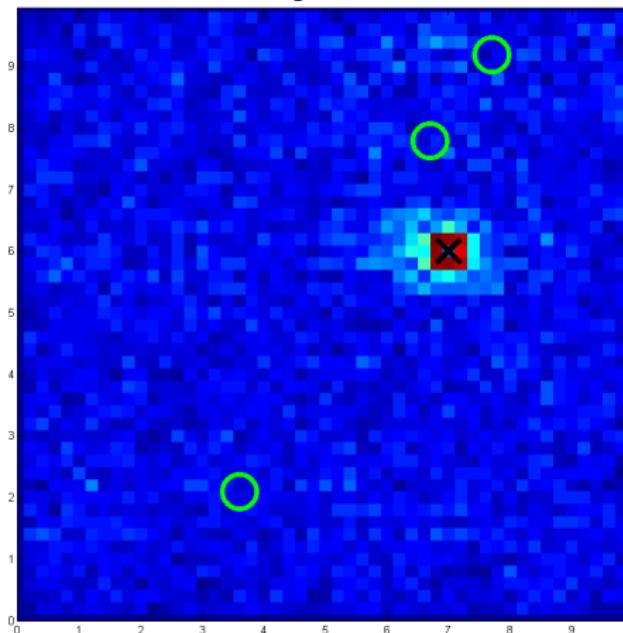
$N_r = 2$, $T = 10$ diagonals, $SNR = 2.5843$



Numerical example - source localization

1 source and increasing number of receivers

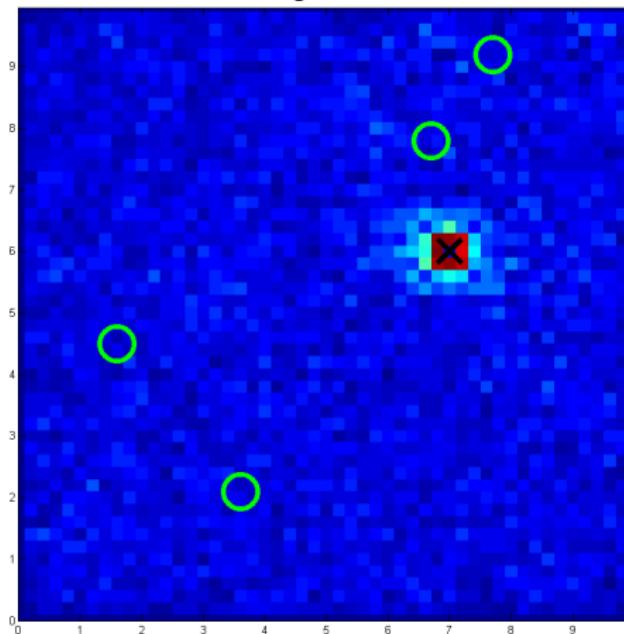
$\text{Nr} = 3, \text{T} = 10$ diagonals, $\text{SNR} = 3.5819$



Numerical example - source localization

1 source and increasing number of receivers

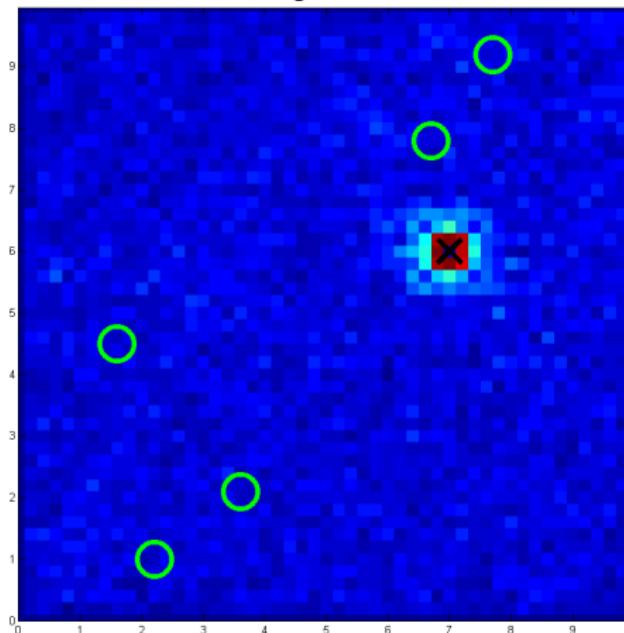
$\text{Nr} = 4$, $T = 10$ diagonals, $\text{SNR} = 4.2135$



Numerical example - source localization

1 source and increasing number of receivers

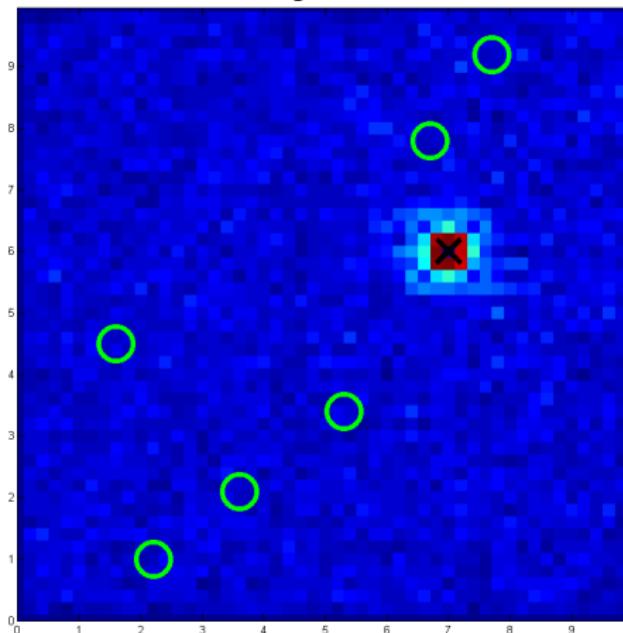
$\text{Nr} = 5, \text{T} = 10 \text{ diagonals}, \text{SNR} = 4.7374$



Numerical example - source localization

1 source and increasing number of receivers

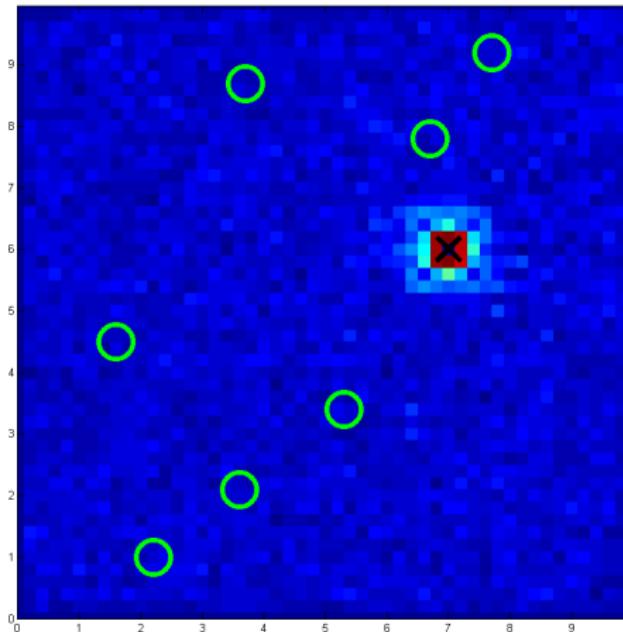
$N_r = 6$, $T = 10$ diagonals, $SNR = 4.8888$



Numerical example - source localization

1 source and increasing number of receivers

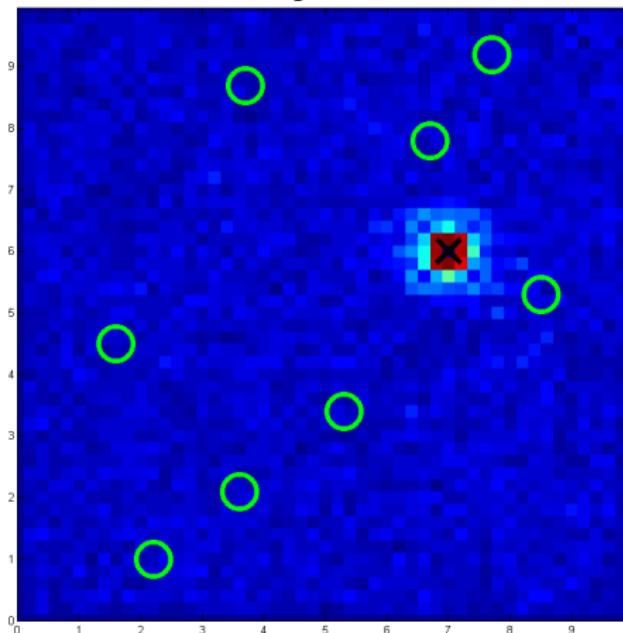
$N_r = 7$, $T = 10$ diagonals, $SNR = 5.3904$



Numerical example - source localization

1 source and increasing number of receivers

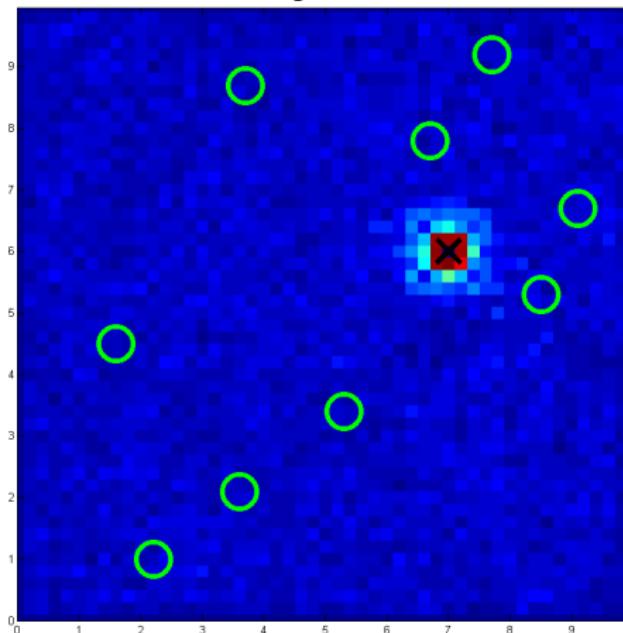
$N_r = 8$, $T = 10$ diagonals, $SNR = 5.5286$



Numerical example - source localization

1 source and increasing number of receivers

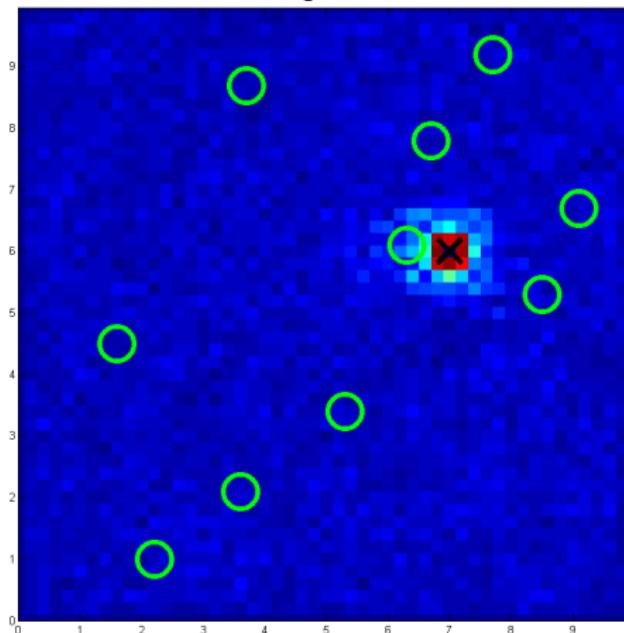
$N_r = 9$, $T = 10$ diagonals, $SNR = 5.9734$



Numerical example - source localization

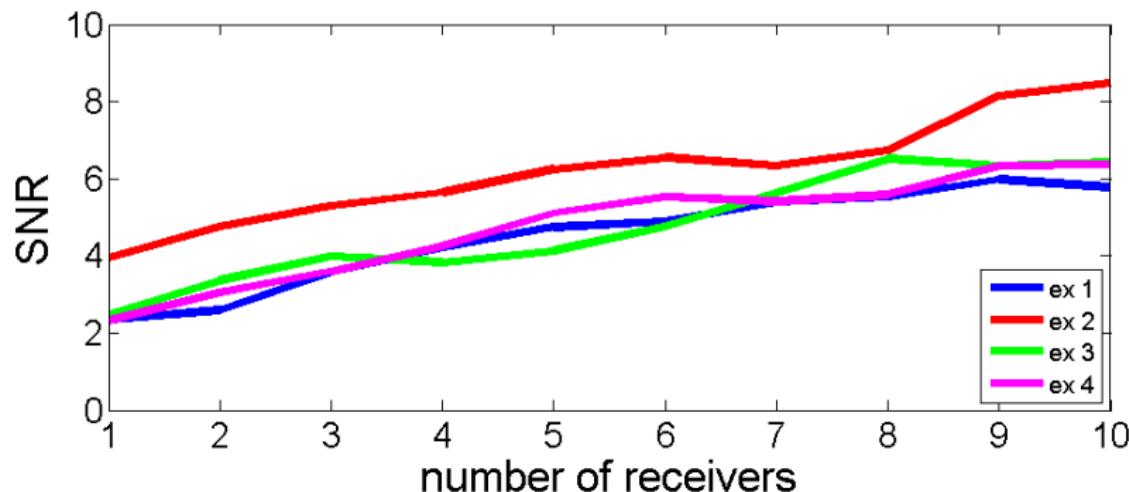
1 source and increasing number of receivers

Nr = 10, T = 10 diagonals, SNR = 5.7849



Numerical example - source localization

1 source and increasing number of receivers

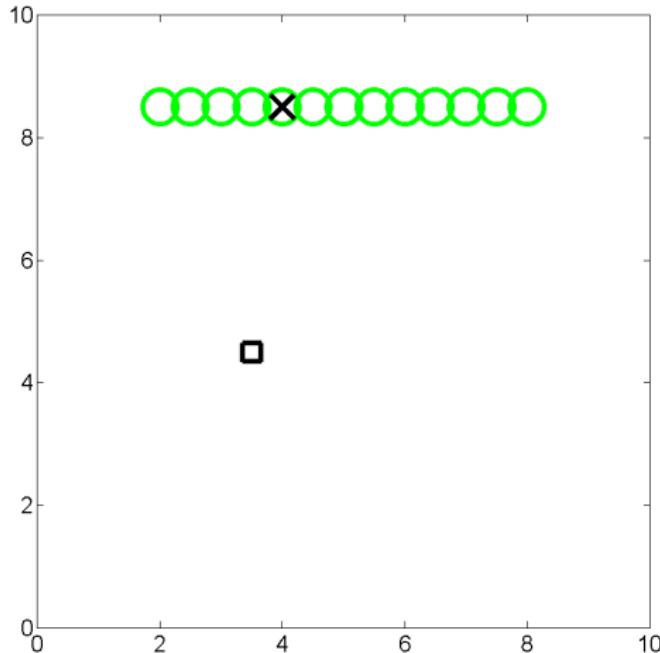


Numerical example - defect localization

- **Geometry** : rectangular domain $L_x = 10.$ and $L_y = 10.$
- **Mesh (numerical solution)** : 200×200 grid with rectangular elements
- **Mesh (Imaging)** : 200×200 grid with rectangular elements
- **Material** : elastic with Lamé coefficients $\lambda = 1.$ and $\mu = 1.$
- **Velocities** : pressure waves $c_p = 1.73$ and shear waves $c_s = 1.$
- **Excitation function** : Ricker pulse with central frequency 4.
- **Defect :**
 - location : (3.5,4.5)
 - size : 0.05×0.05
 - material : $\lambda = 0.5$ and $\mu = 1.$

Numerical example - defect localization

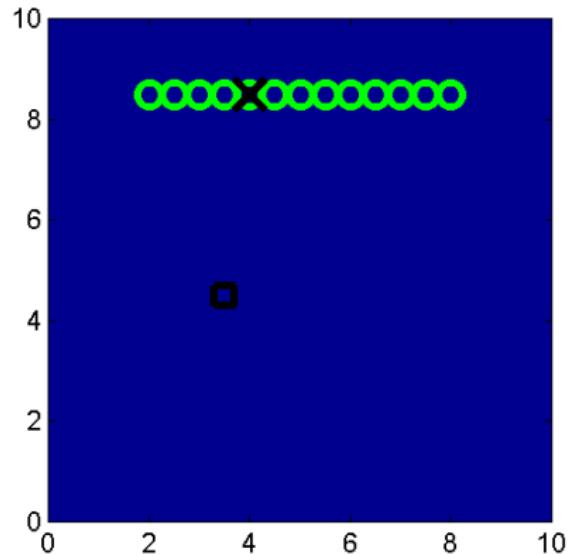
1 defect, 1 source and array of 13 receivers



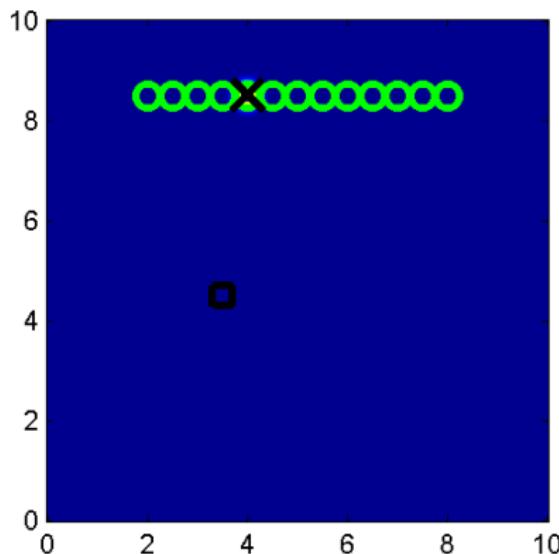
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



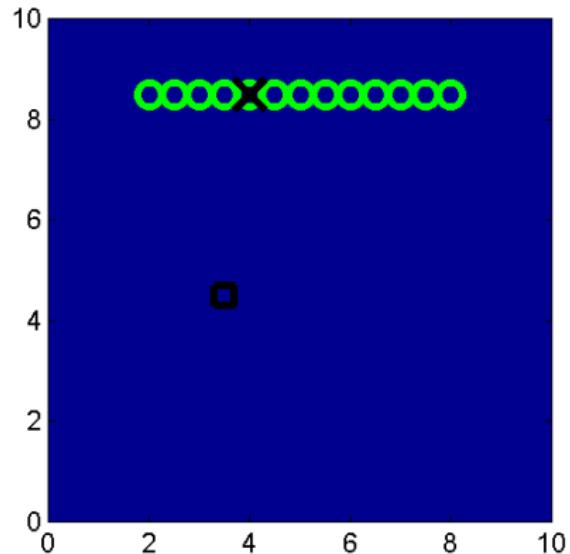
total field



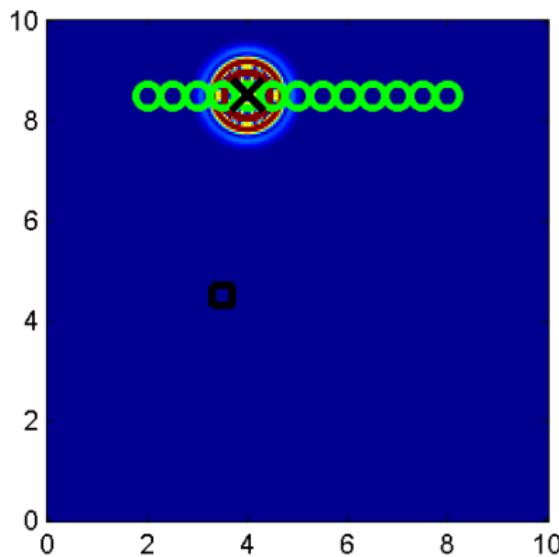
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



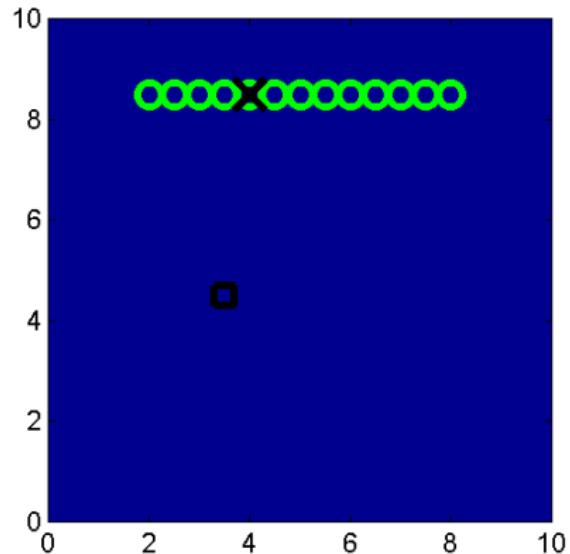
total field



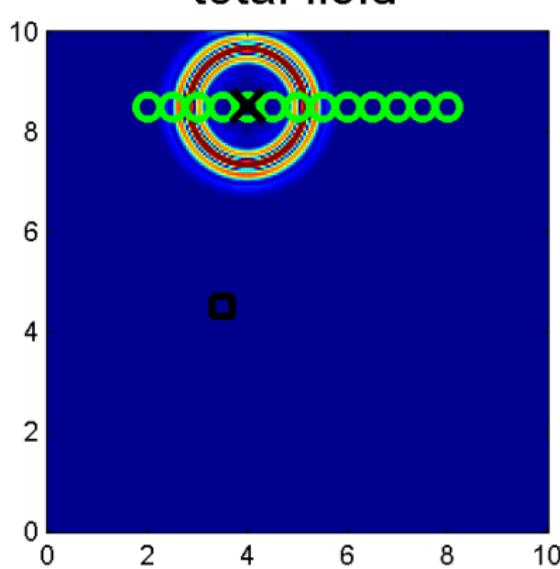
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



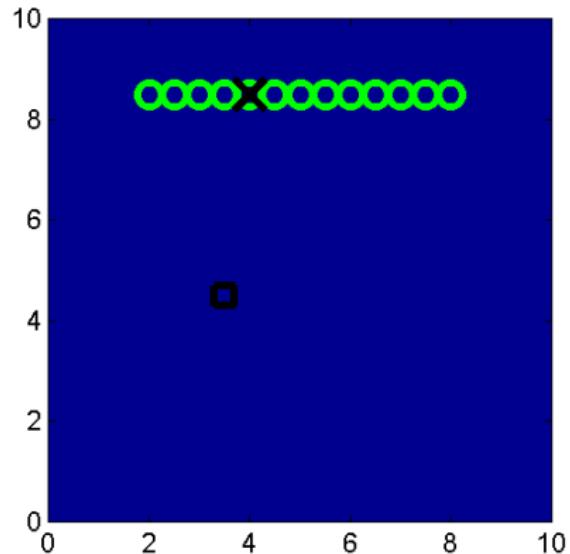
total field



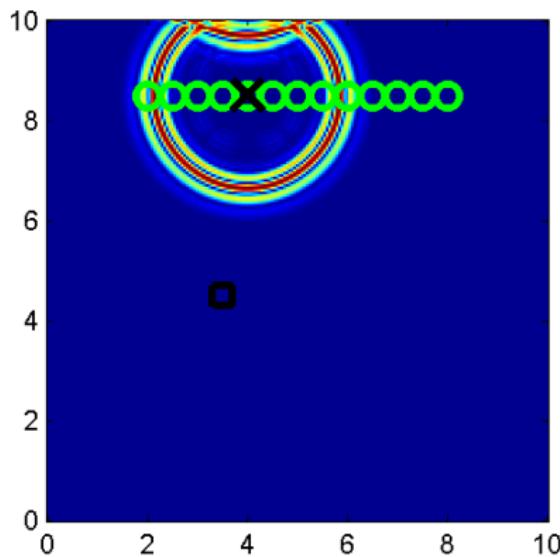
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



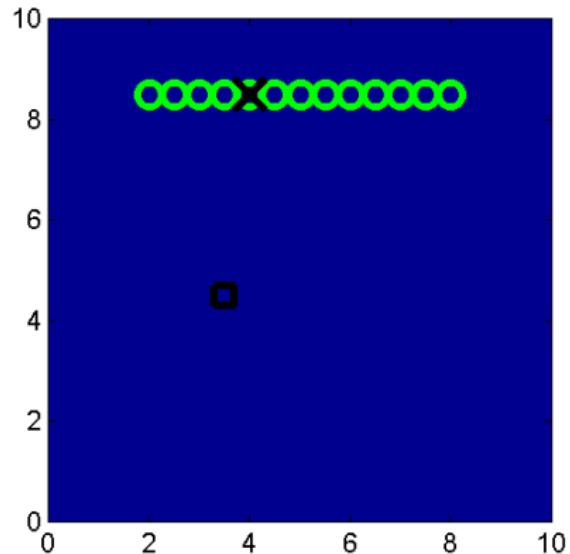
total field



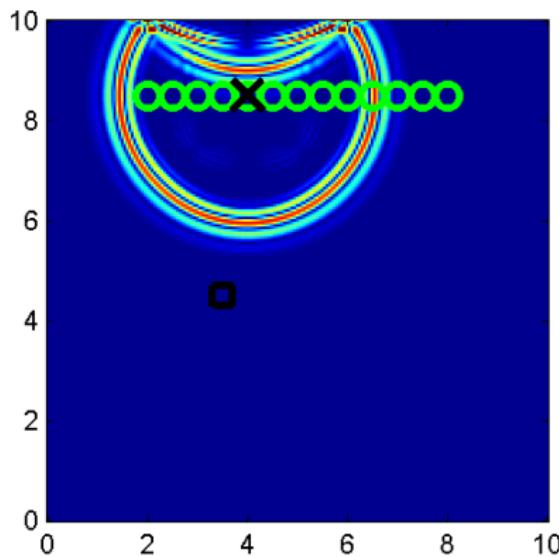
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



