

# Selection of defective components in unknown backgrounds

**Housseem HADDAR**

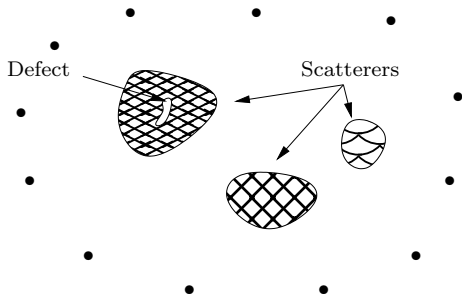
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Joint work with **Lorenzo AUDIBERT** and **Alexandre GIRARD**

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# Model problem



**Goal:** Determine **defects** or **defective components** in a complex and **unknown medium** from multi-static measurements of scattered waves at a given frequency.

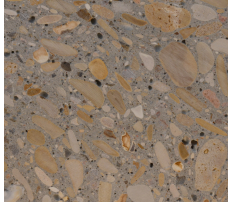
## **Constraints:**

- ▶ The background is unknown and cannot be accurately reconstructed.
- ▶ The background components diameters are comparable to the wavelength.

**But:** We have access to **differential measures**: measures with and without defects.

# The original motivation

Detection of defects in a concrete like material using ultrasounds

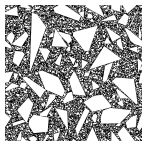


An example of a concrete structure

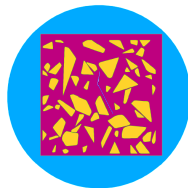
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**A simulation using the Linear Sampling Method** (without differential measurements)



An example of synthetic background

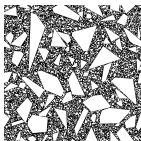


A filtered background + crack

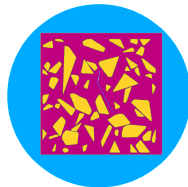
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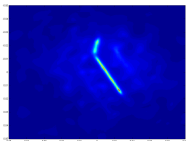
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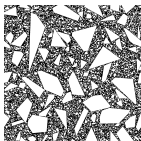


Reconstructed crack using  
the exact background

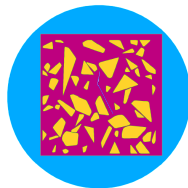
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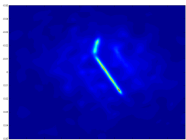
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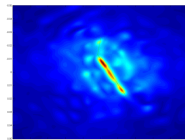
An example of synthetic background



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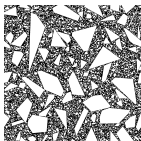


Reconstructed crack using  
a homogeneous background: weak perturbation

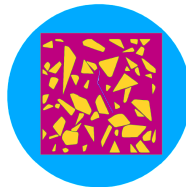
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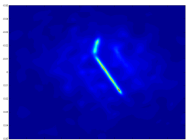
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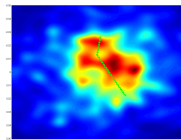
An example of synthetic background



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Reconstructed crack using  
the exact background



Reconstructed crack using  
a homogeneous background: strong perturbation

# Outline

- ▶ A model problem (not for cracks)
- ▶ The Linear Sampling Method revisited
- ▶ Application to the case of differential measurements
- ▶ Numerical results and perspectives



# A simple model problem

## Scalar acoustic equation for inhomogeneous media

The background index  $n_0$  :  $n_0 = 1$  in  $\mathbb{R}^d \setminus D_0$  and  $\mathbb{R}^d \setminus D_0$  is connected.

The modified index  $n$  :  $n = 1$  in  $\mathbb{R}^d \setminus D$  and  $\mathbb{R}^d \setminus D$  is connected.

The total fields  $u_0 \in H_{loc}^1(\mathbb{R}^d)$  and  $u \in H_{loc}^1(\mathbb{R}^d)$

$$\Delta u_0 + k^2 n_0 u_0 = 0 \text{ and } \Delta u + k^2 n u = 0 \text{ in } \mathbb{R}^d$$

We assume that the field is generated by incident plane waves:

$$u^i(\theta, x) := e^{ikx \cdot \theta} \quad \theta \in \mathbb{S}^{d-1}$$

The scattered fields

$$u_