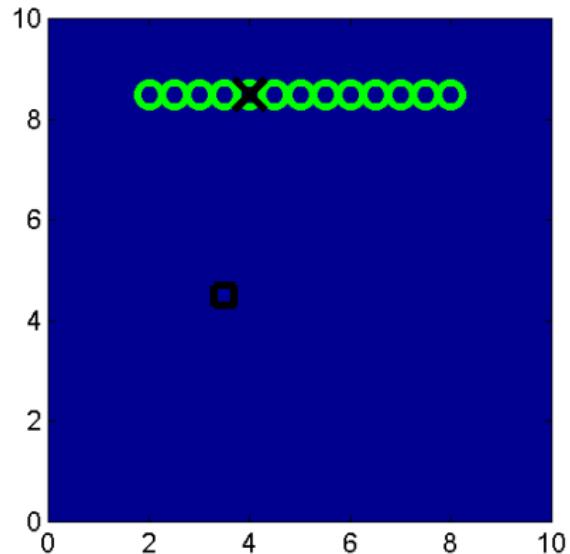
total field



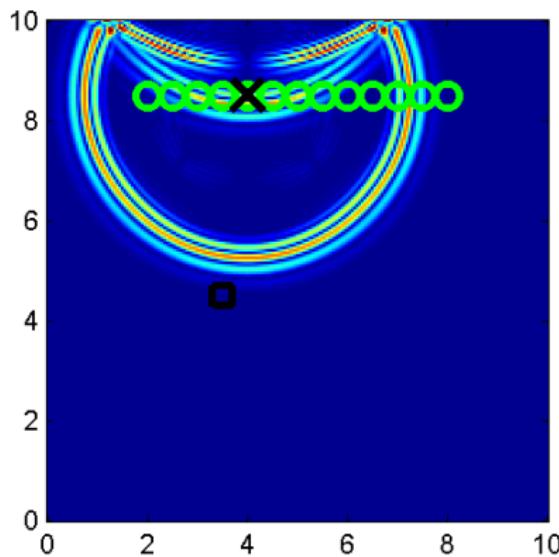
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



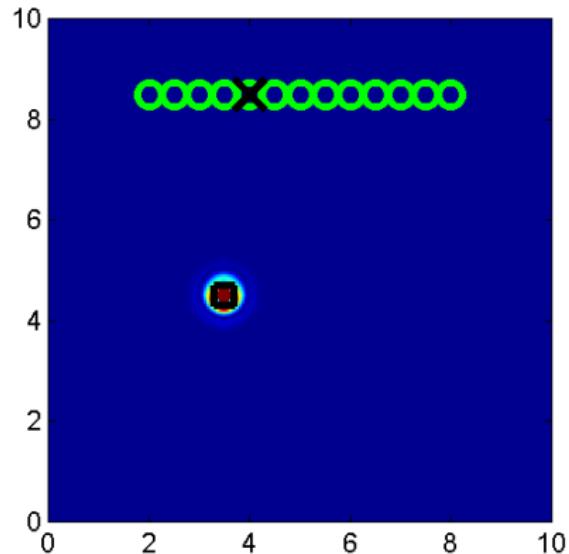
total field



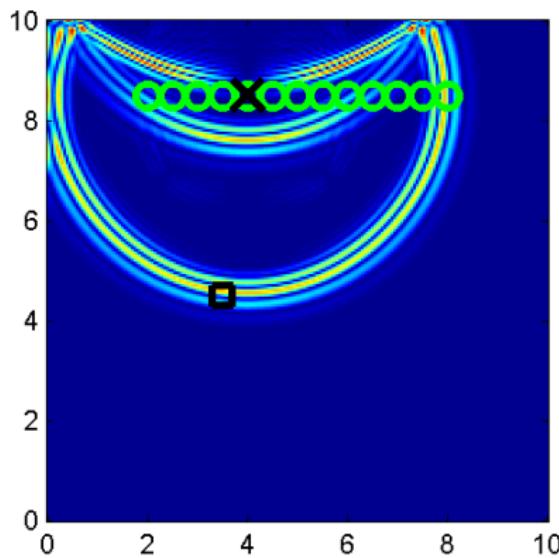
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



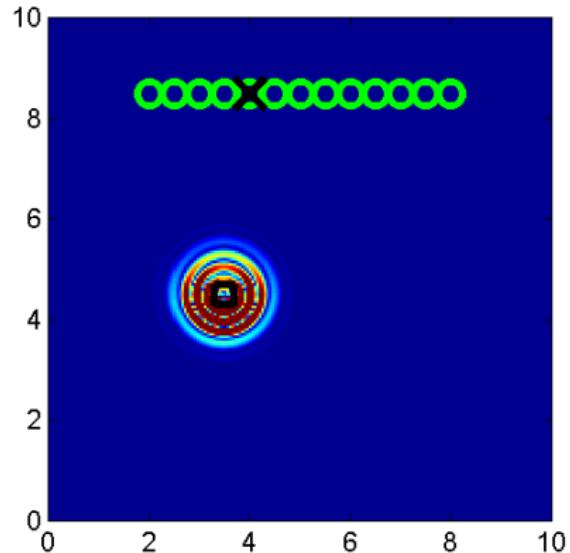
total field



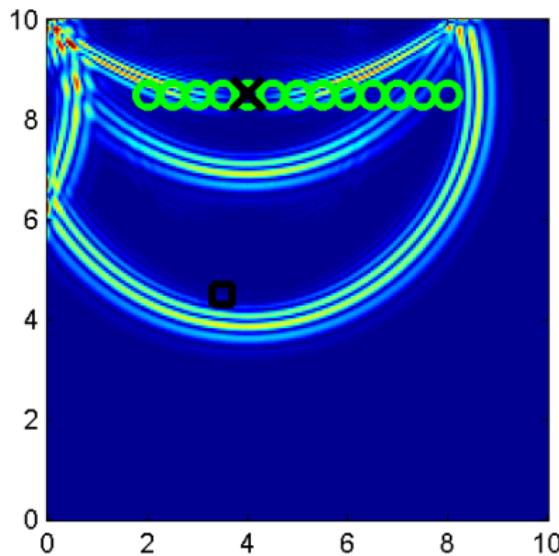
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



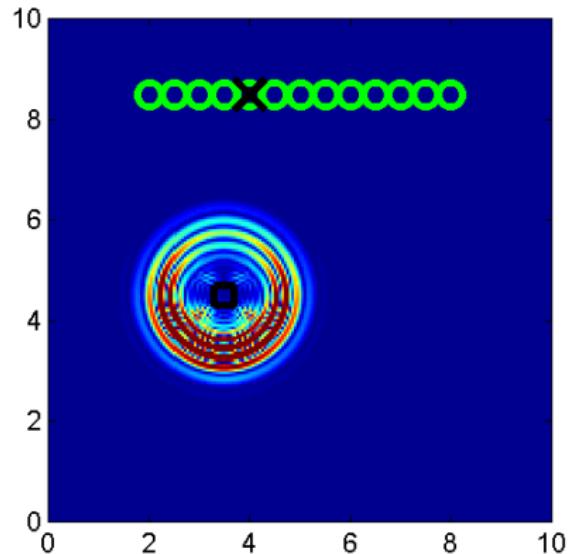
total field



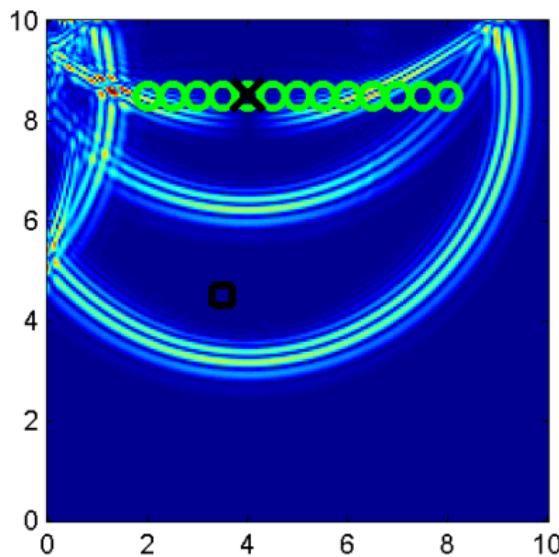
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



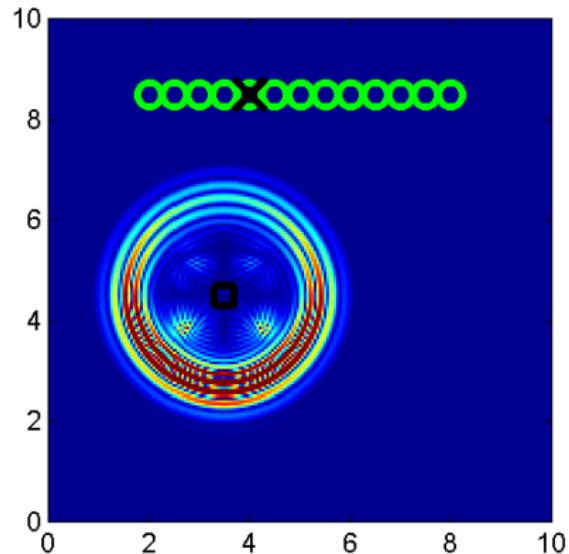
total field



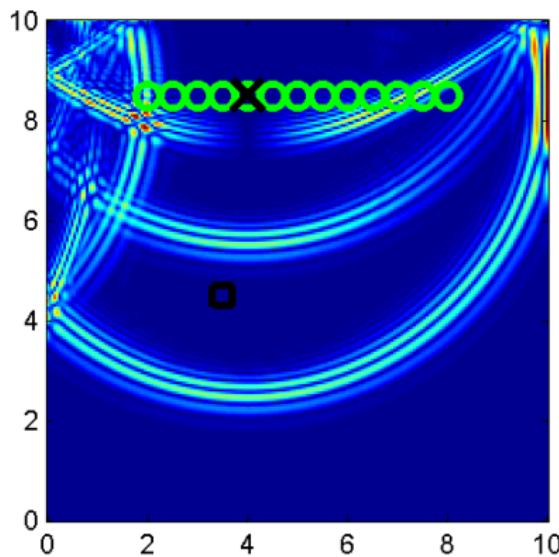
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



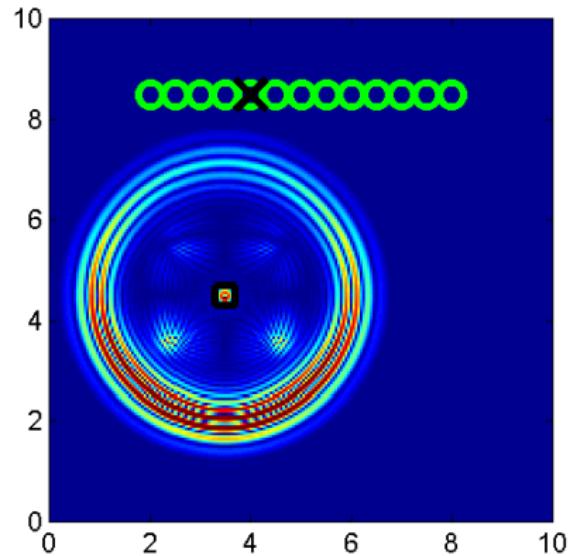
total field



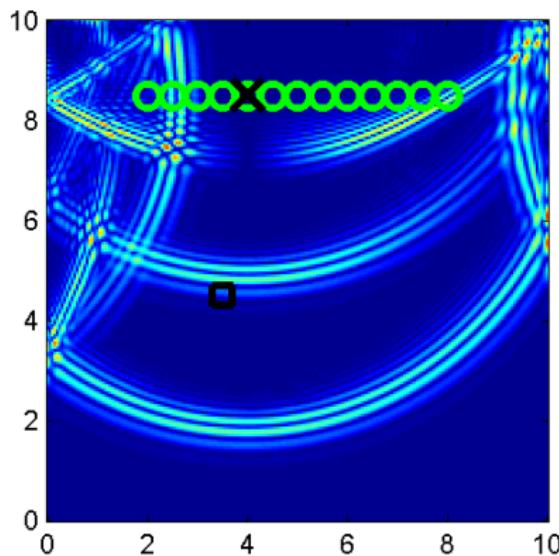
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



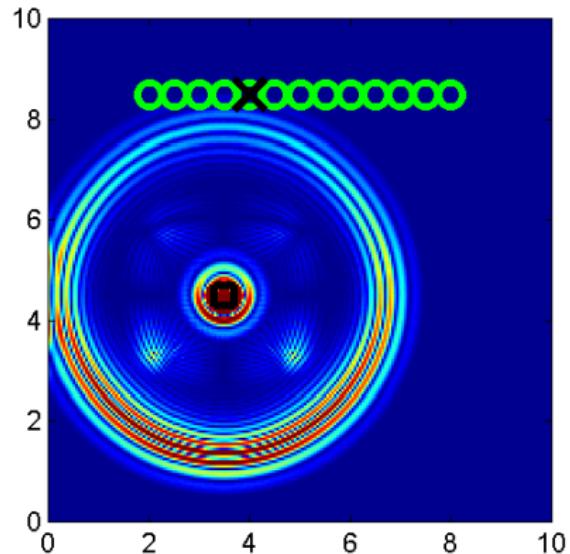
total field



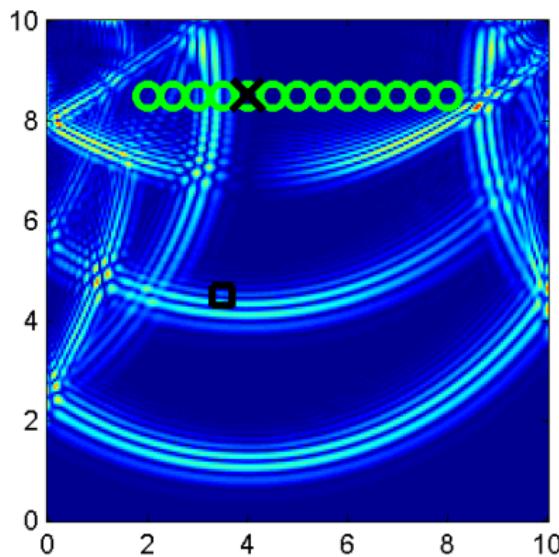
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



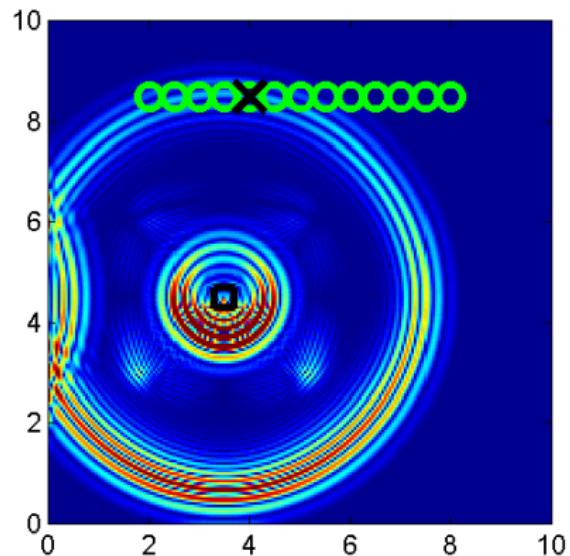
total field



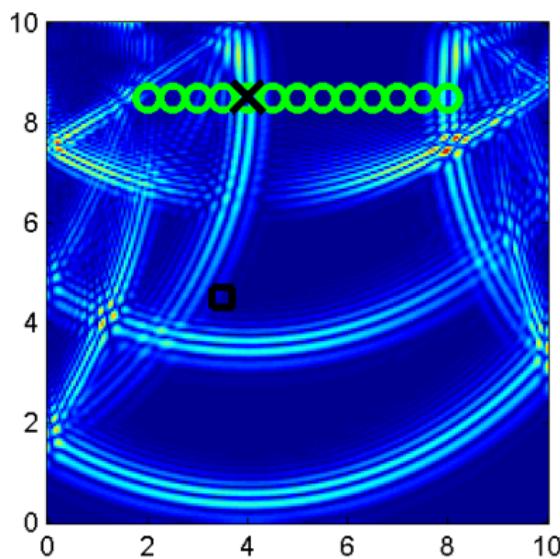
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



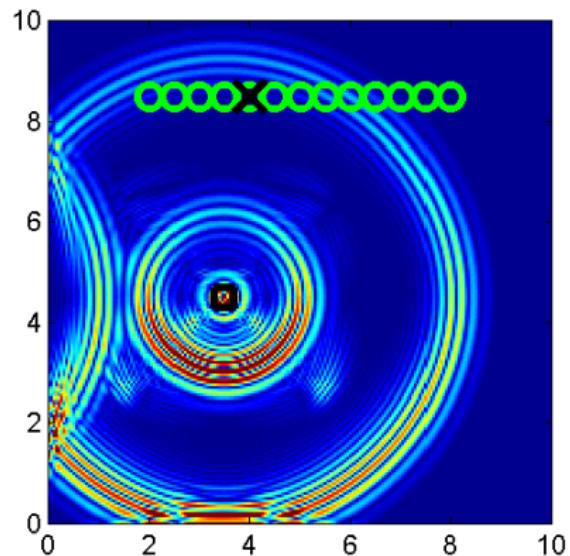
total field



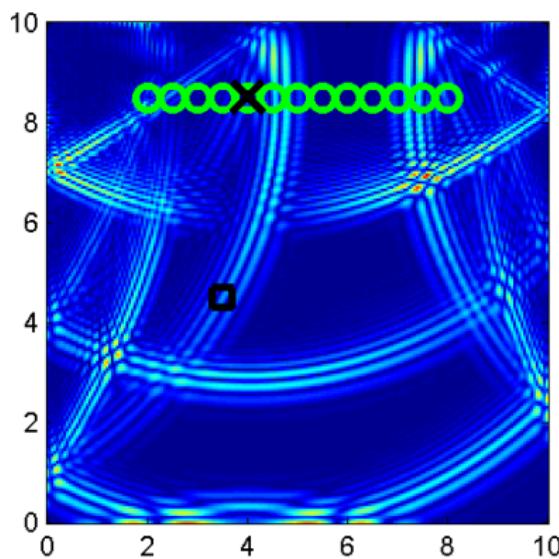
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



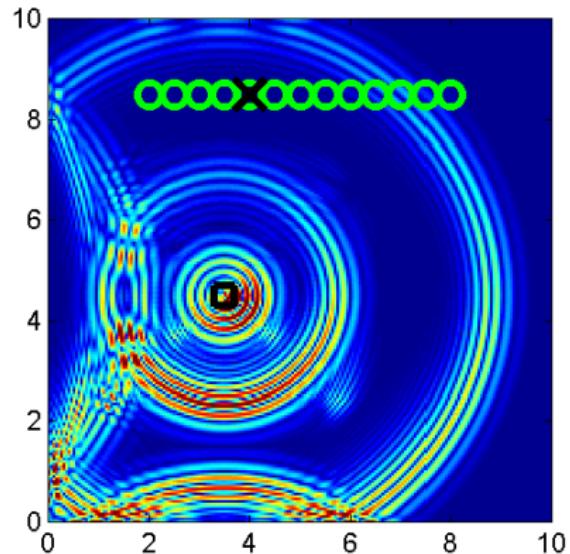
total field



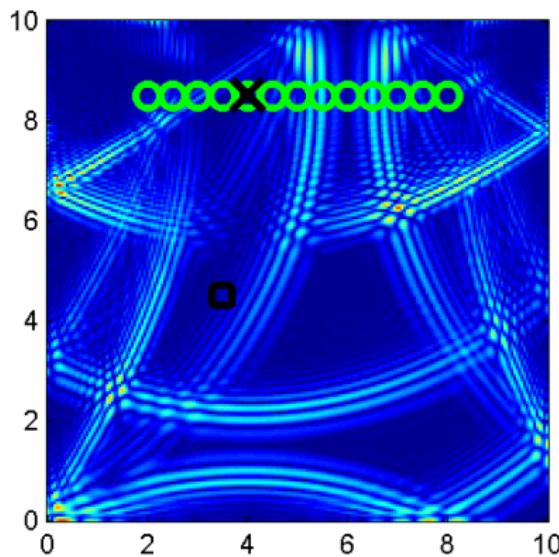
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



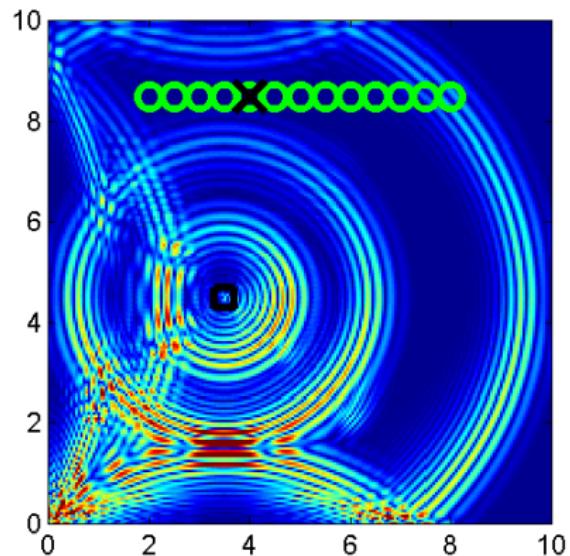
total field



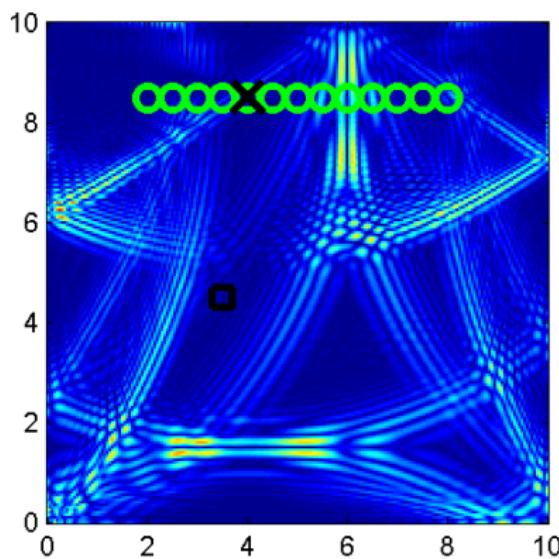
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

scattered field



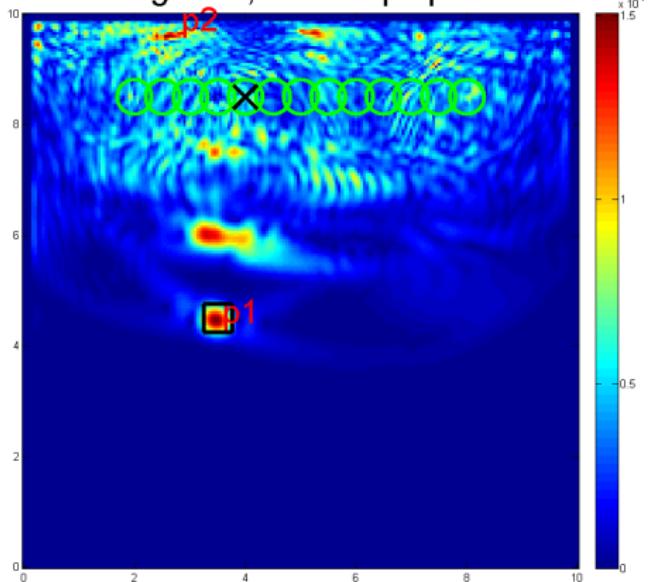
total field



Numerical example - defect localization

1 defect, 1 source and array of 13 receivers

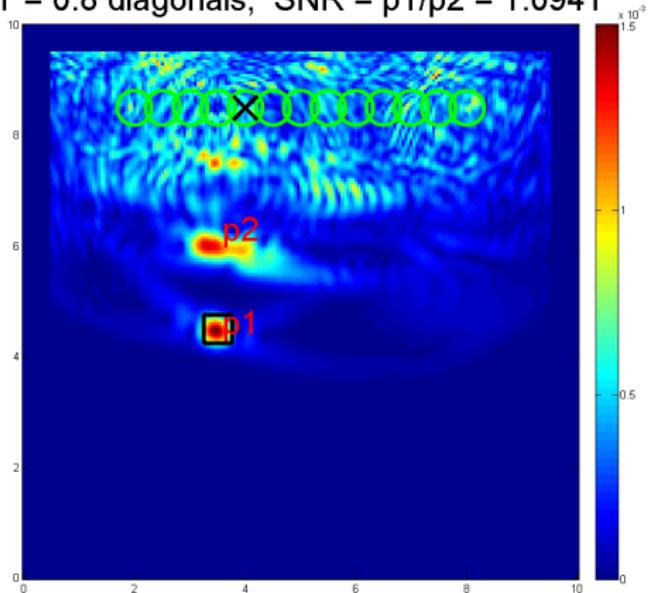
$T = 0.8$ diagonals, $\text{SNR} = p_1/p_2 = 1.0163$



Numerical example - defect localization

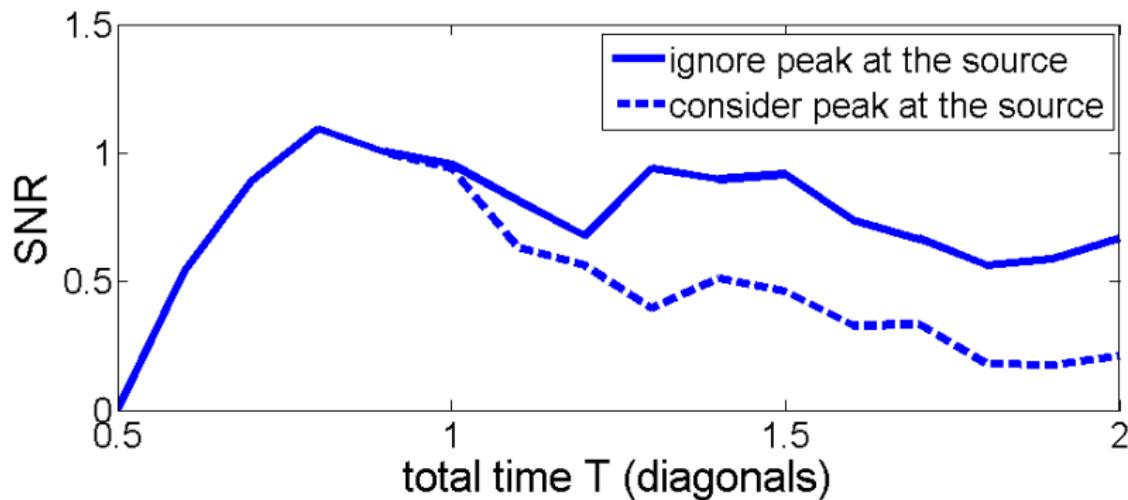
1 defect, 1 source and array of 13 receivers

$T = 0.8$ diagonals, $\text{SNR} = p_1/p_2 = 1.0941$



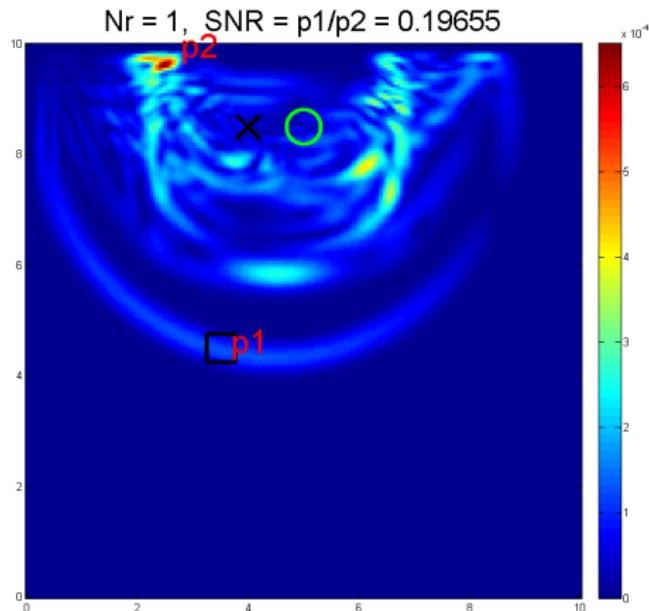
Numerical example - defect localization

1 defect, 1 source and array of 13 receivers



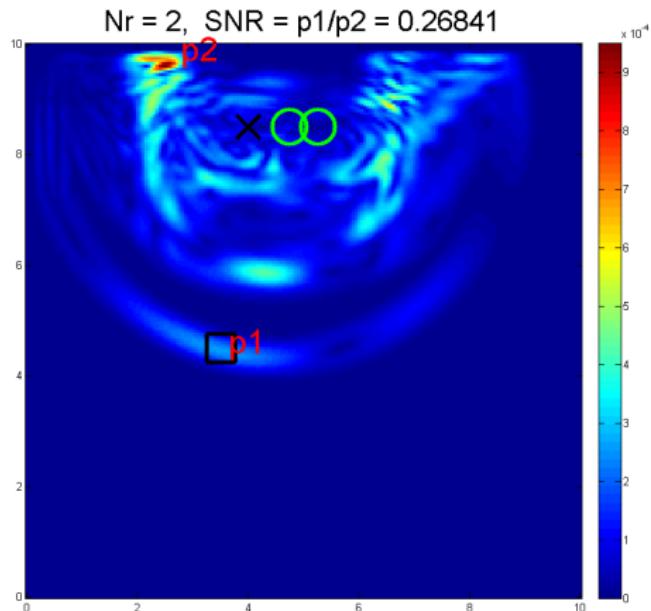
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



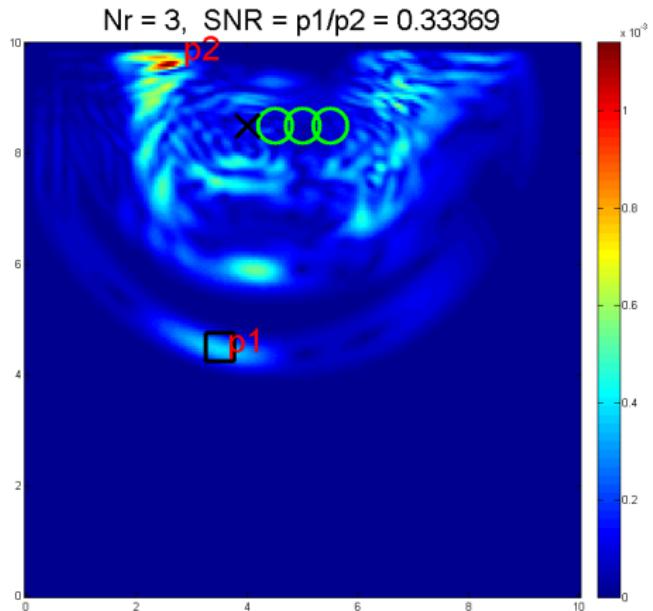
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



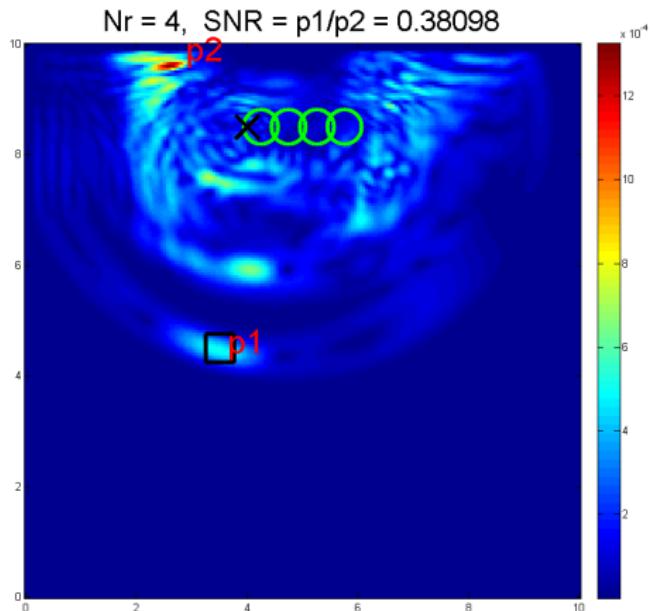
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



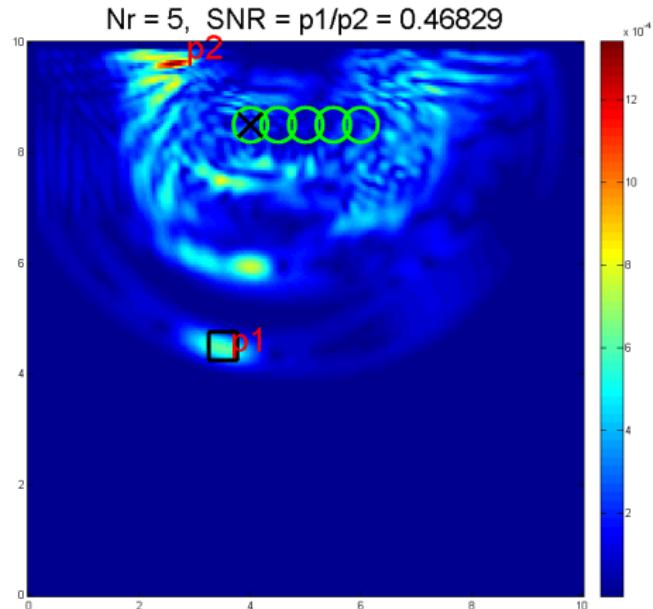
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



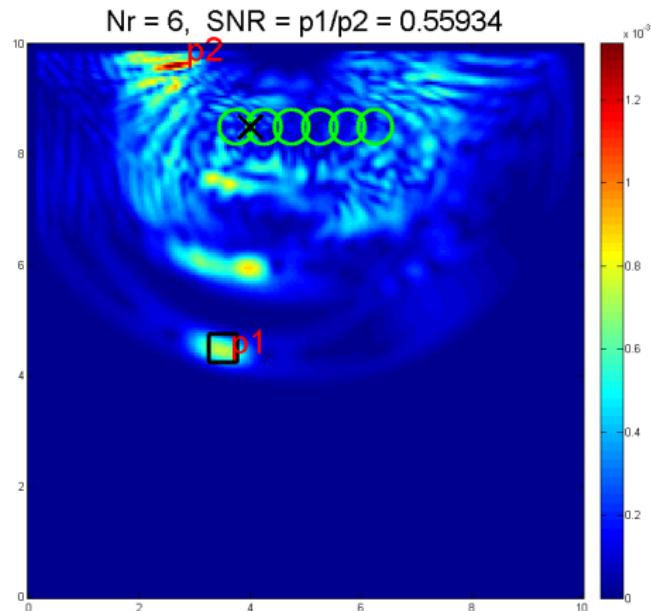
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



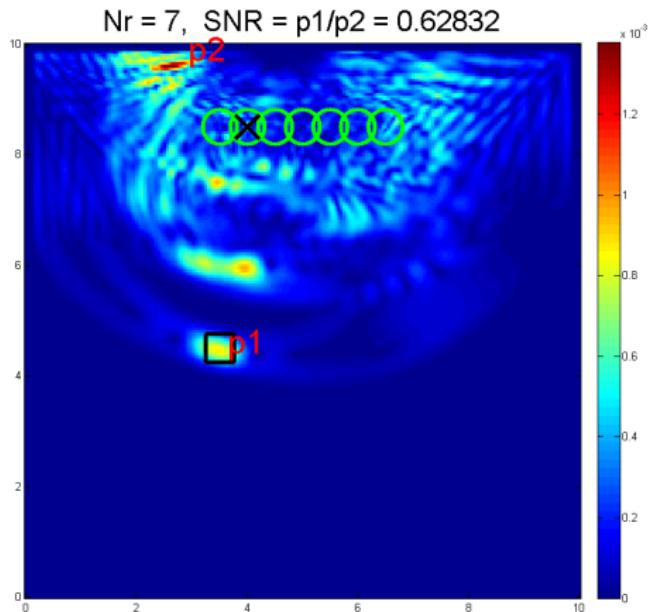
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



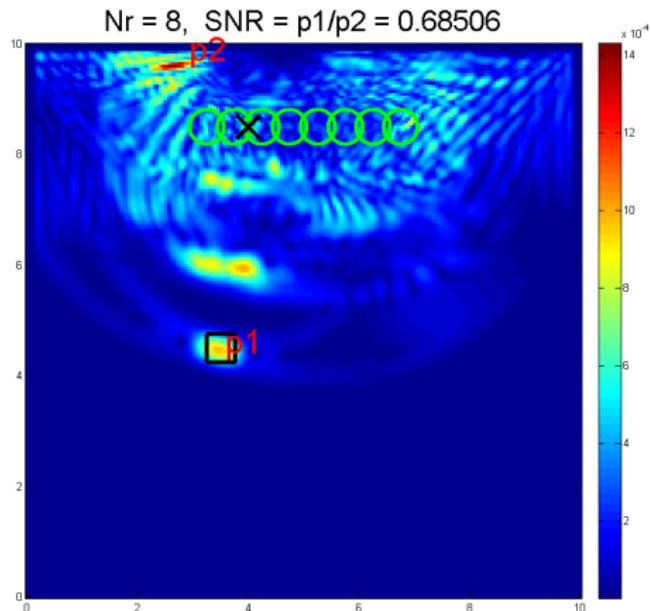
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



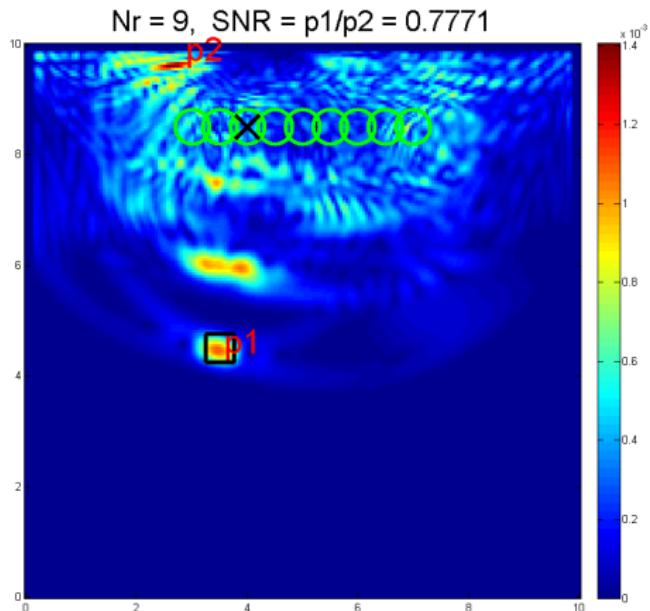
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



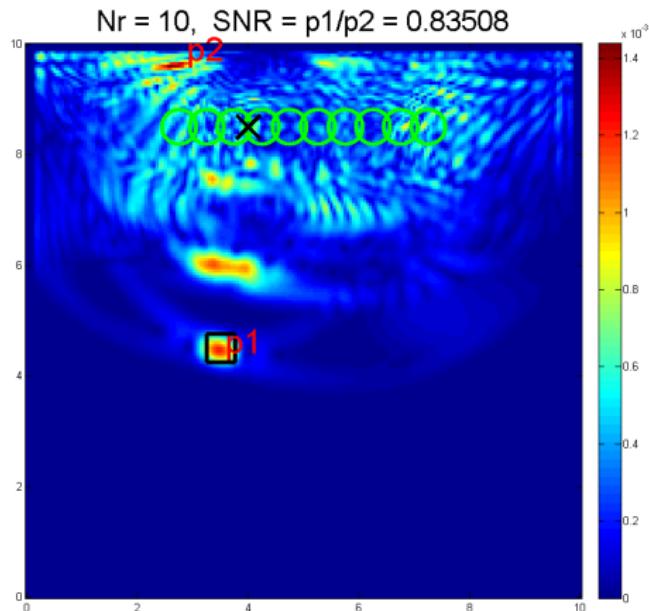
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



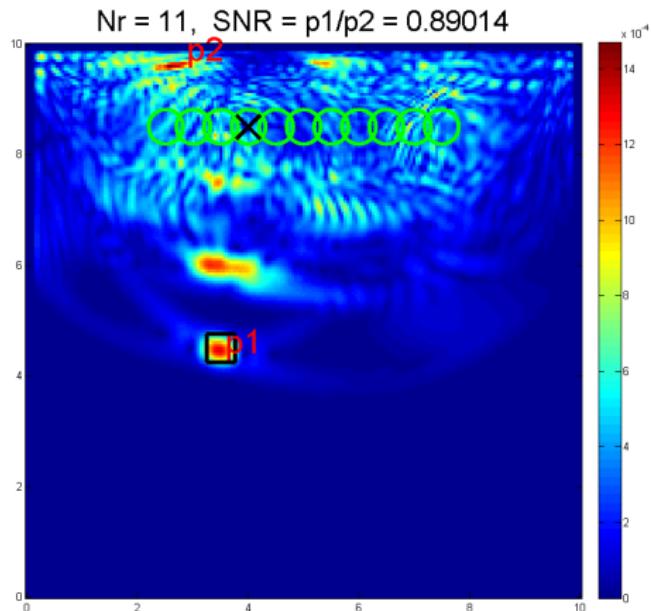
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



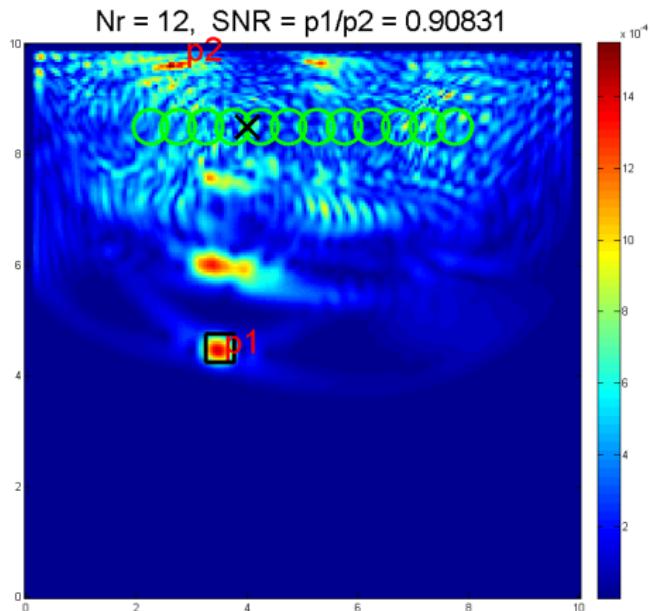
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



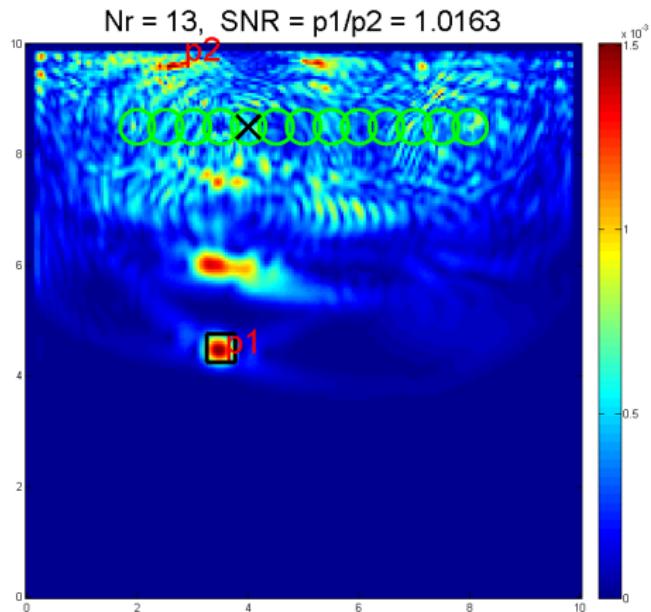
Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



Numerical example - defect localization

1 defect, 1 source and increasing number of receivers



Numerical example - defect localization

1 defect, 1 source and increasing number of receivers

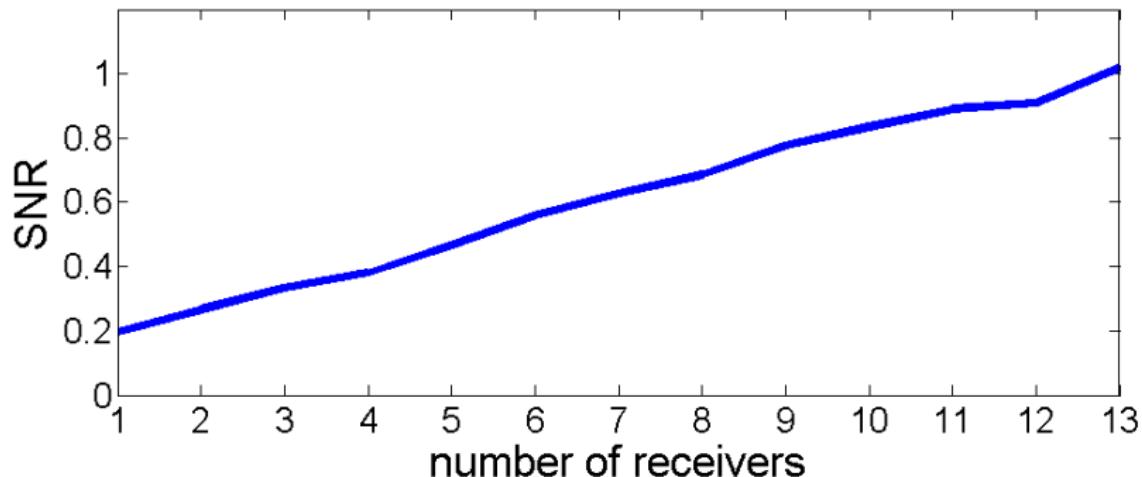


Table of contents

1 Introduction

2 Source localization

3 Defect localization

4 Numerical examples

5 Conclusions

Summary and Conclusions

- Application of TR based imaging techniques⁶ for source and scatterer localization in elastic bounded domains
- Very efficient compared to TR, the Green's functions are compute only once
- Difficulties in the elastic medium due to the two types of waves (pressure and shear) and their conversions
- Source localization :
 - sensor configuration : distributed or array
 - steady increase and convergence of SNR for increasing total time T
 - approximately linear increase of SNR for increasing number of sensors
 - Boundaries : positive influence
- Defect localization :
 - sensor configuration : array
 - total time T is very important, should be carefully chosen
 - approximately linear increase of SNR for increasing number of sensors
 - Boundaries : negative influence

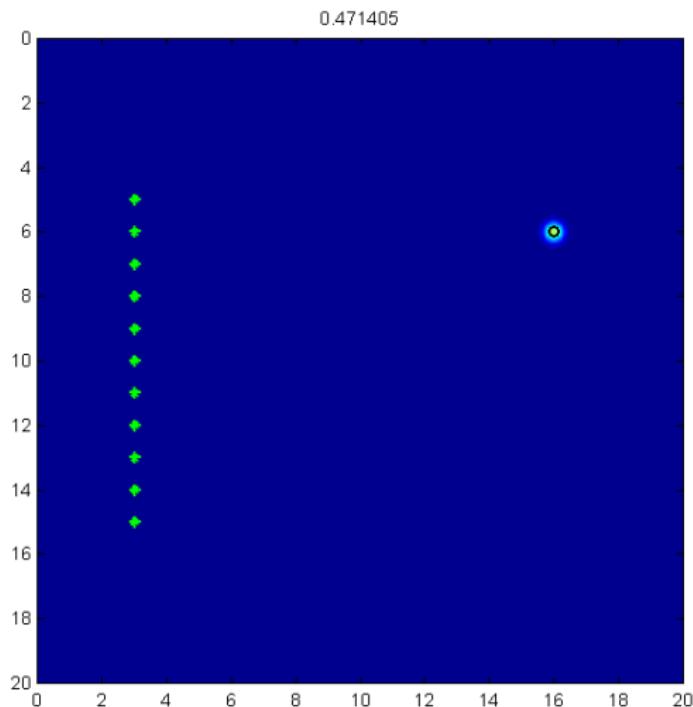
6. L Borcea, G Papanicolaou, C Tsogka and J Berryman, (2002) Imaging and time reversal in random media

Future work

- Extensive investigation of the distributed sensor configuration
- Propose optimal total experiment time
- Investigation of the methodology using passive noisy recordings as input data
- Account for dissipation (damping) and dispersion
- Application to structures with complex geometry

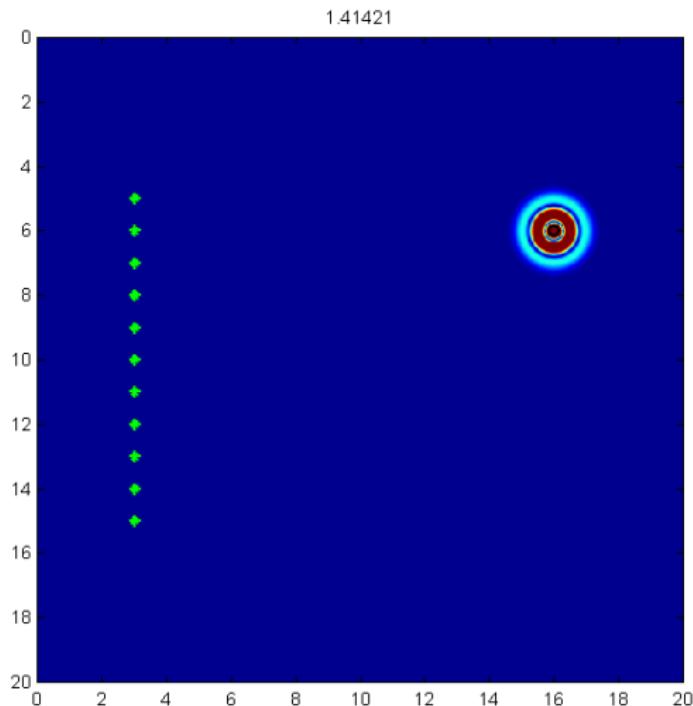
Thank you !

Source localization - Forward step



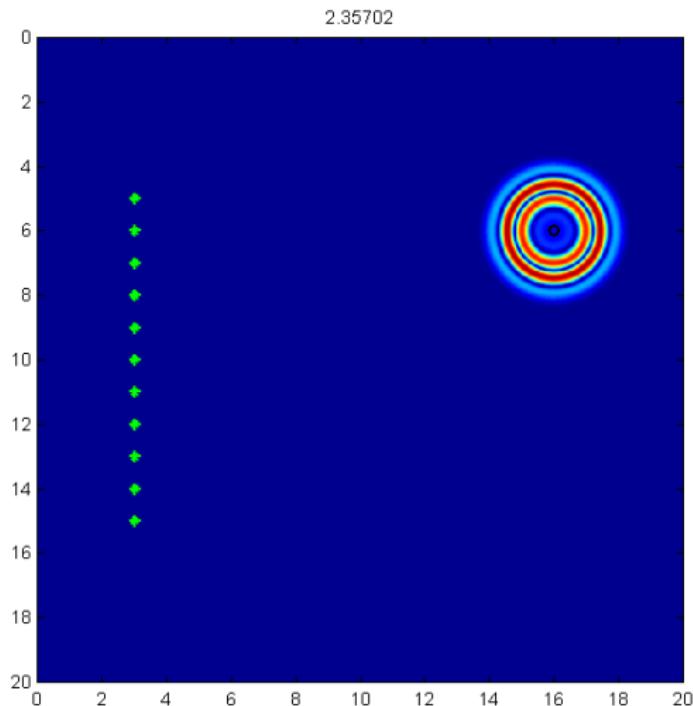
main

Source localization - Forward step



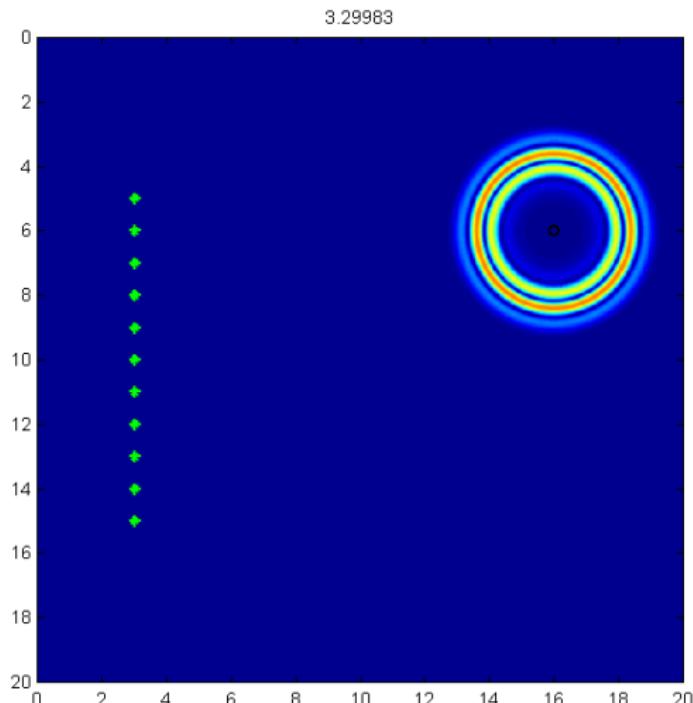
main

Source localization - Forward step



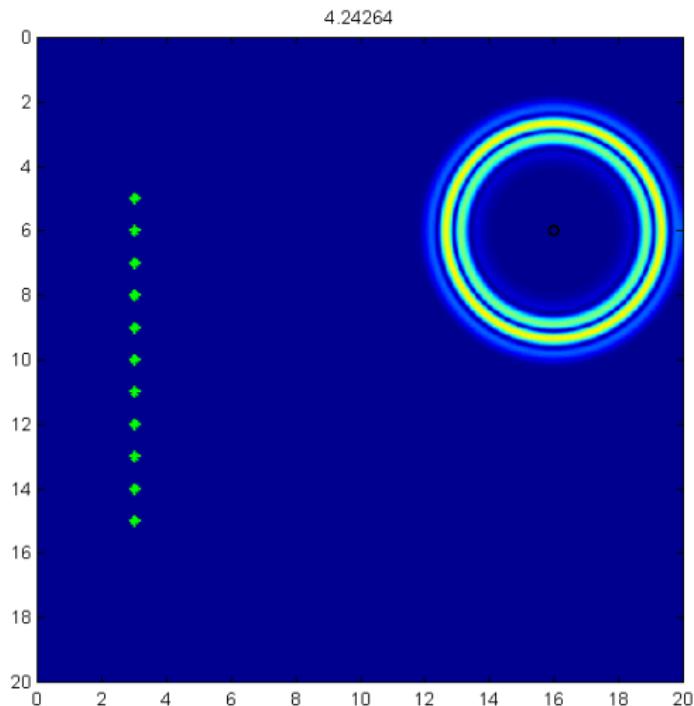
main

Source localization - Forward step



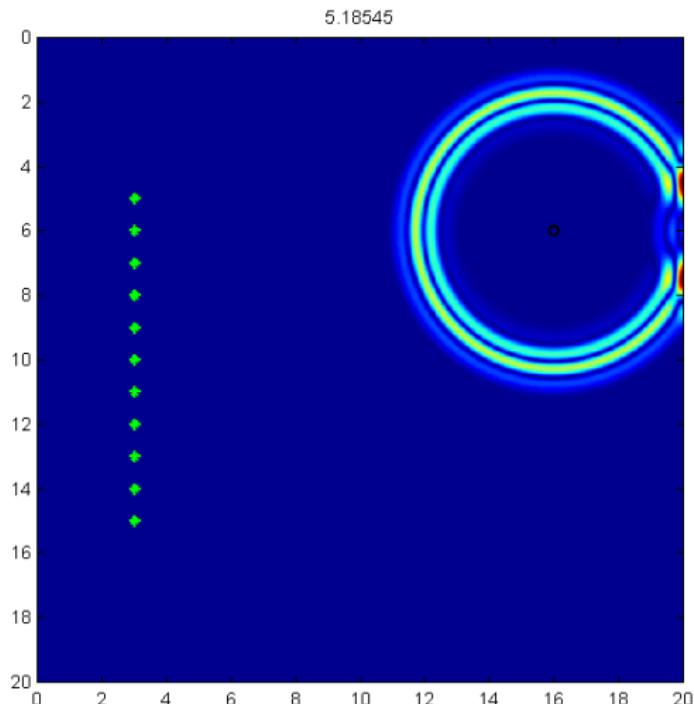
main

Source localization - Forward step



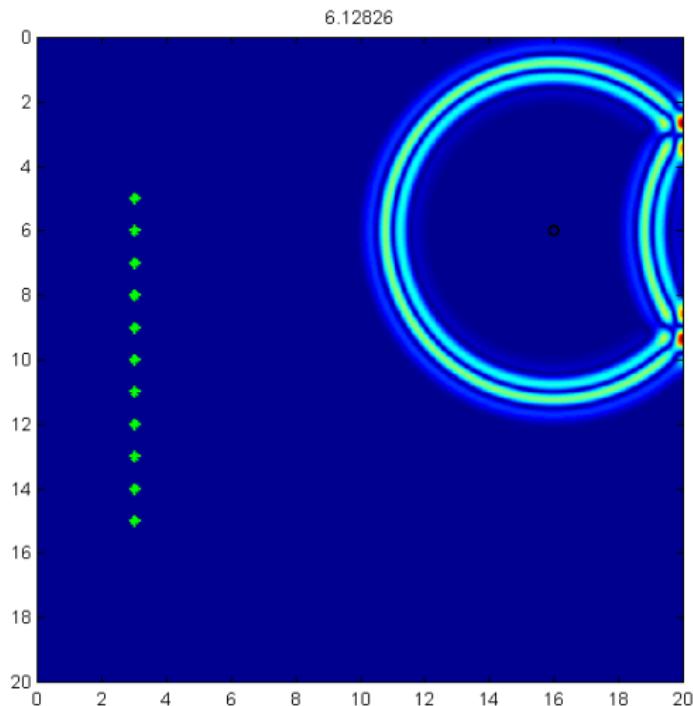
main

Source localization - Forward step



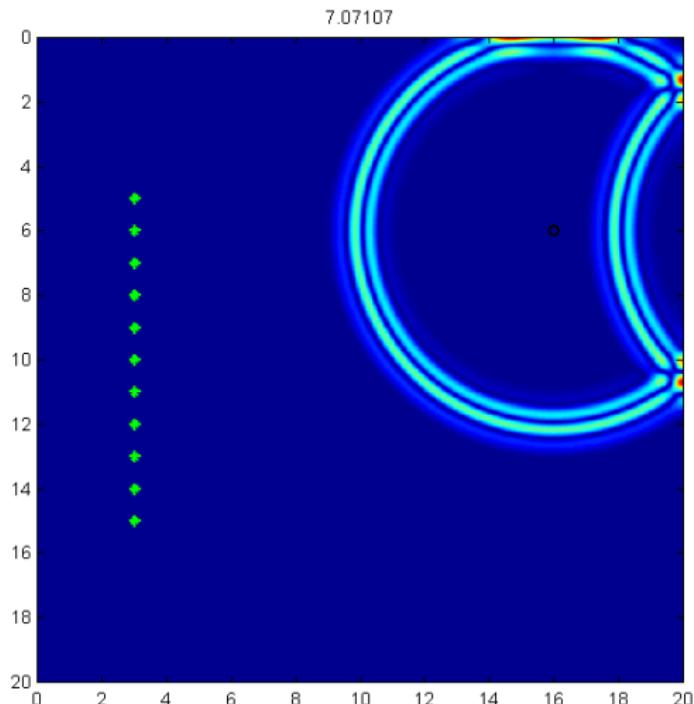
main

Source localization - Forward step



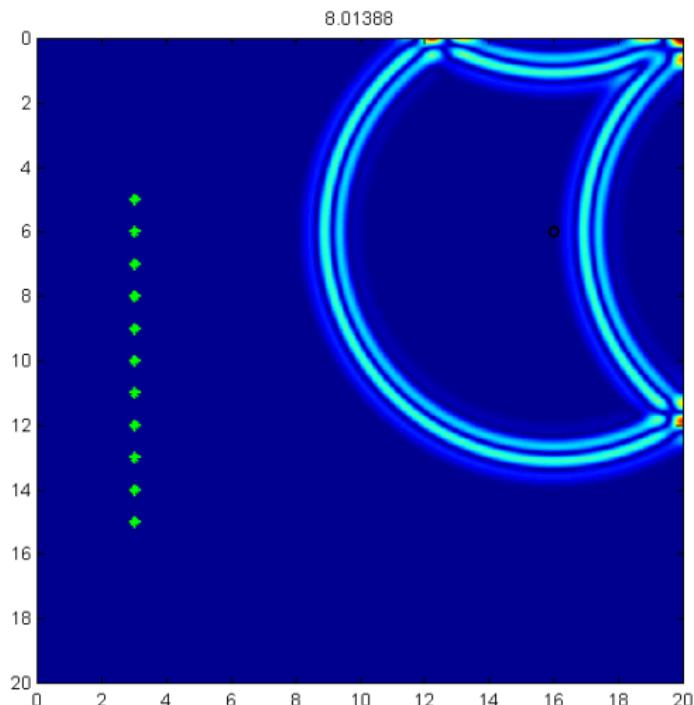
main

Source localization - Forward step



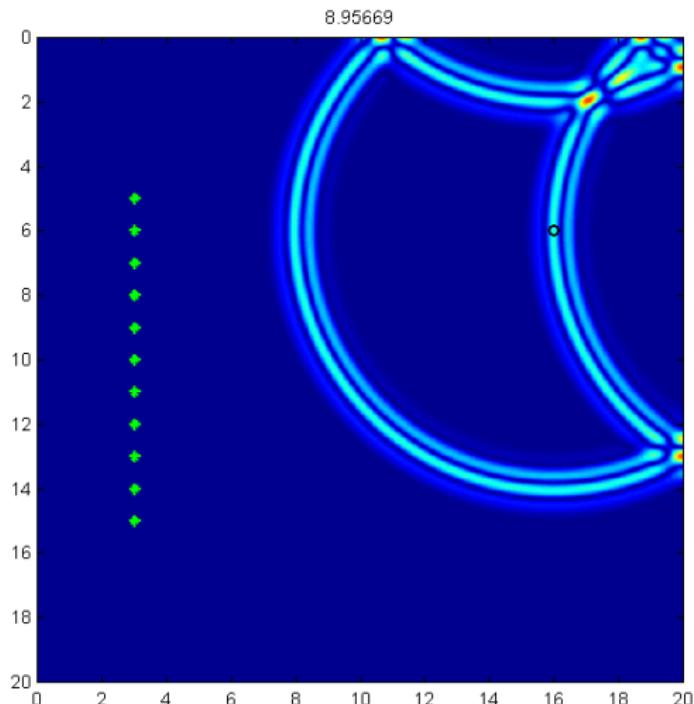
main

Source localization - Forward step



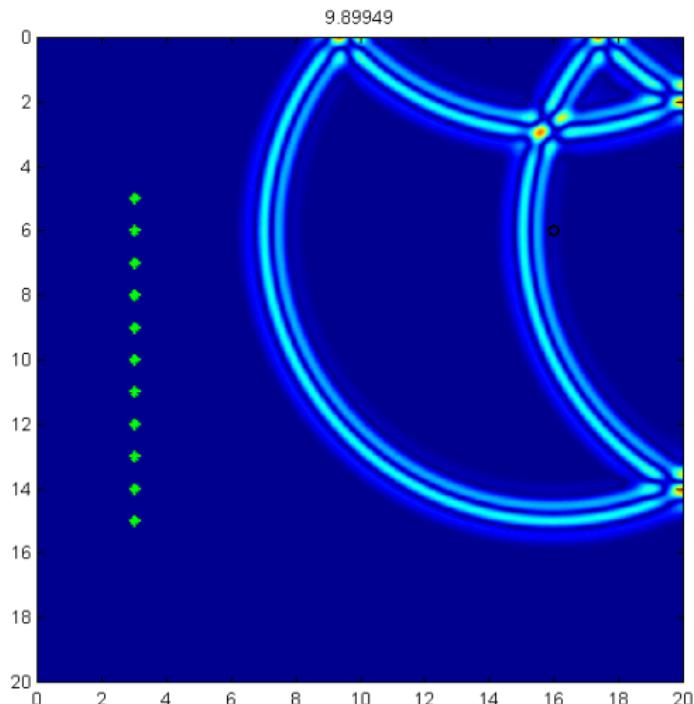
main

Source localization - Forward step



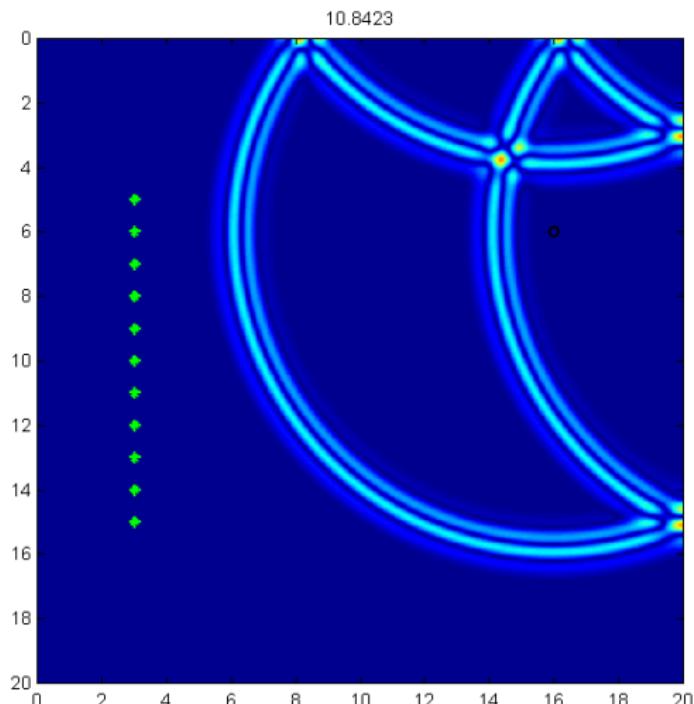
main

Source localization - Forward step



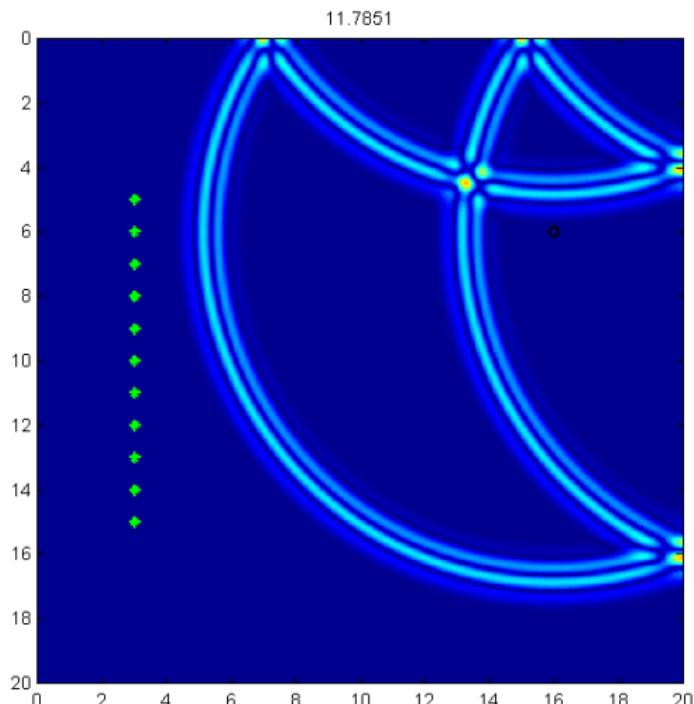
main

Source localization - Forward step



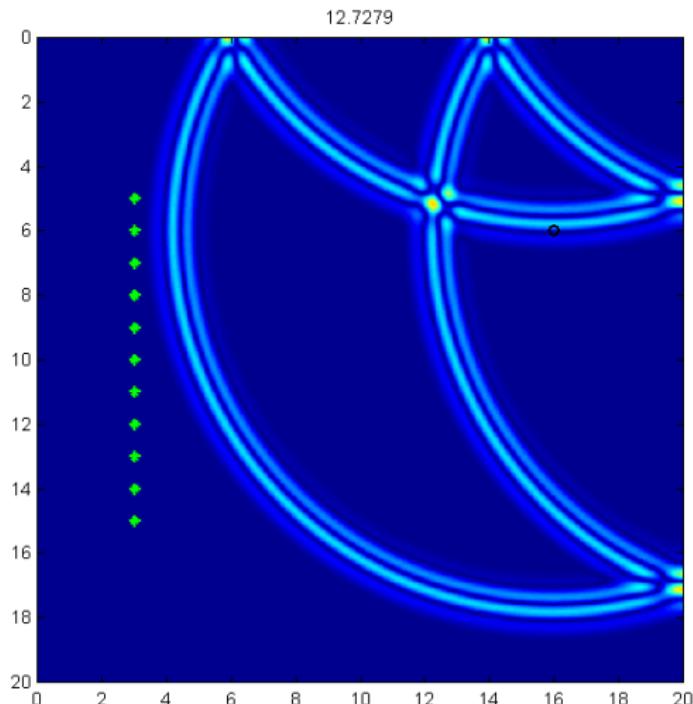
main

Source localization - Forward step



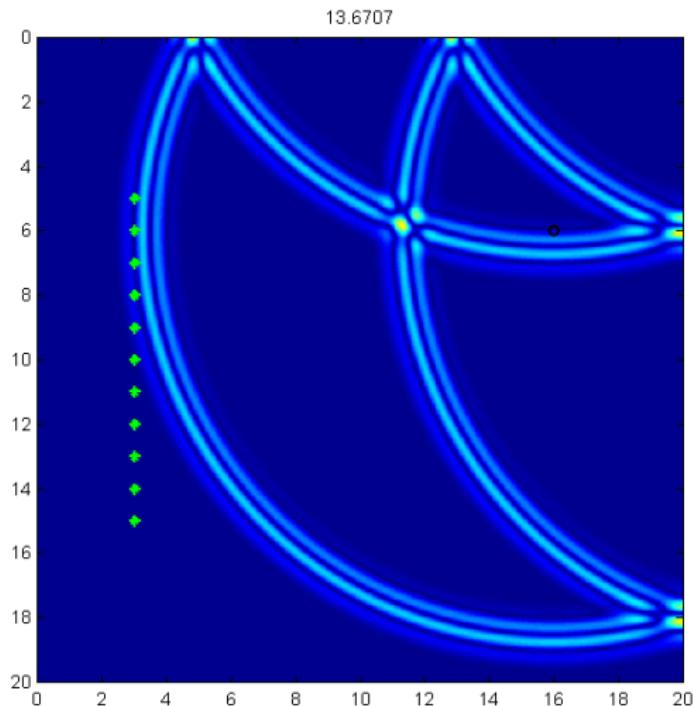
main

Source localization - Forward step



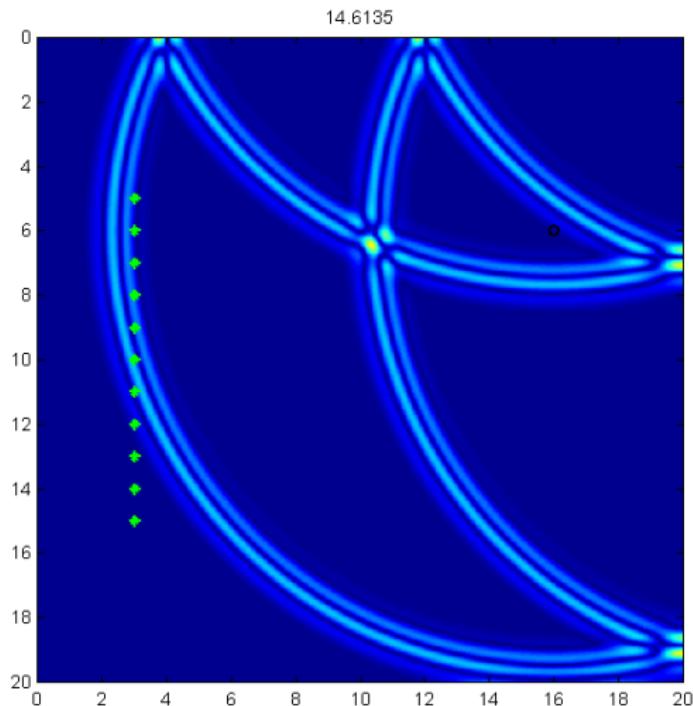
main

Source localization - Forward step



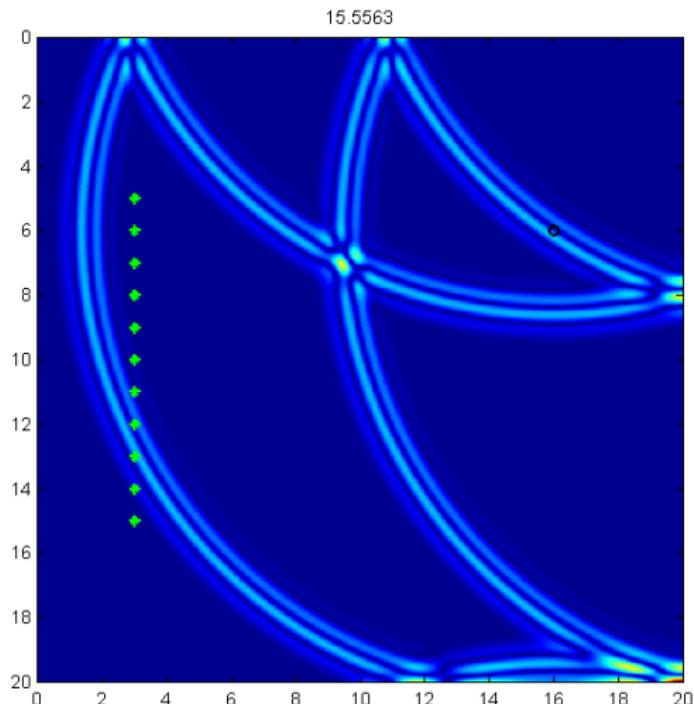
main

Source localization - Forward step



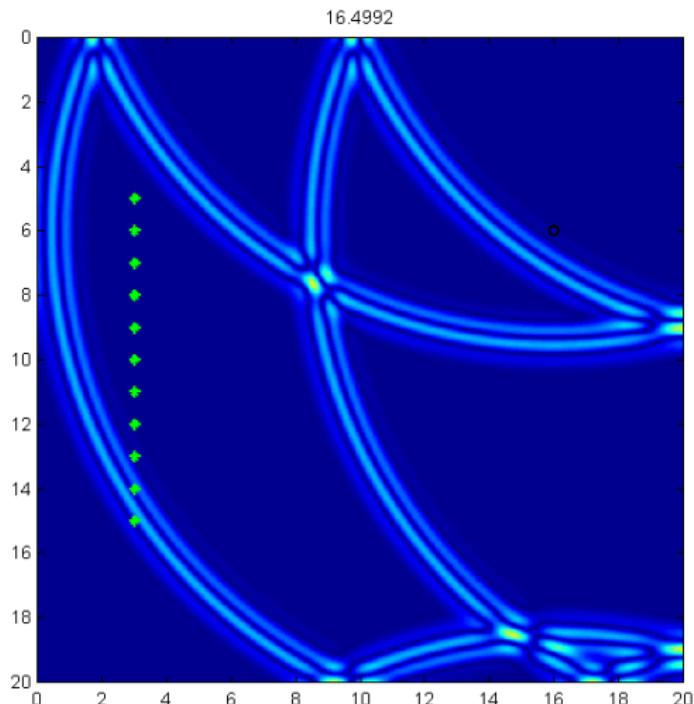
main

Source localization - Forward step

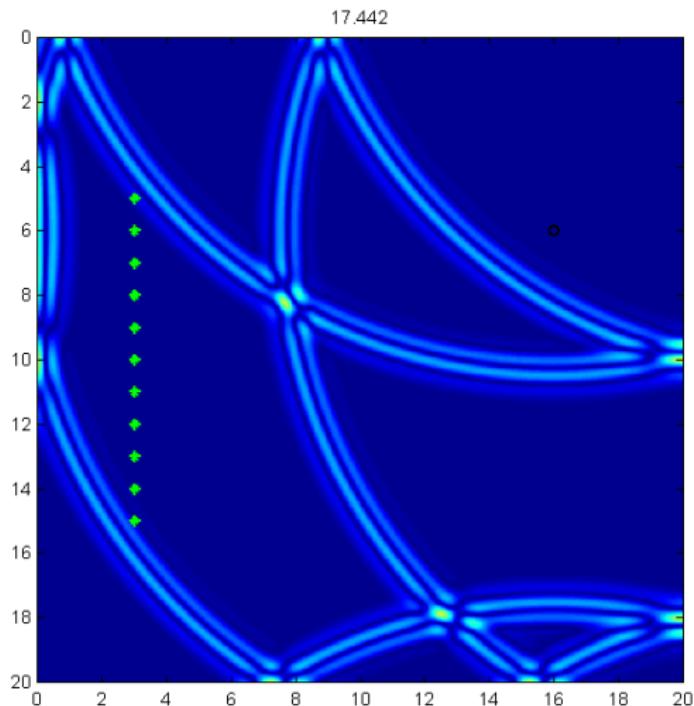


main

Source localization - Forward step

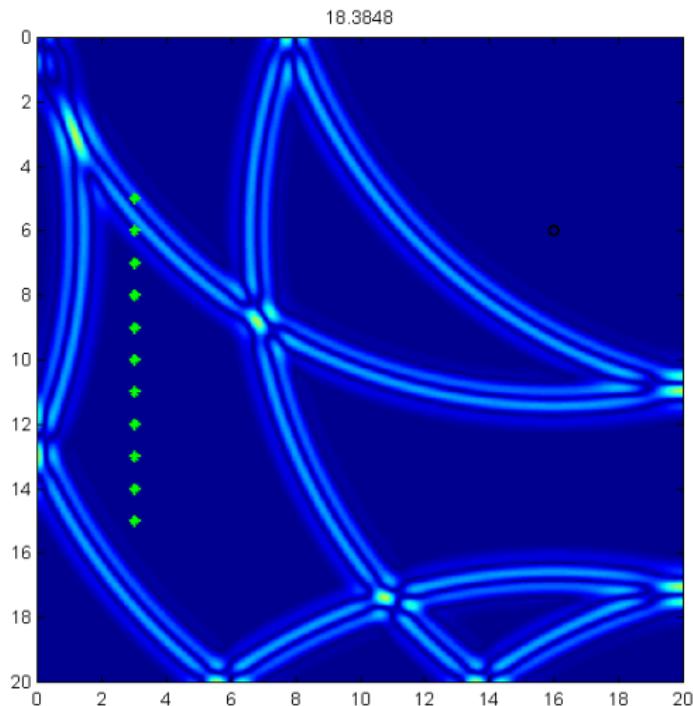


Source localization - Forward step

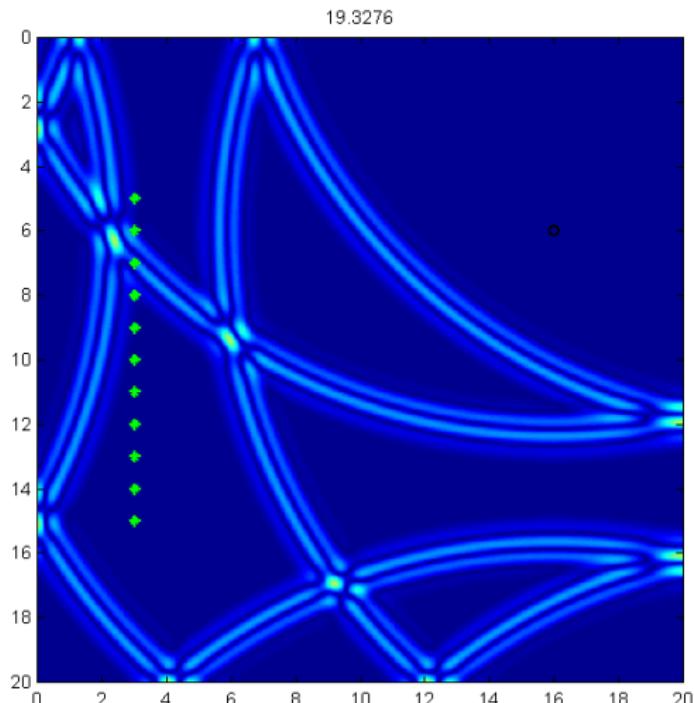


main

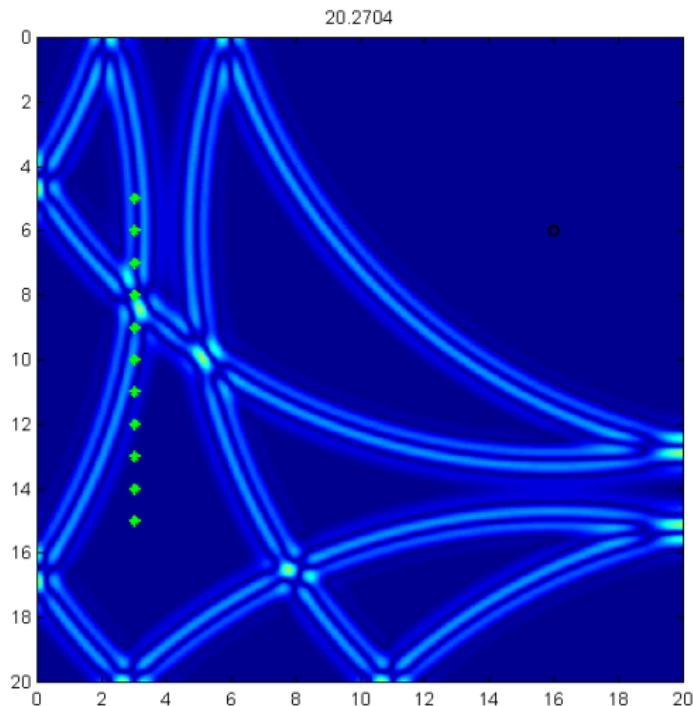
Source localization - Forward step



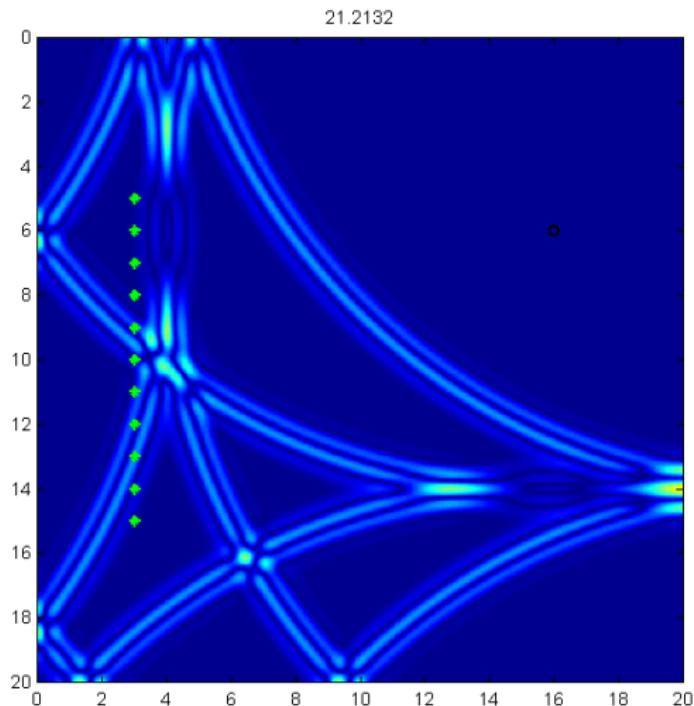
Source localization - Forward step



Source localization - Forward step

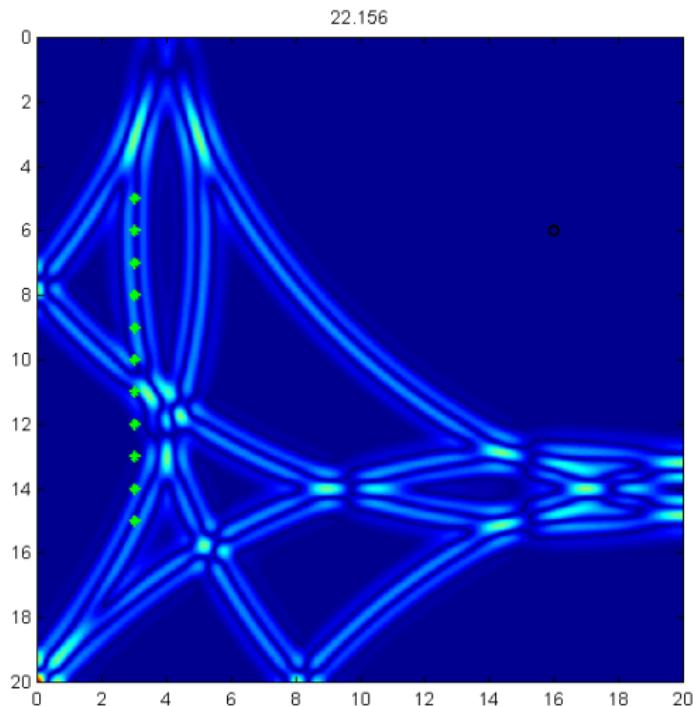


Source localization - Forward step



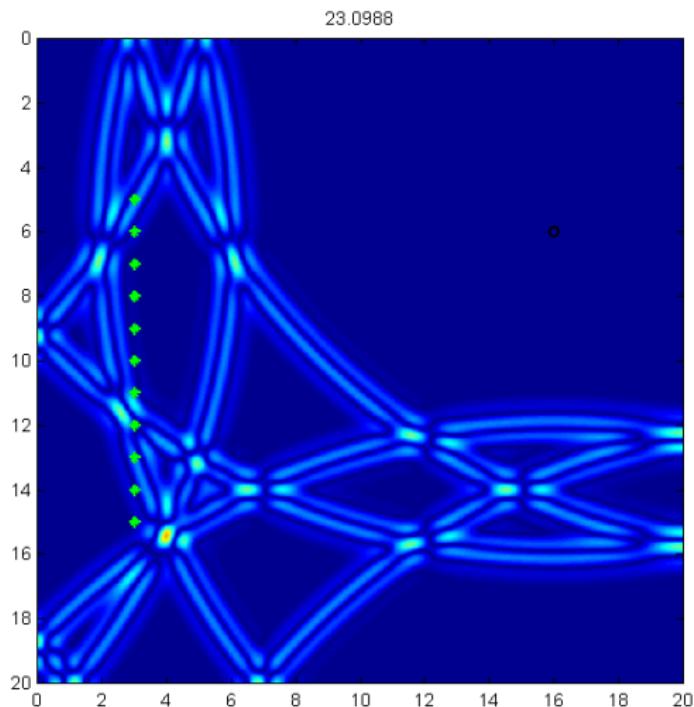
main

Source localization - Forward step

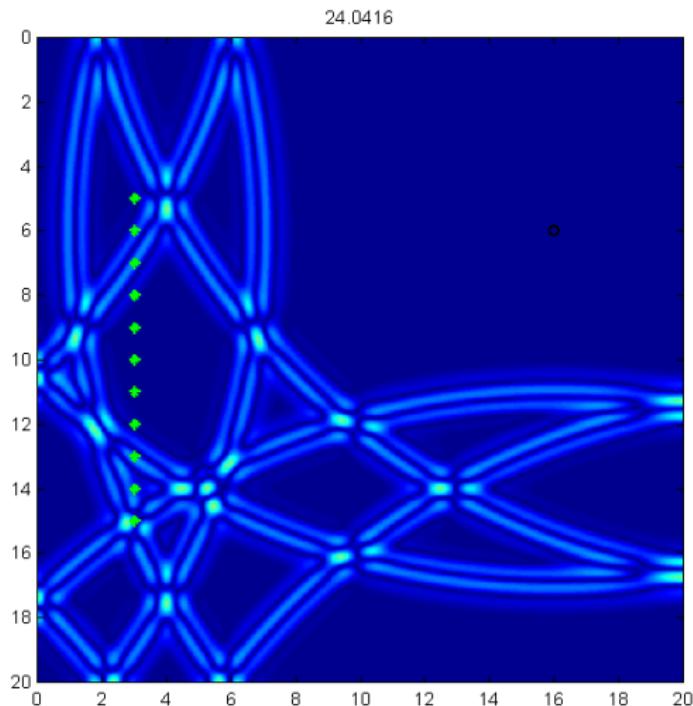


main

Source localization - Forward step

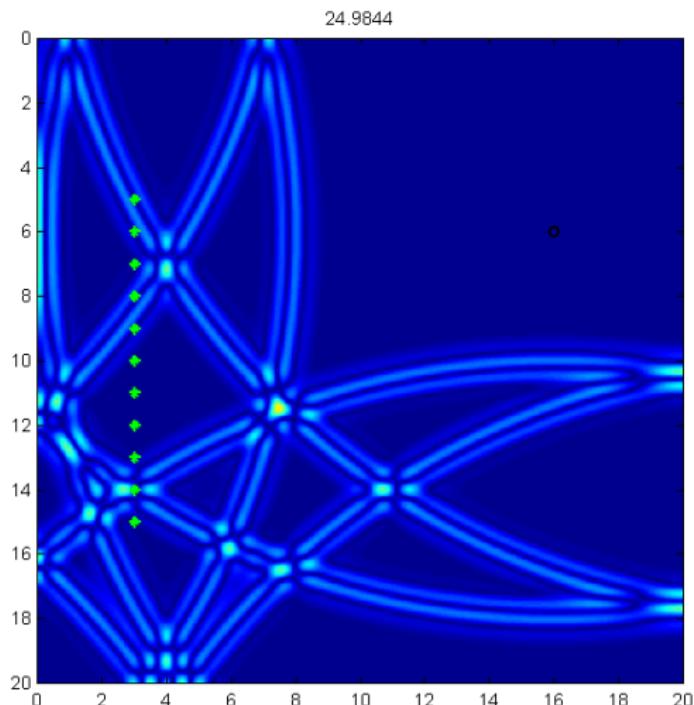


Source localization - Forward step



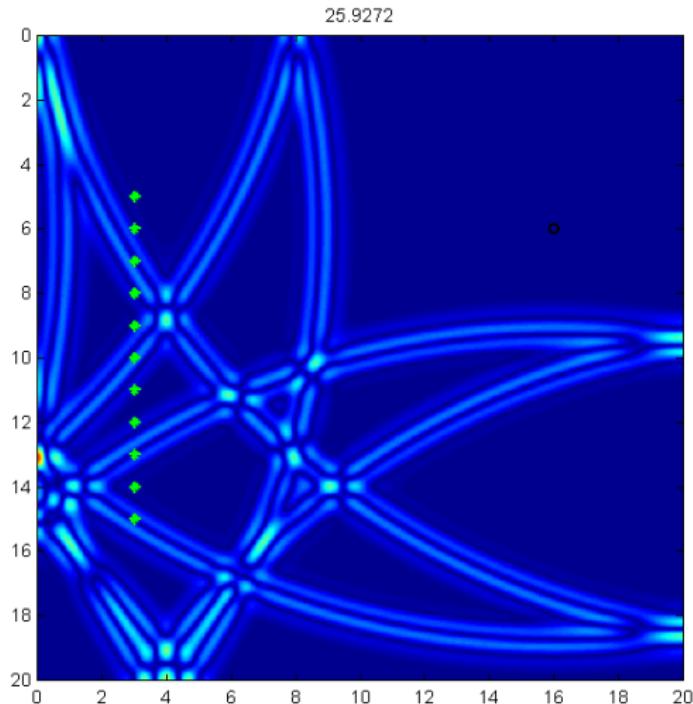
main

Source localization - Forward step



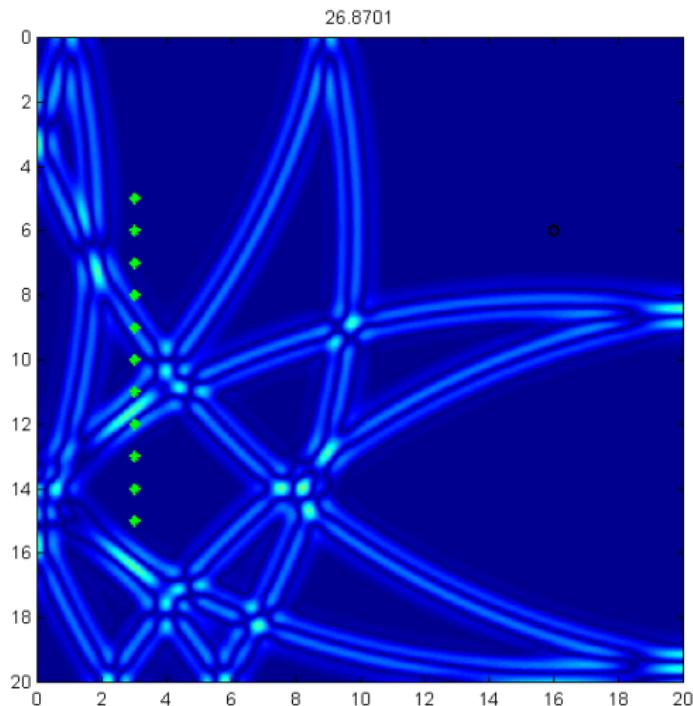
main

Source localization - Forward step



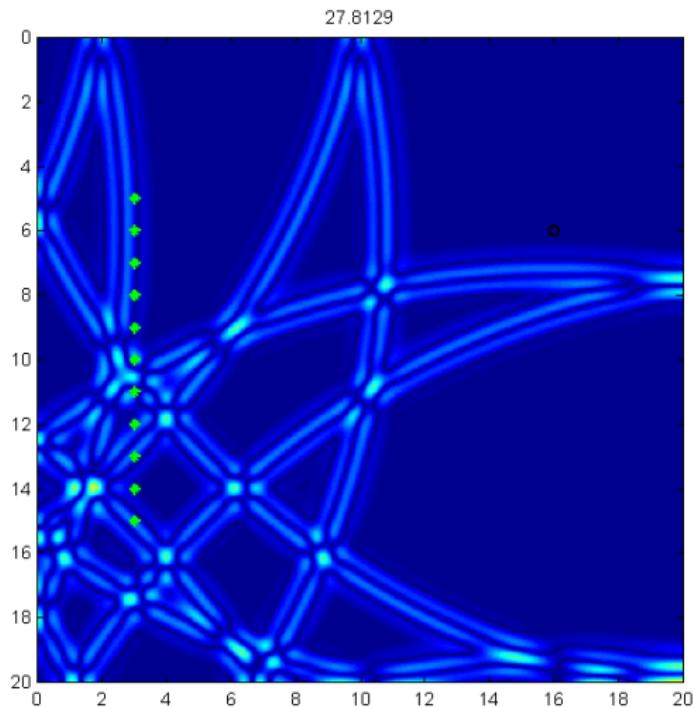
main

Source localization - Forward step



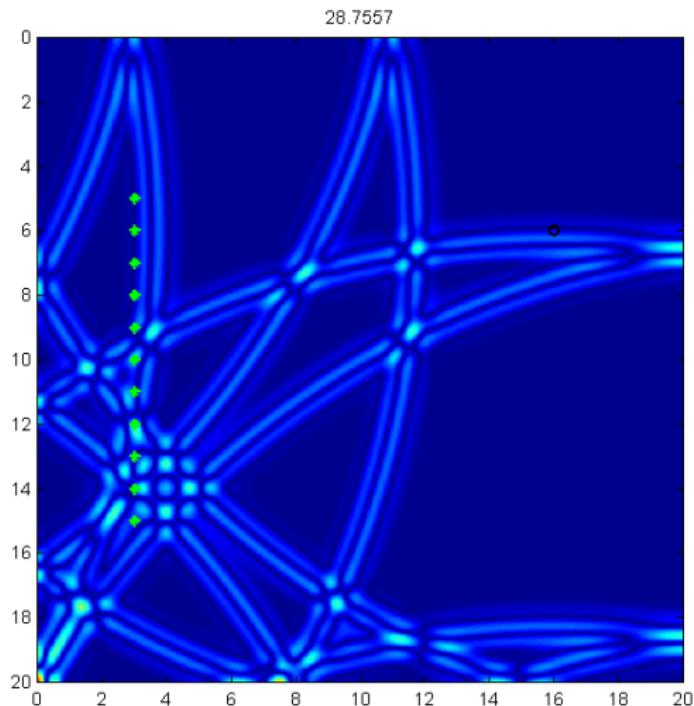
main

Source localization - Forward step



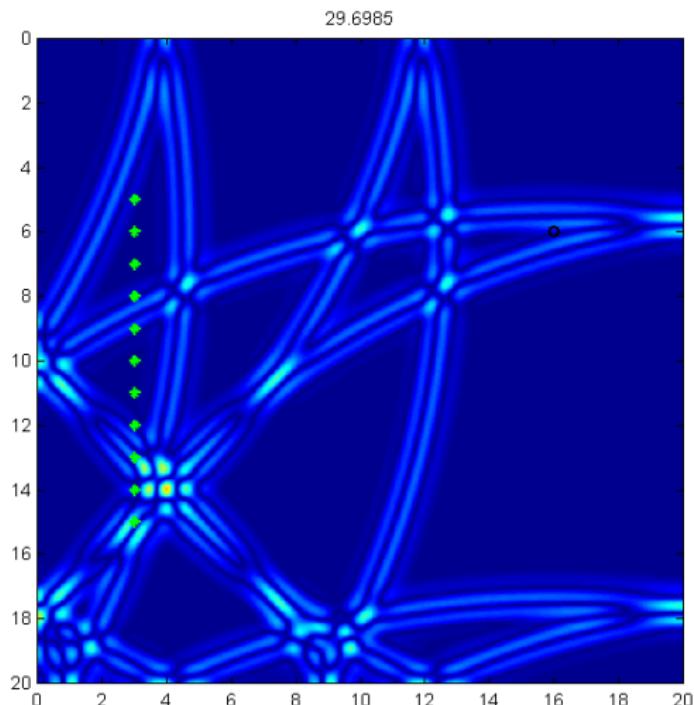
main

Source localization - Forward step



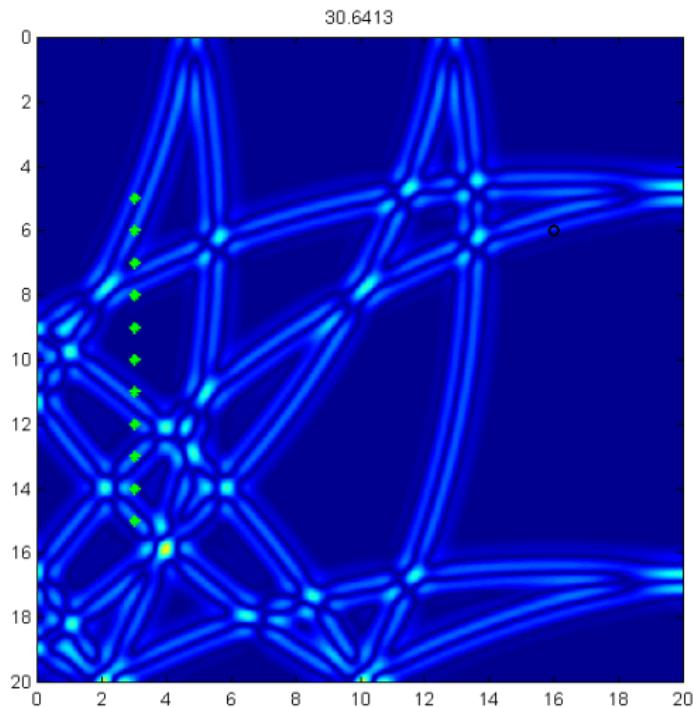
main

Source localization - Forward step



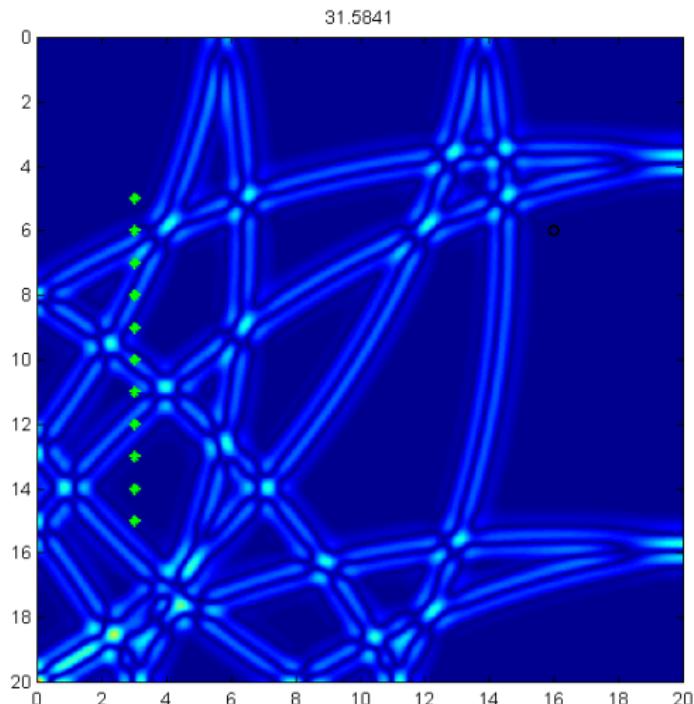
main

Source localization - Forward step



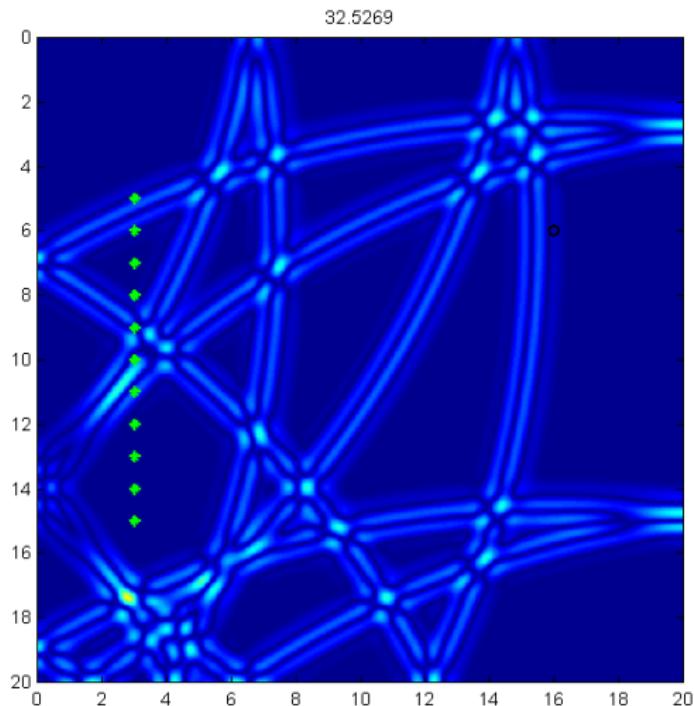
main

Source localization - Forward step



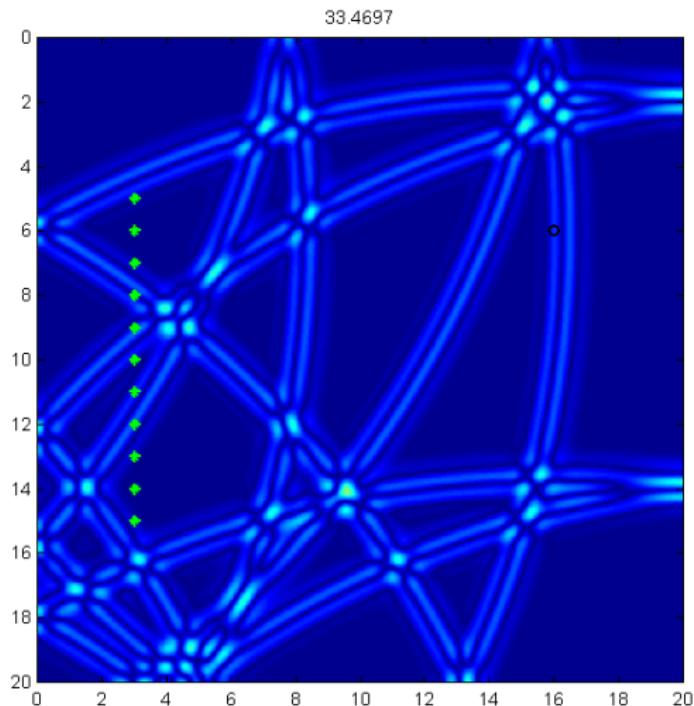
main

Source localization - Forward step



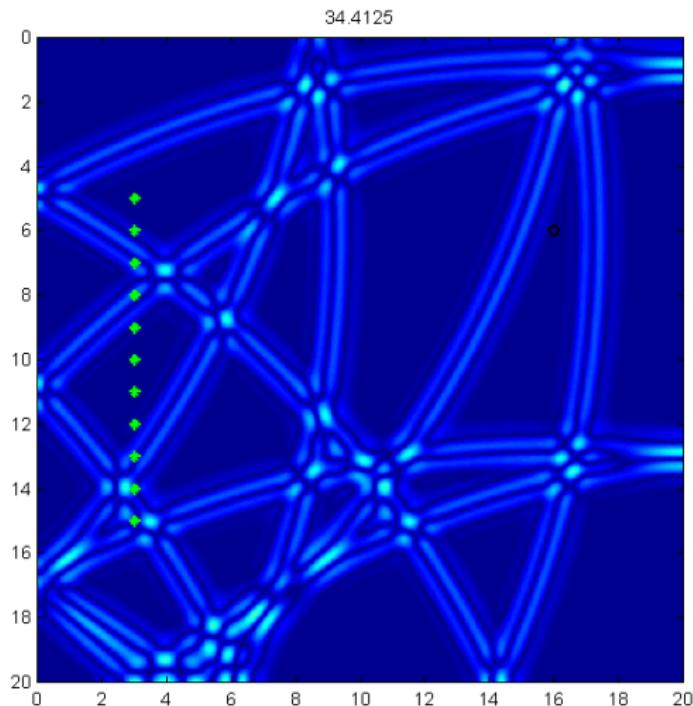
main

Source localization - Forward step

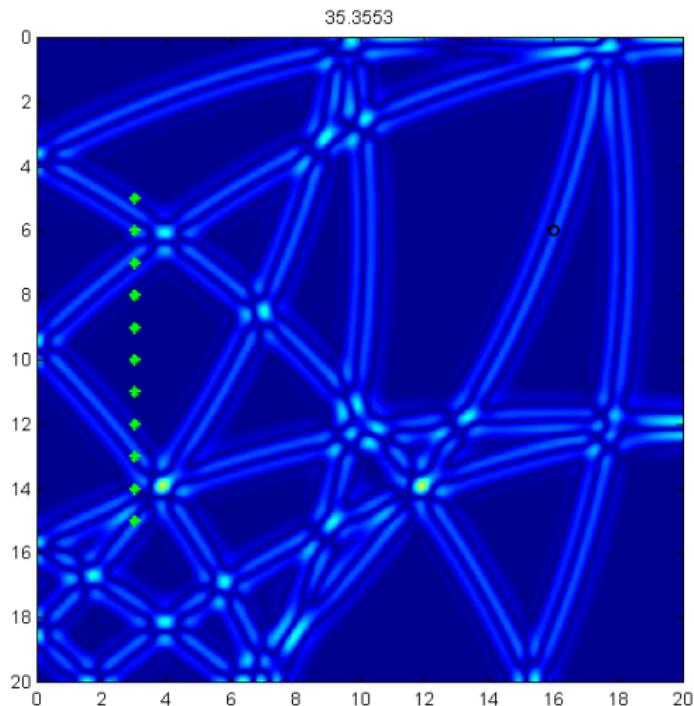


main

Source localization - Forward step

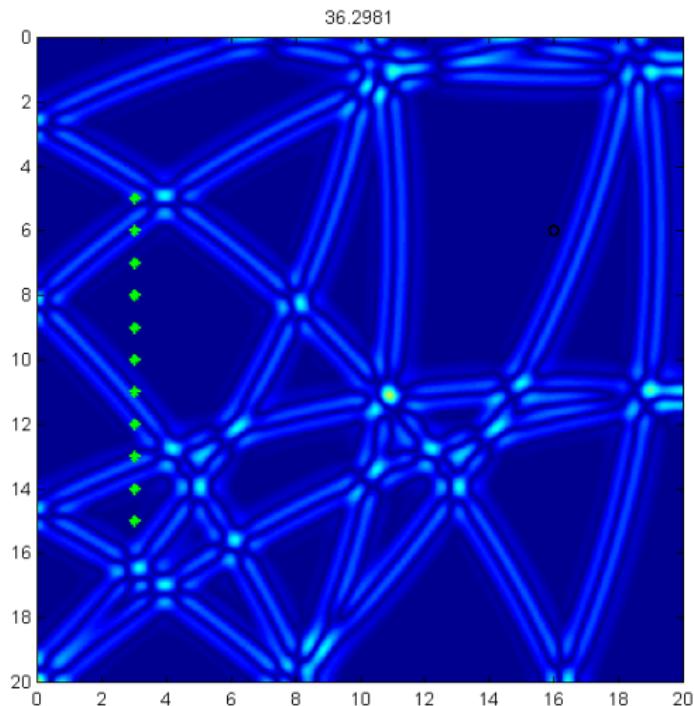


Source localization - Forward step

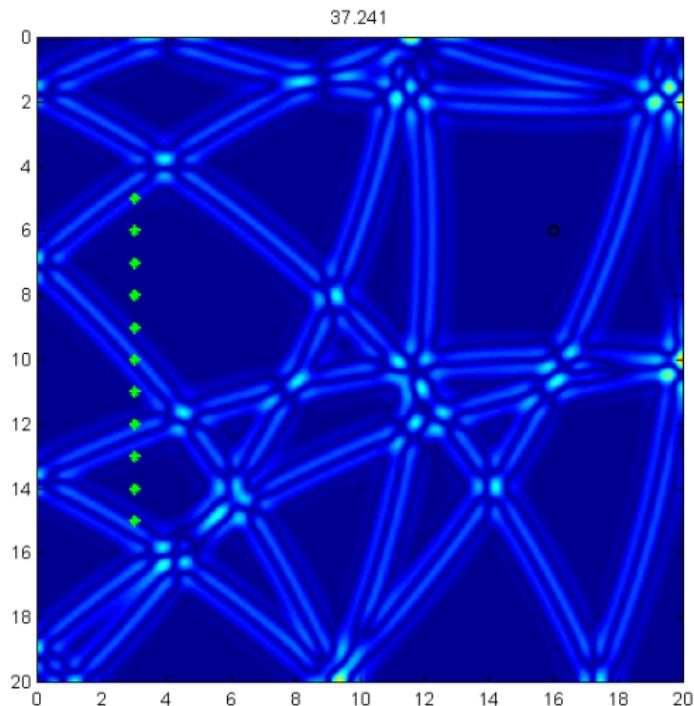


main

Source localization - Forward step

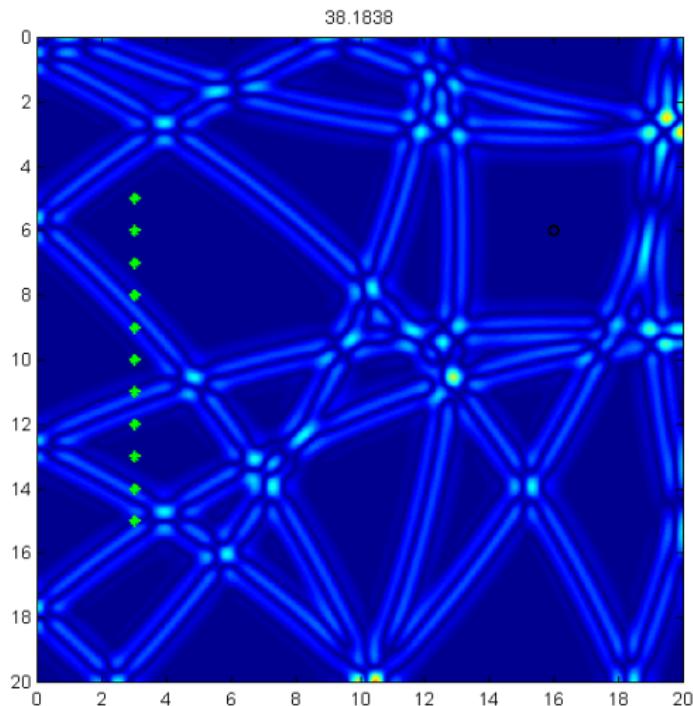


Source localization - Forward step



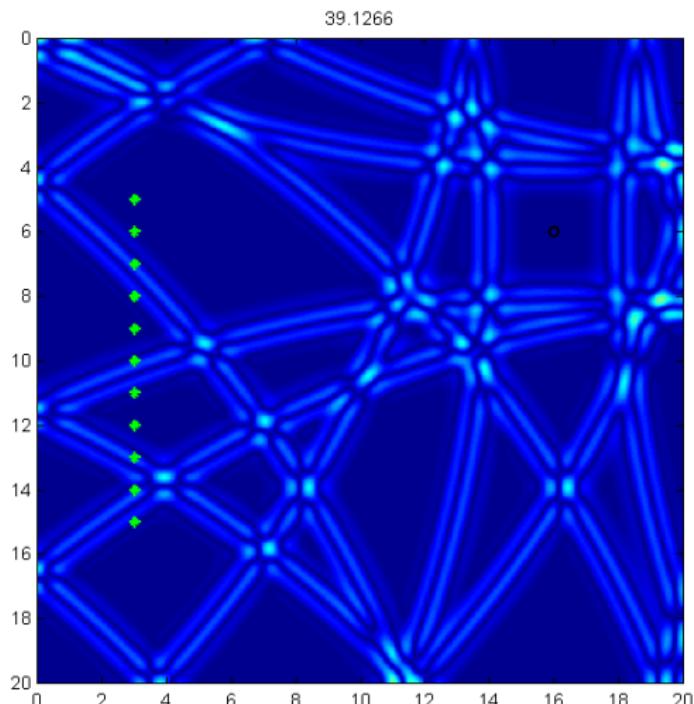
main

Source localization - Forward step



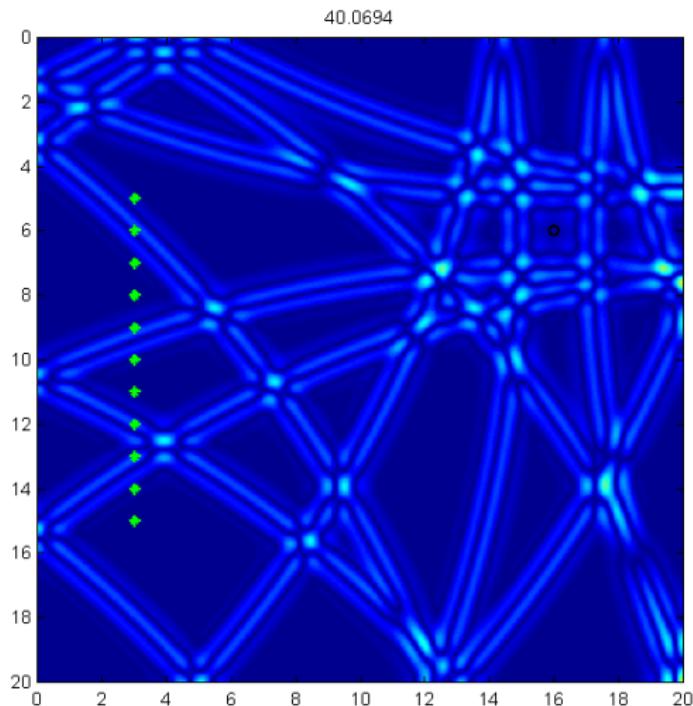
main

Source localization - Forward step



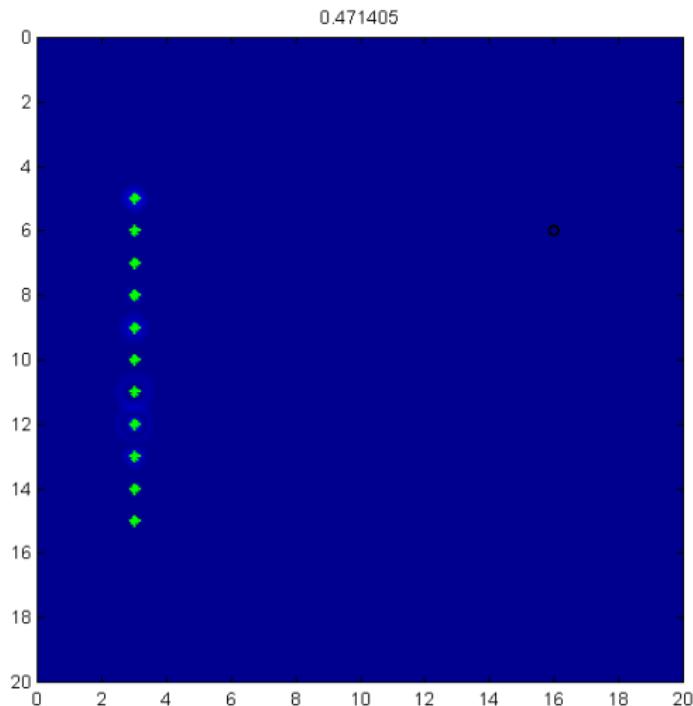
main

Source localization - Forward step

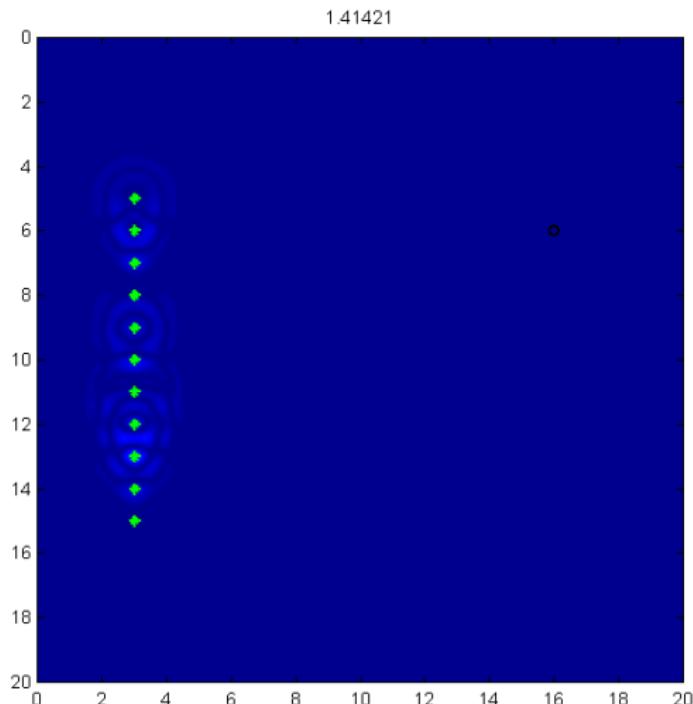


main

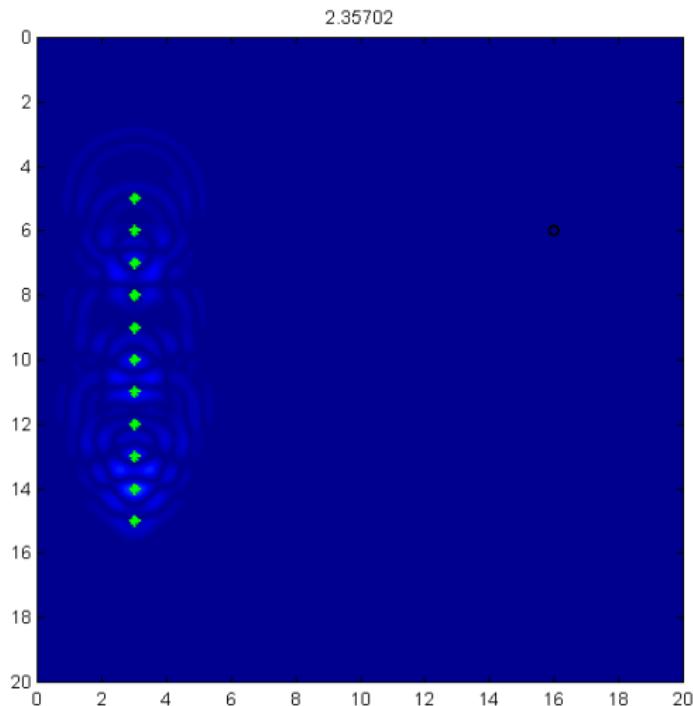
Source localization - Backward step



Source localization - Backward step

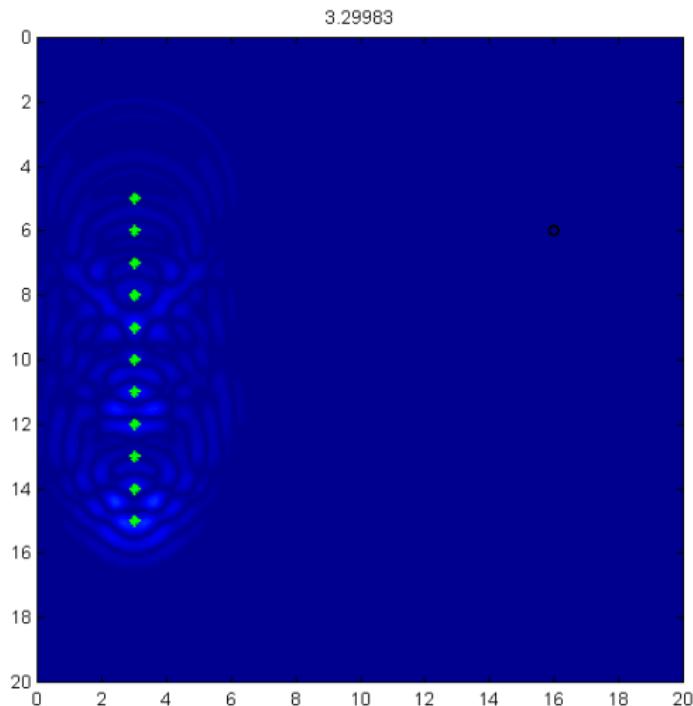


Source localization - Backward step

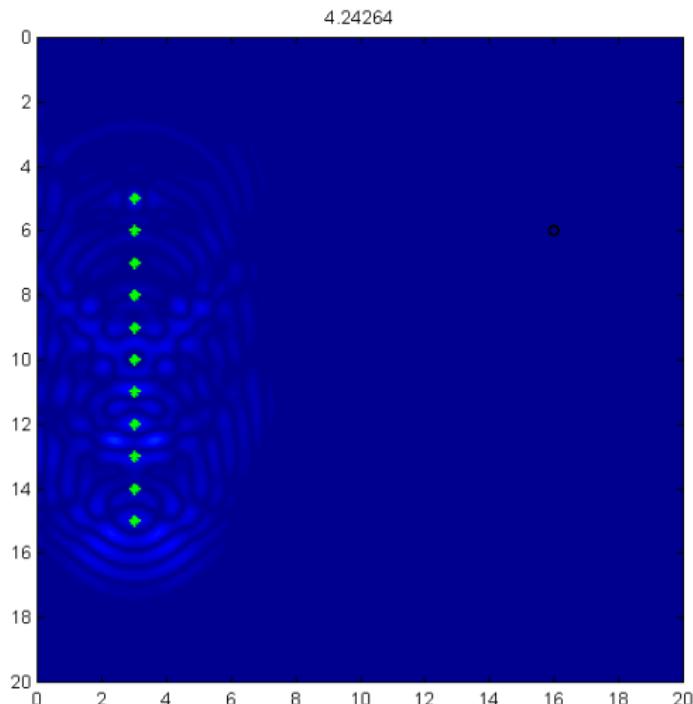


main

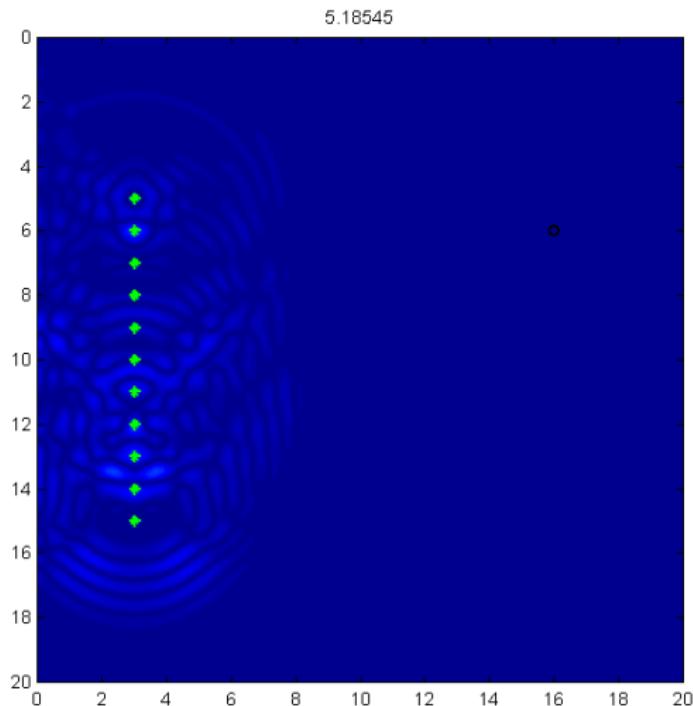
Source localization - Backward step



Source localization - Backward step

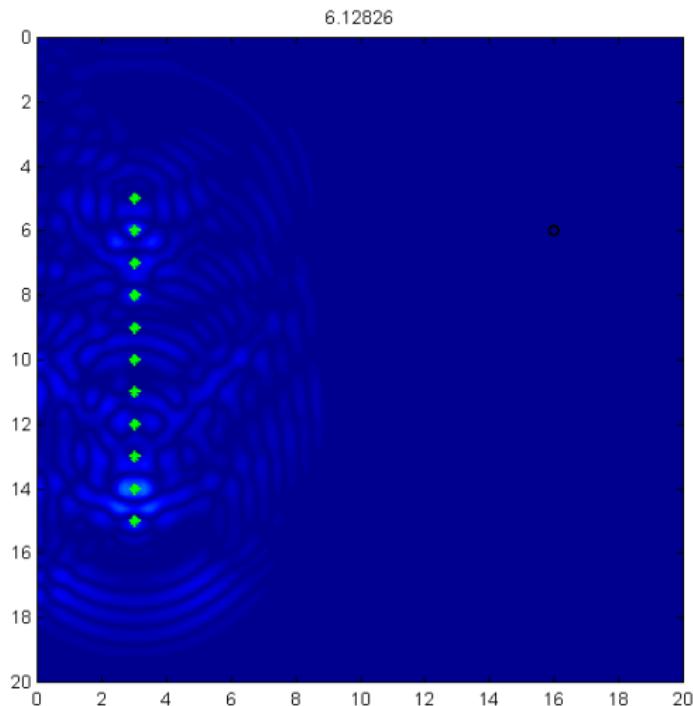


Source localization - Backward step



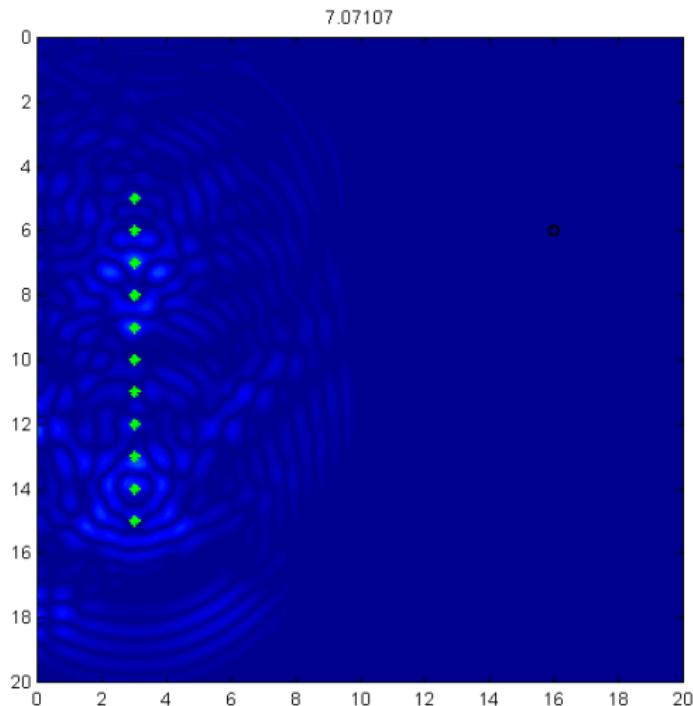
main

Source localization - Backward step



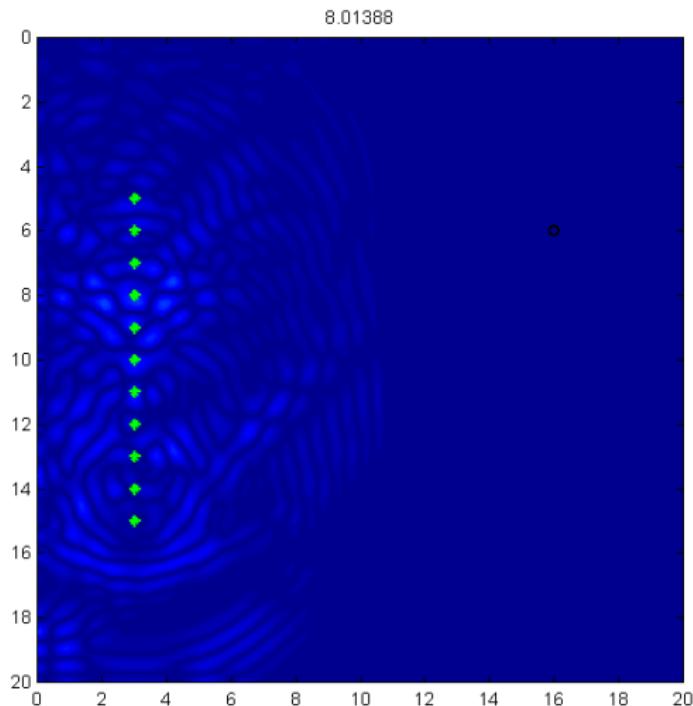
main

Source localization - Backward step

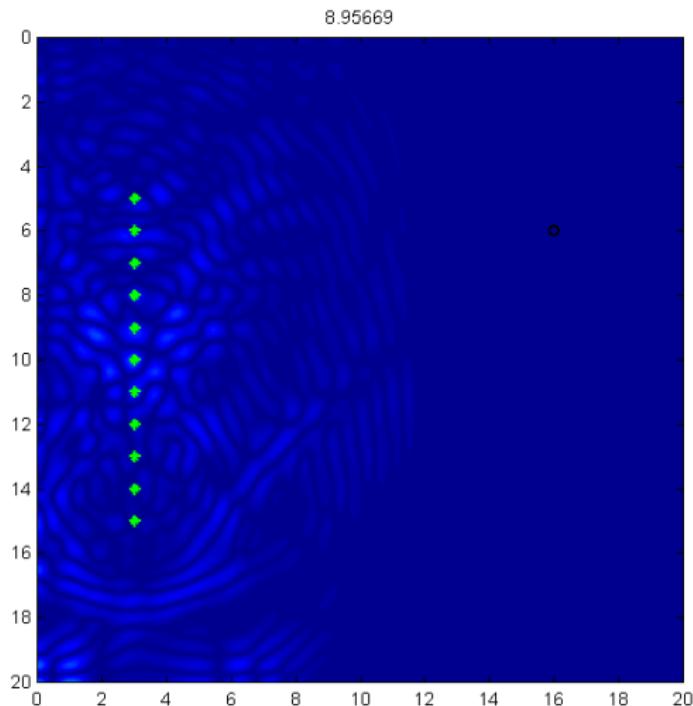


main

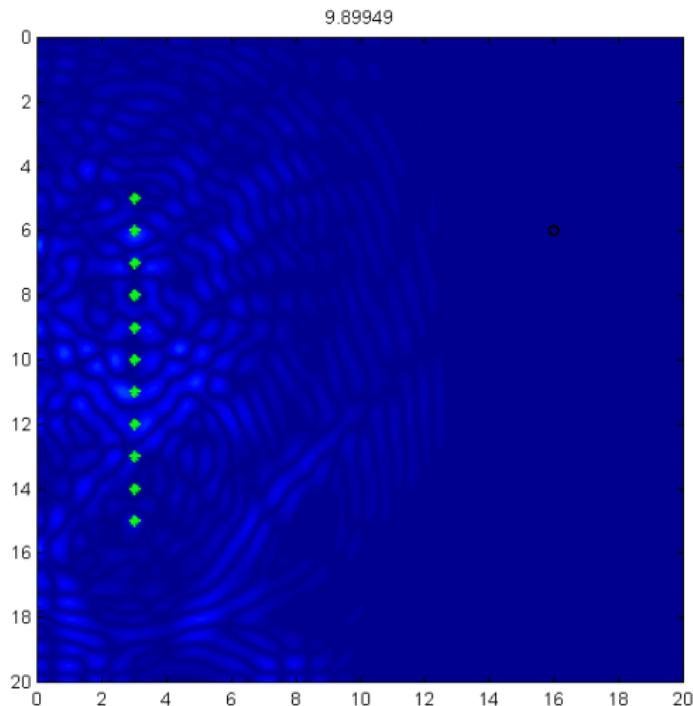
Source localization - Backward step



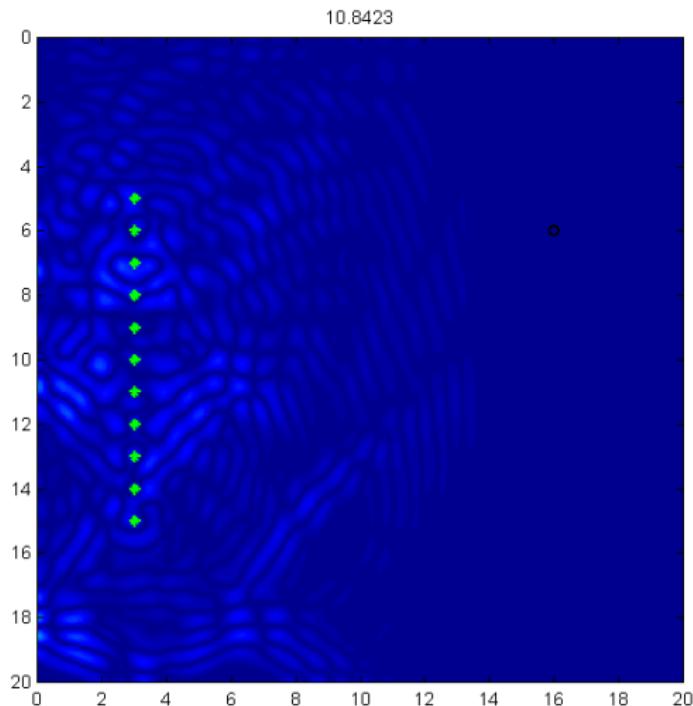
Source localization - Backward step



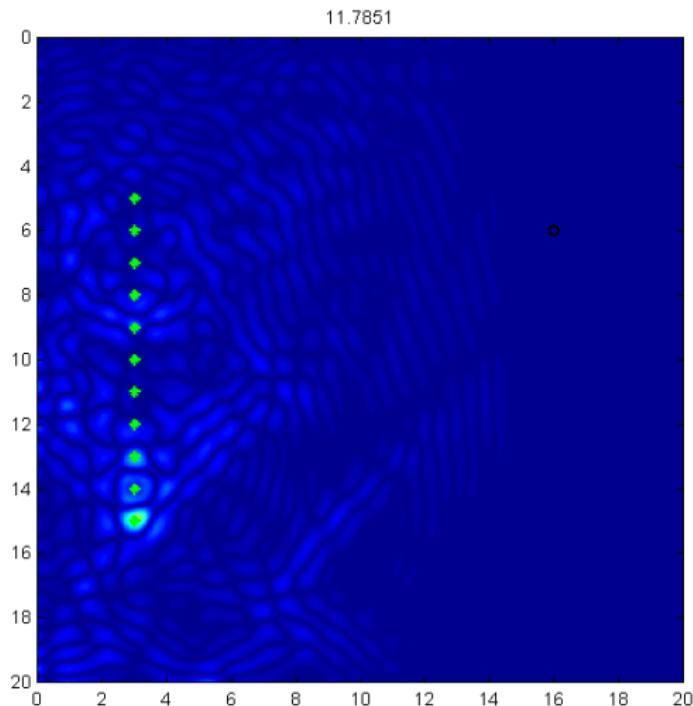
Source localization - Backward step



Source localization - Backward step

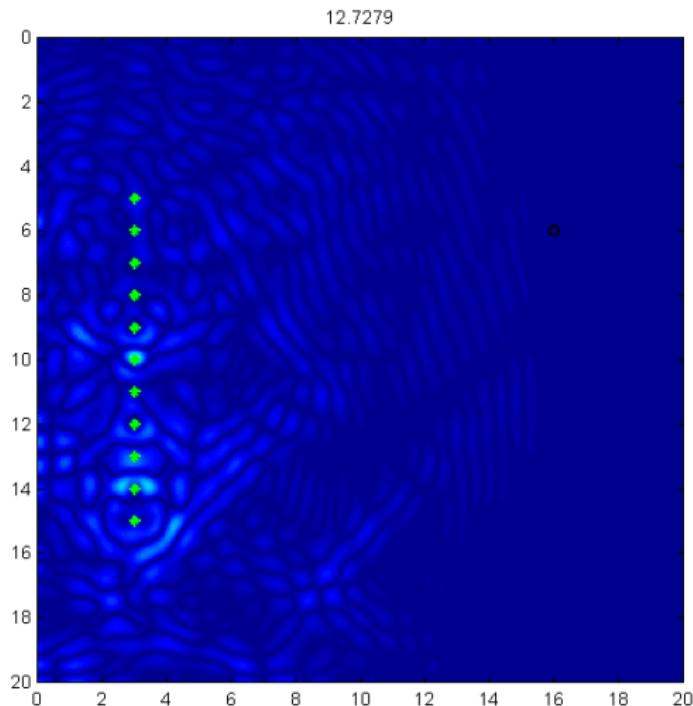


Source localization - Backward step

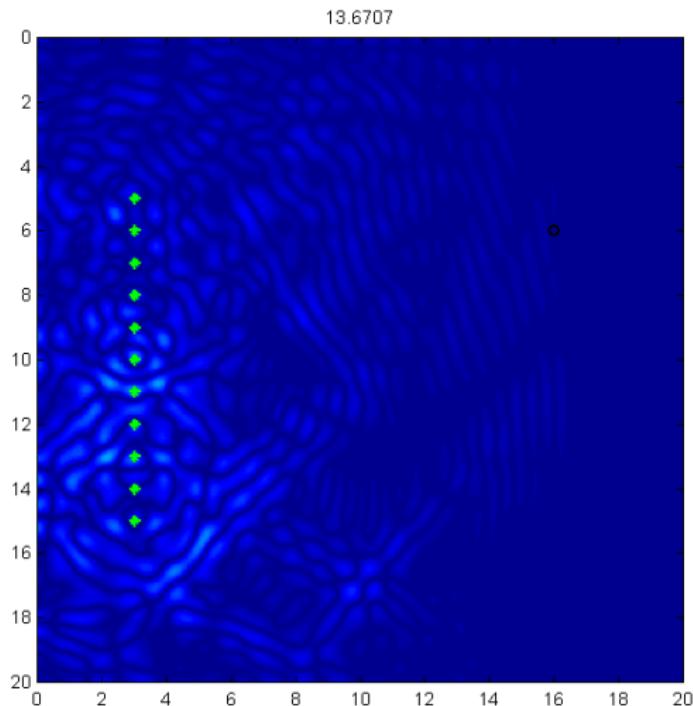


main

Source localization - Backward step

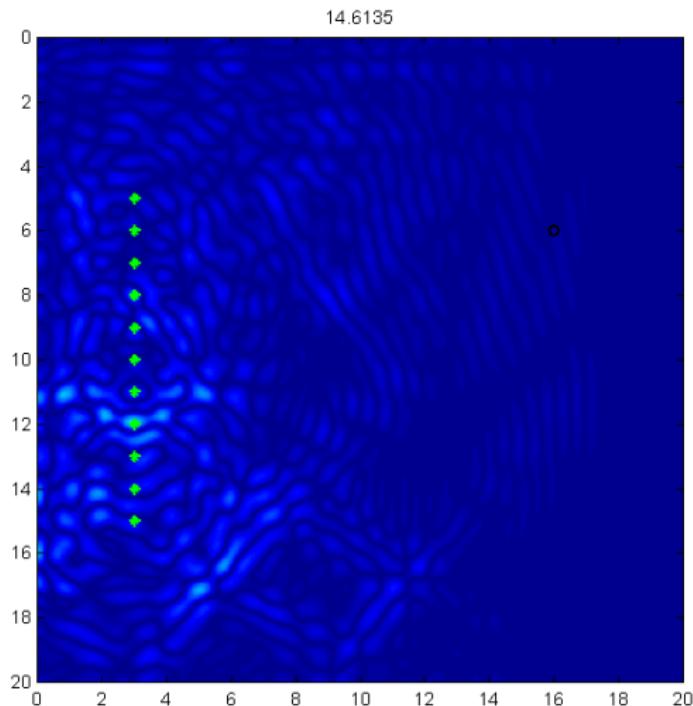


Source localization - Backward step

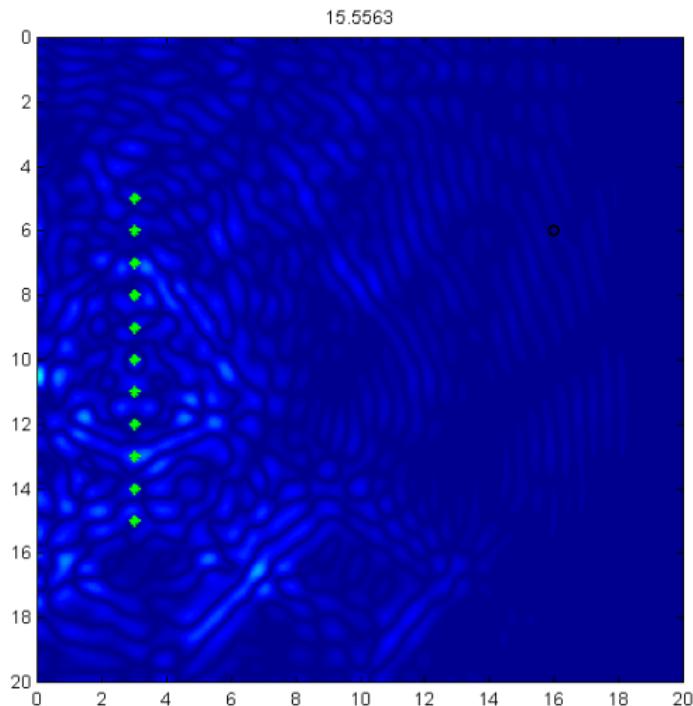


main

Source localization - Backward step

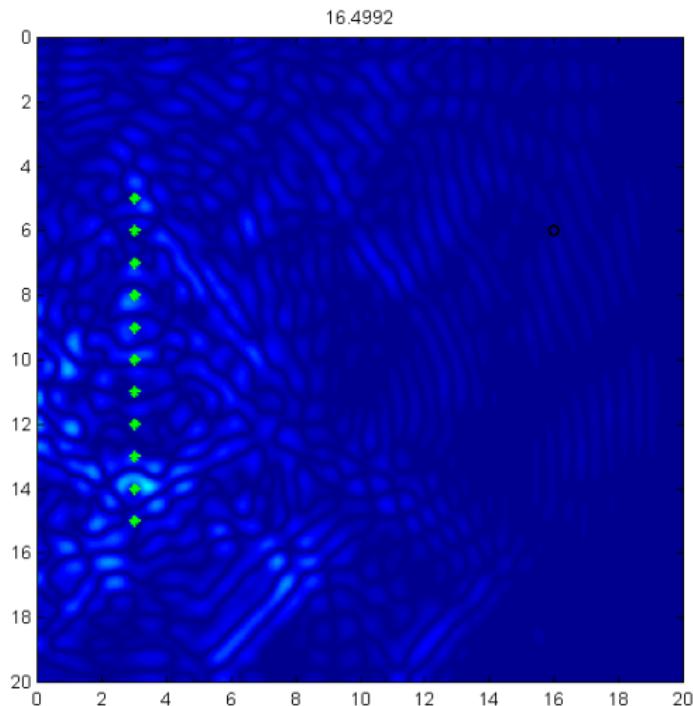


Source localization - Backward step



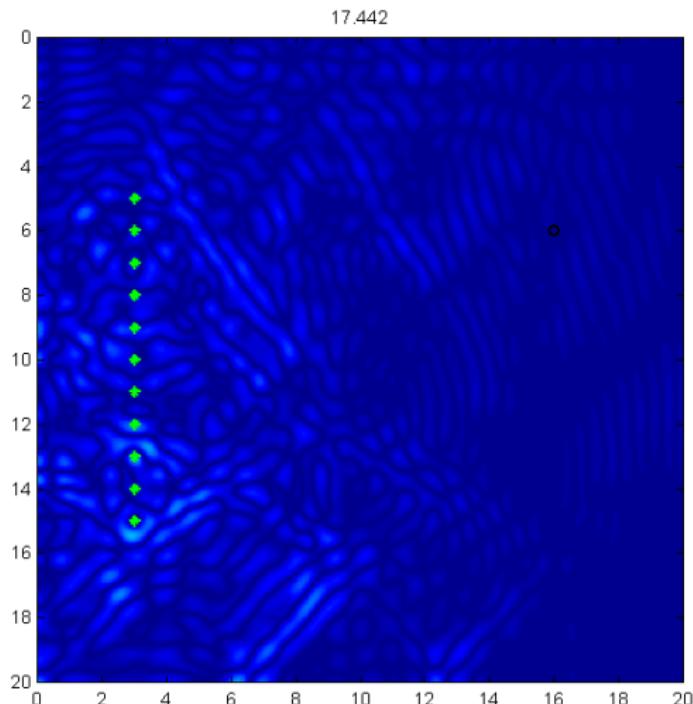
main

Source localization - Backward step



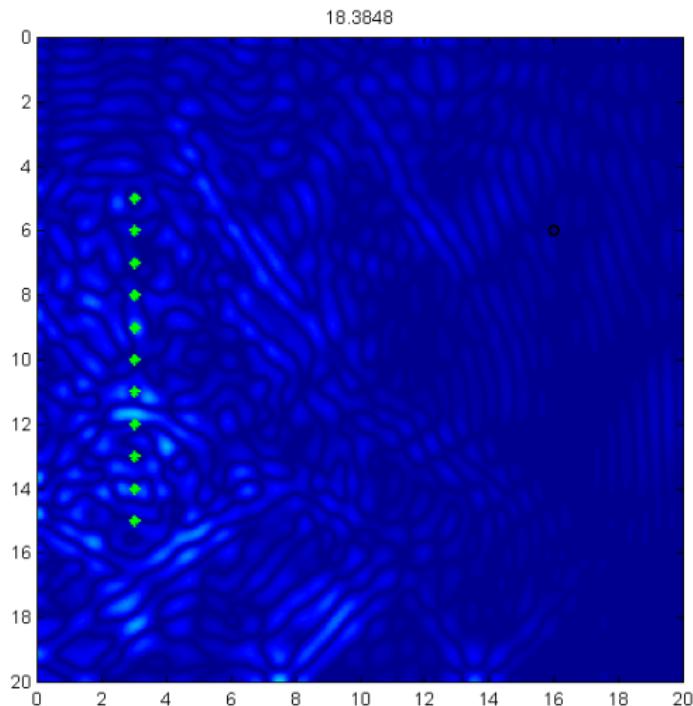
main

Source localization - Backward step



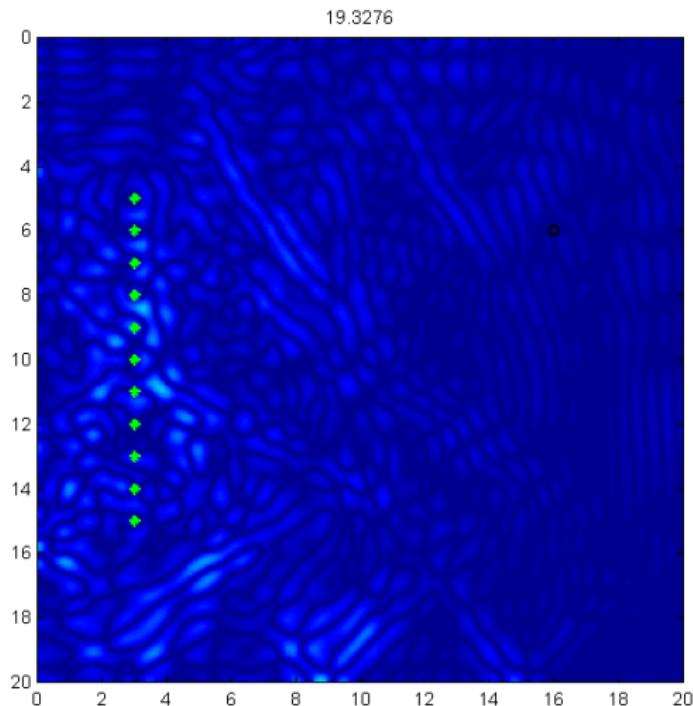
main

Source localization - Backward step



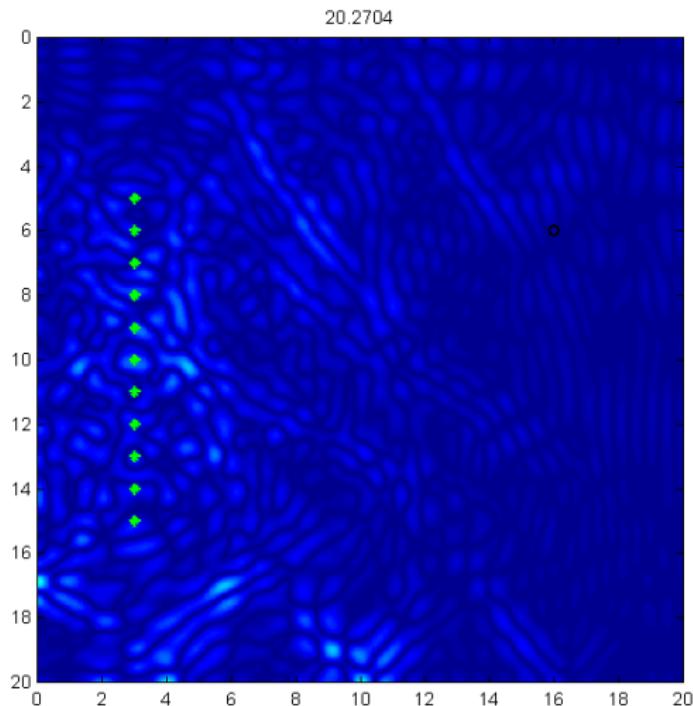
main

Source localization - Backward step

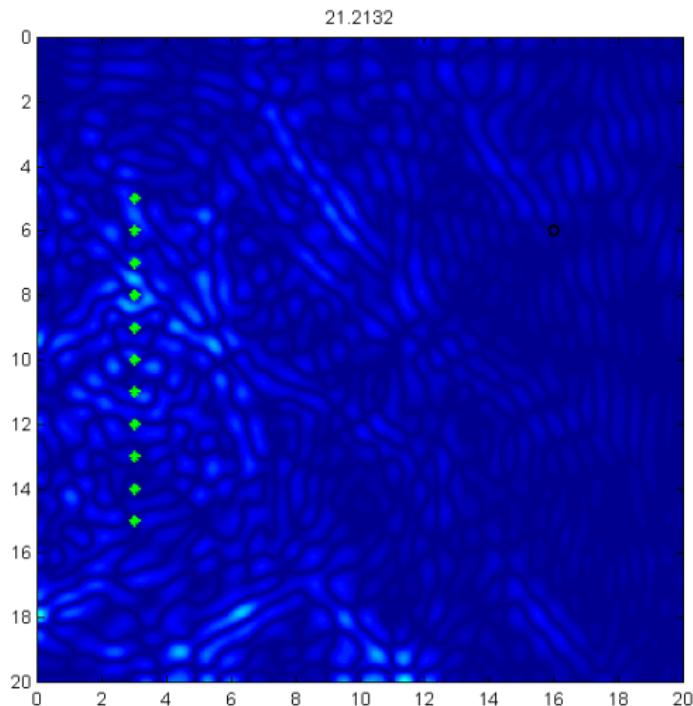


main

Source localization - Backward step

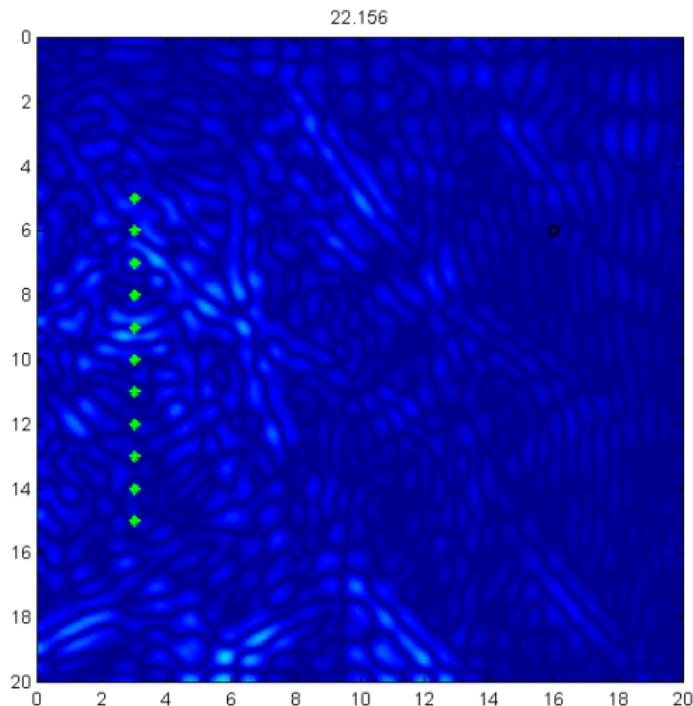


Source localization - Backward step

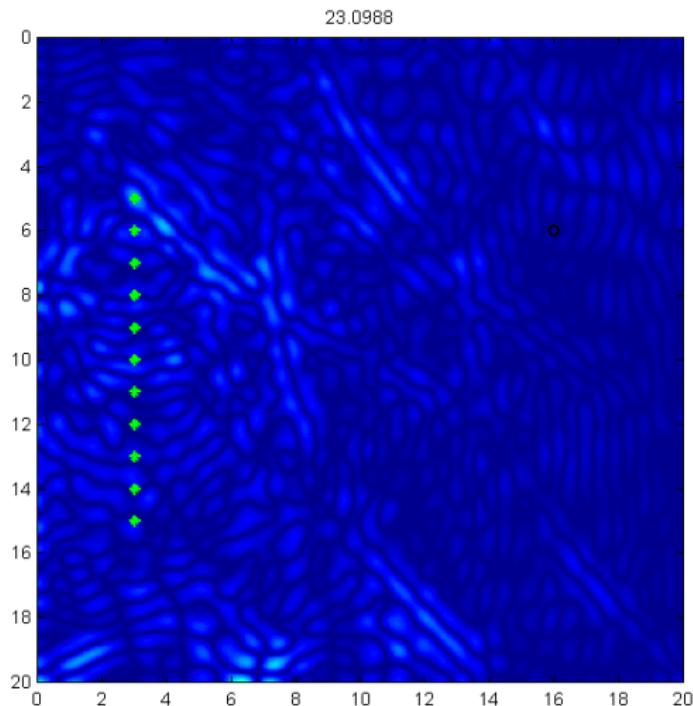


main

Source localization - Backward step

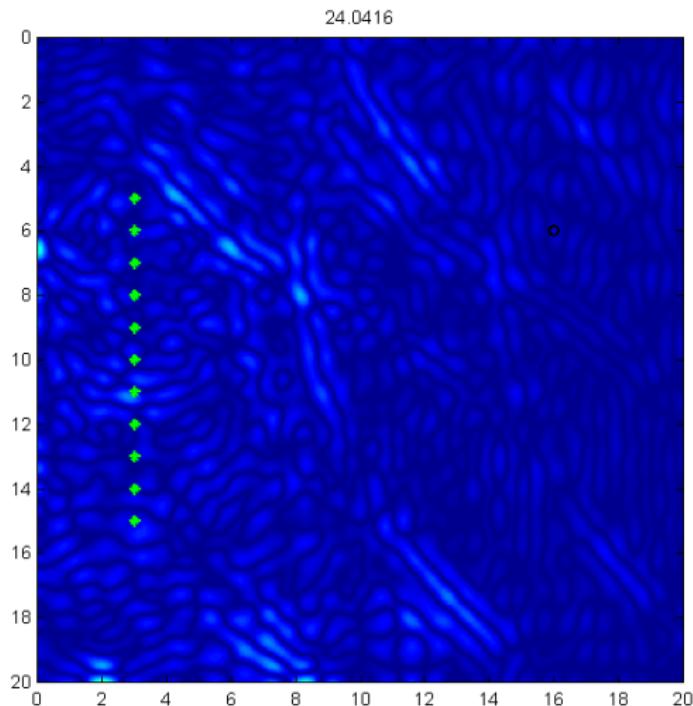


Source localization - Backward step



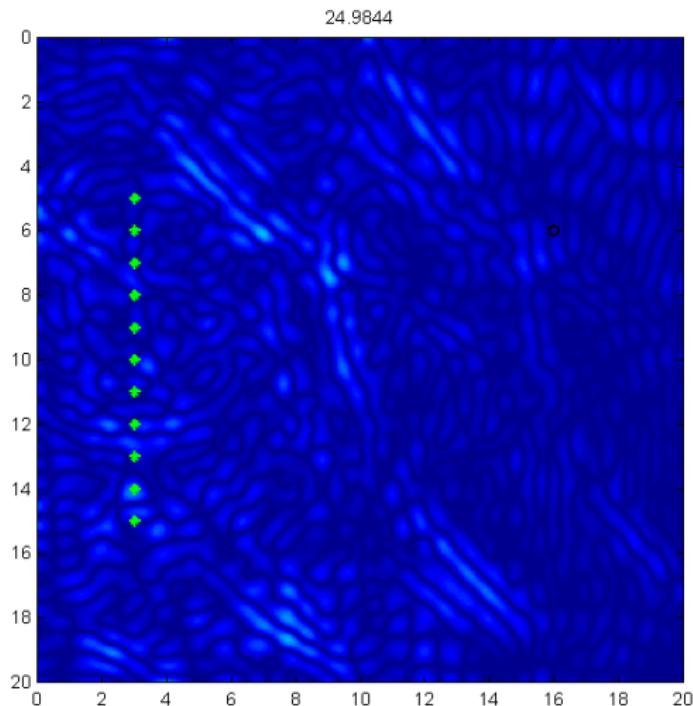
main

Source localization - Backward step



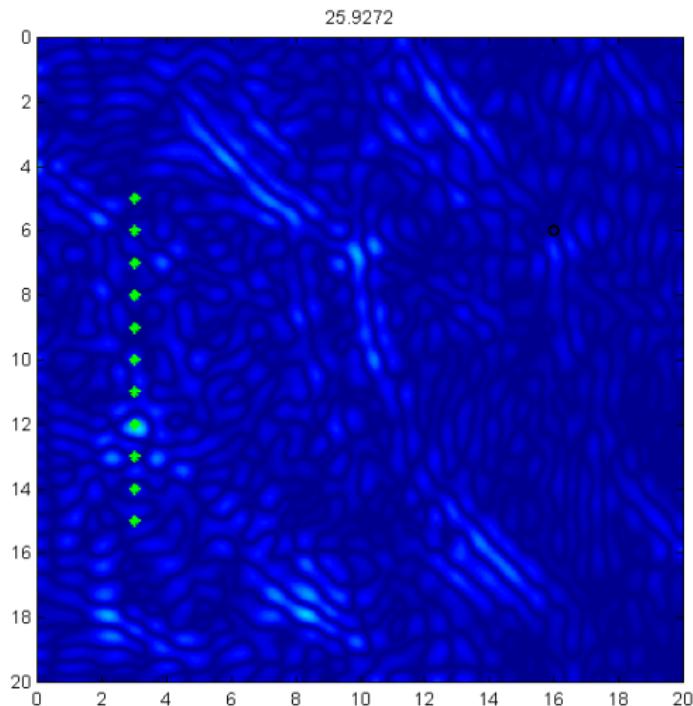
main

Source localization - Backward step



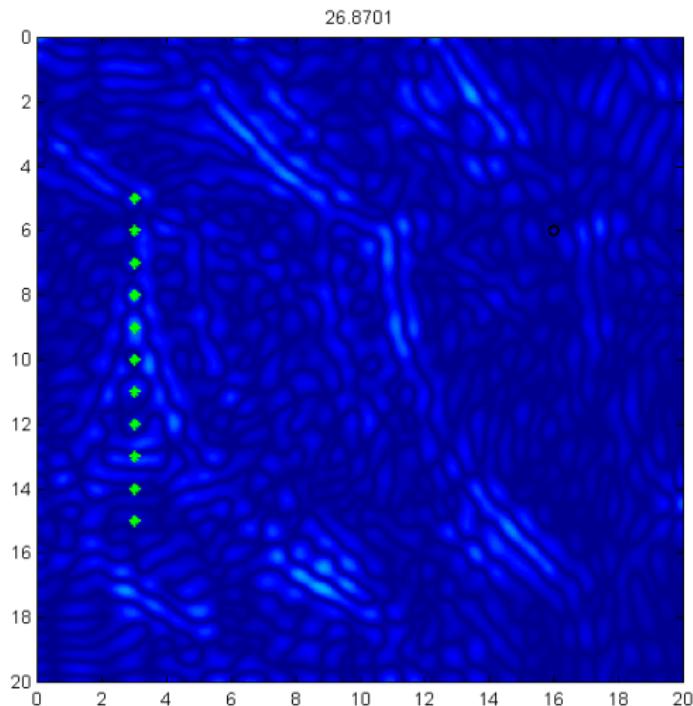
main

Source localization - Backward step

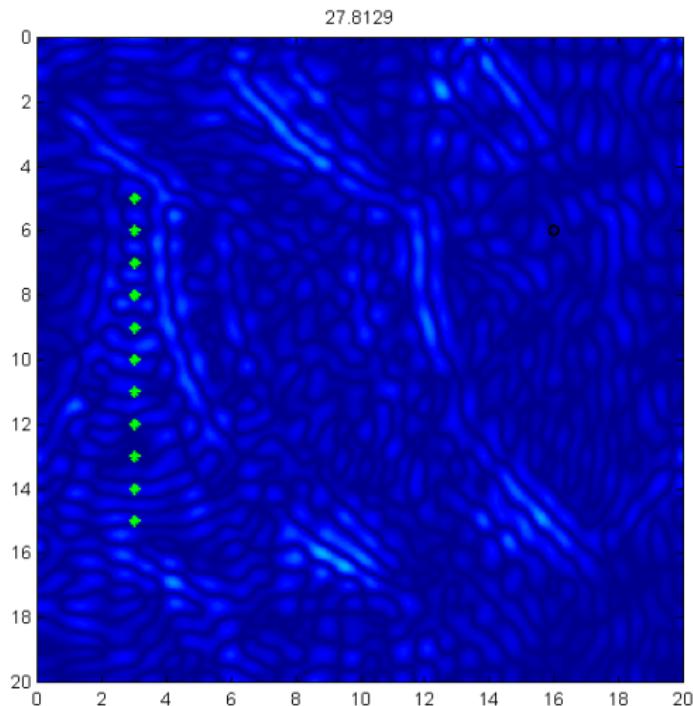


main

Source localization - Backward step

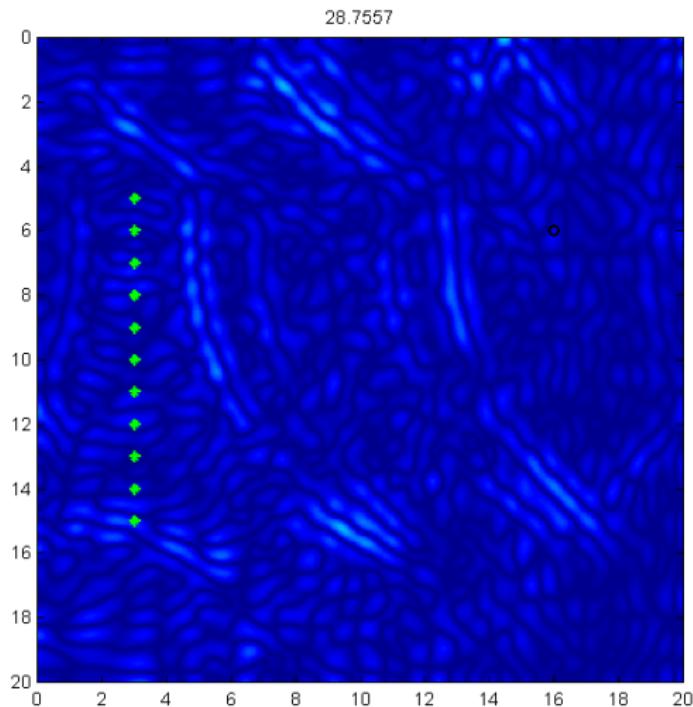


Source localization - Backward step



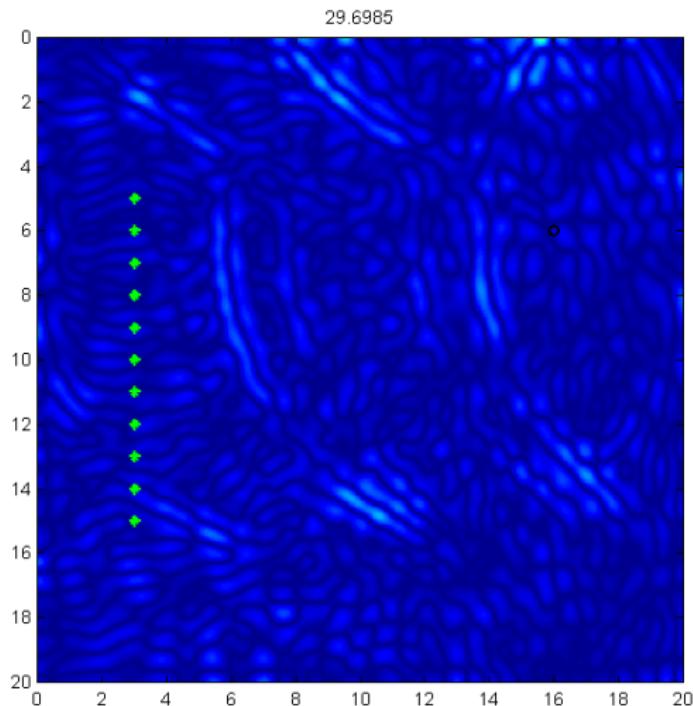
main

Source localization - Backward step



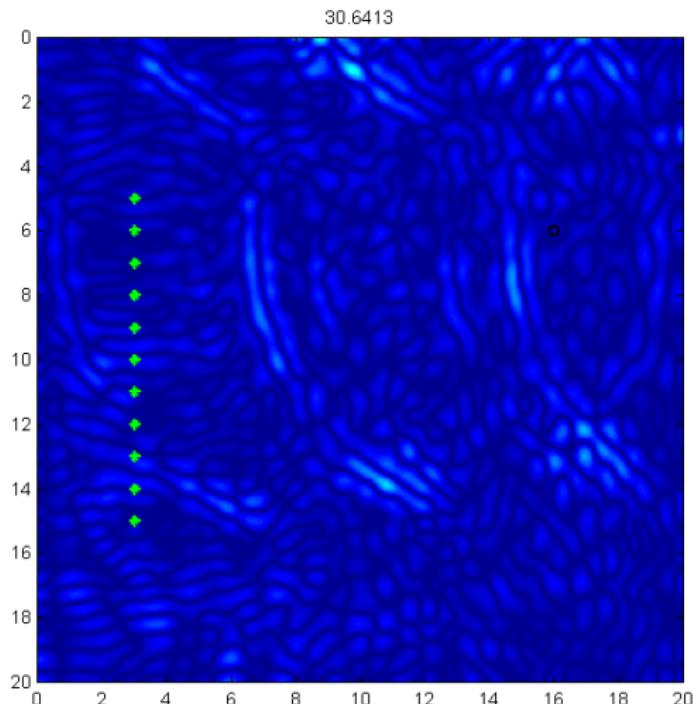
main

Source localization - Backward step



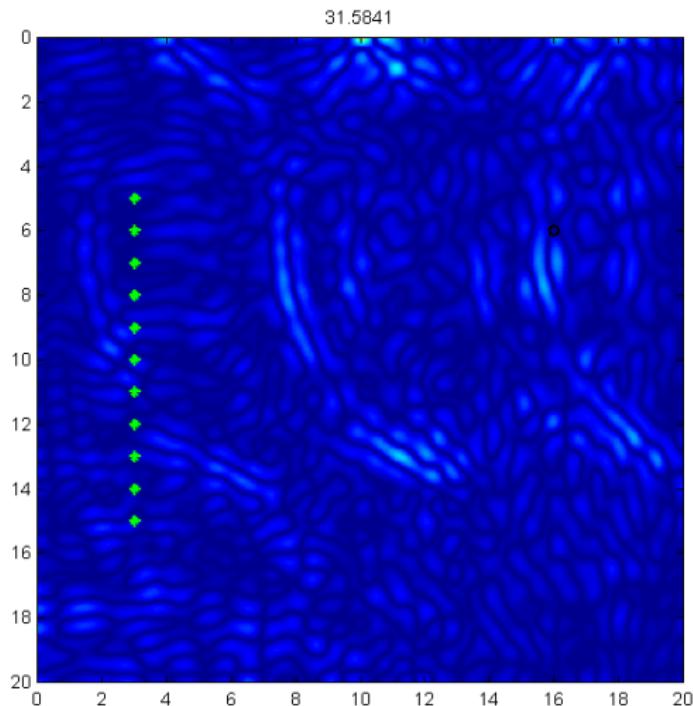
main

Source localization - Backward step



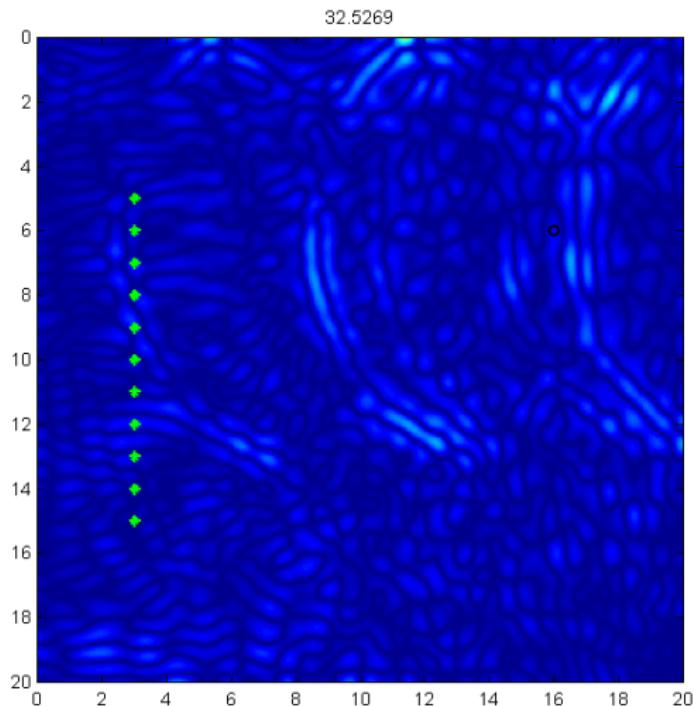
main

Source localization - Backward step

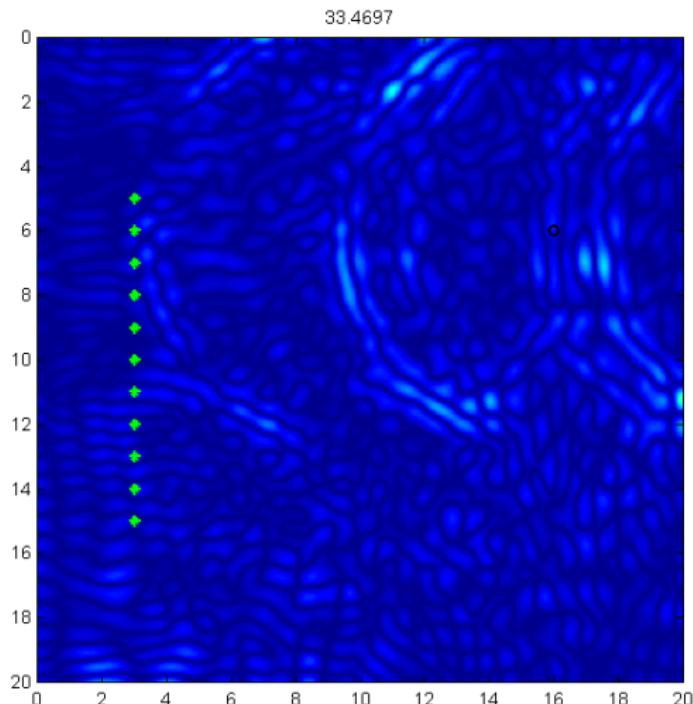


main

Source localization - Backward step

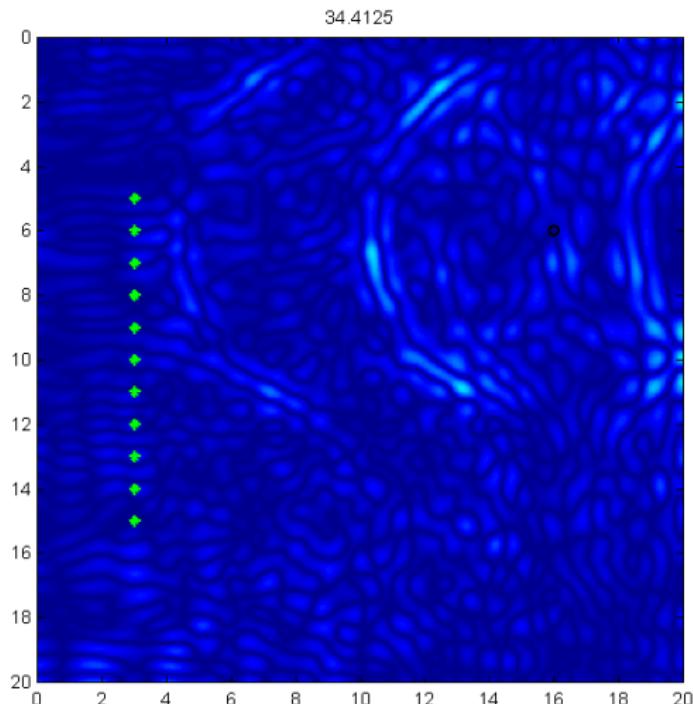


Source localization - Backward step



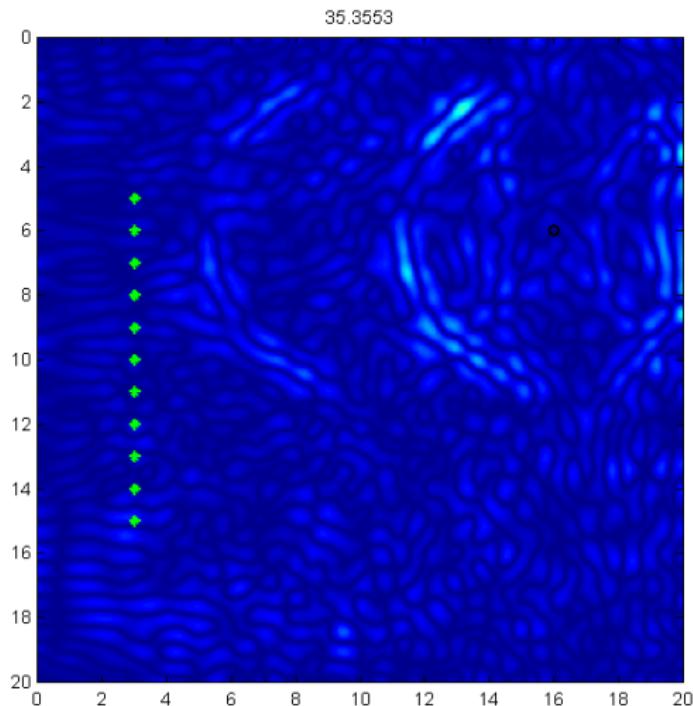
main

Source localization - Backward step



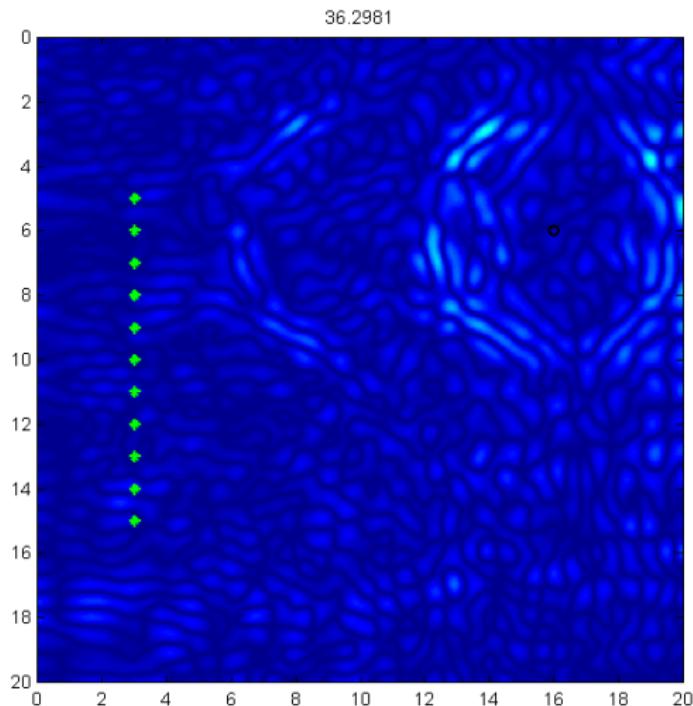
main

Source localization - Backward step



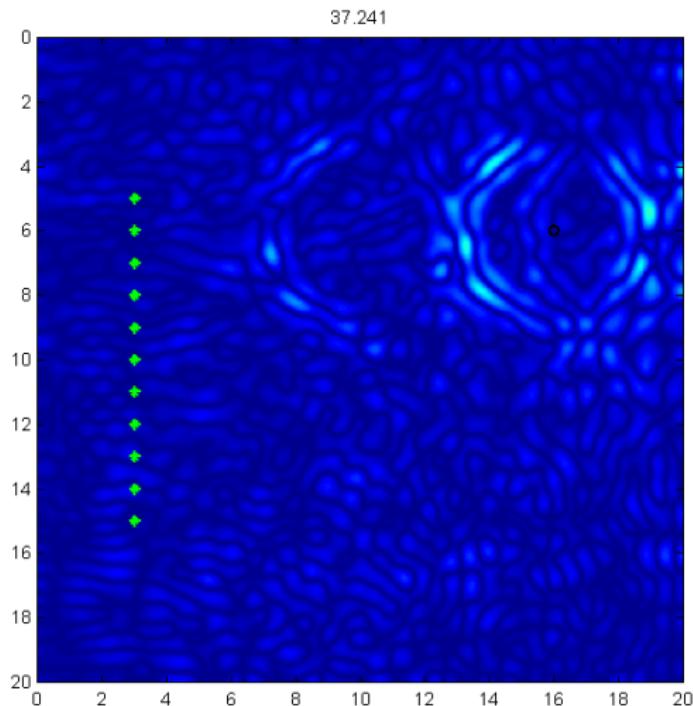
main

Source localization - Backward step



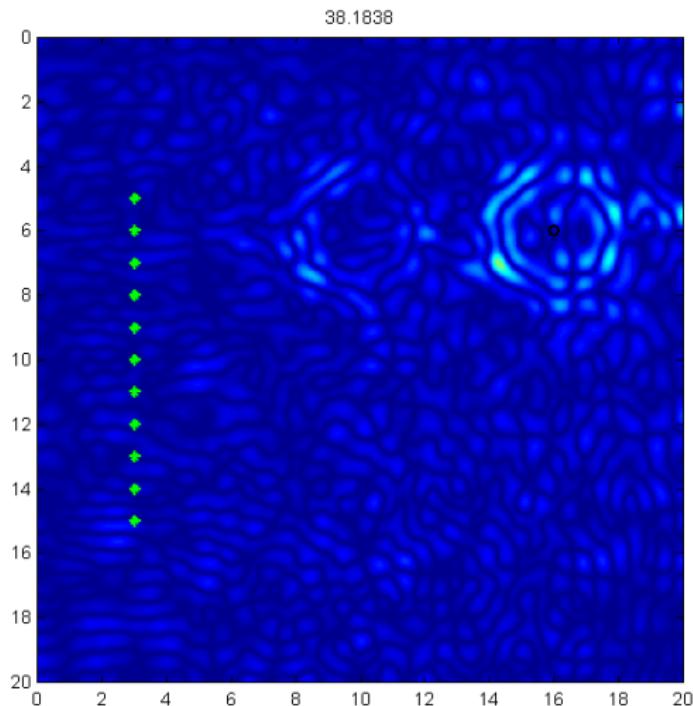
main

Source localization - Backward step



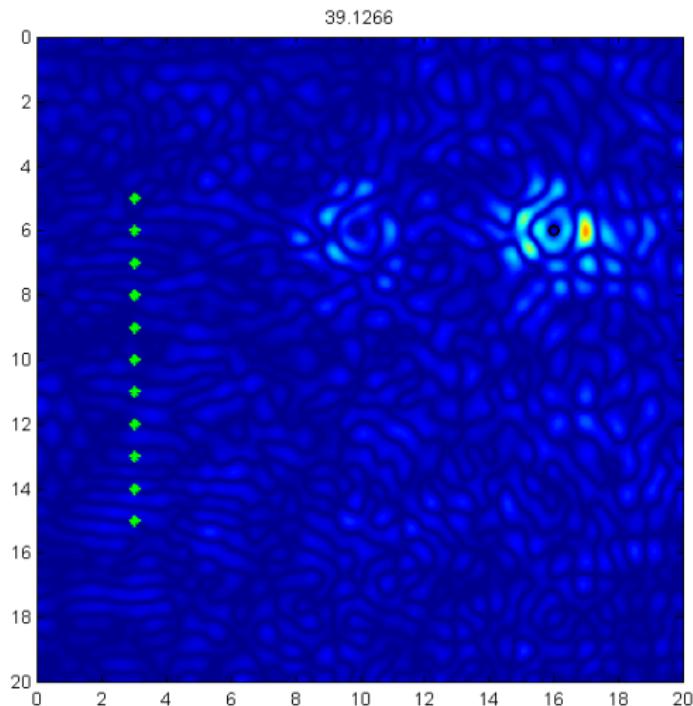
main

Source localization - Backward step



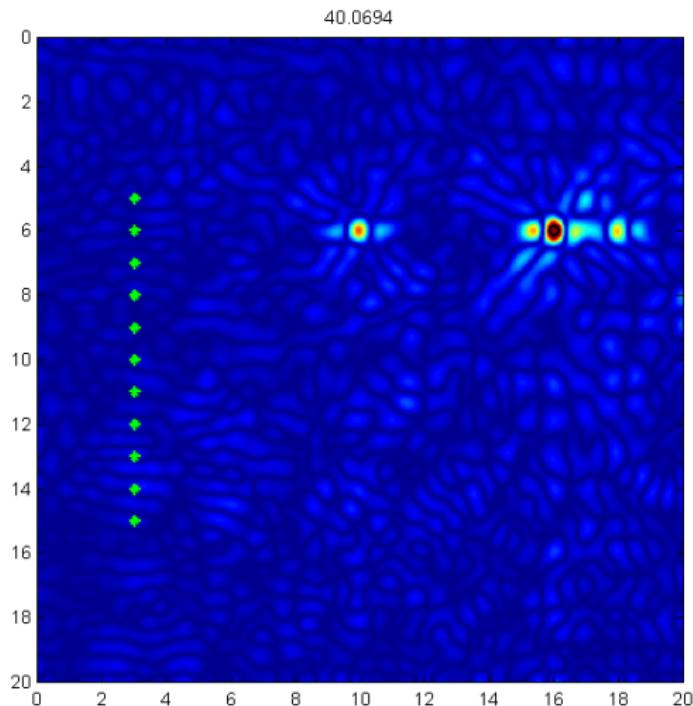
main

Source localization - Backward step



main

Source localization - Backward step



main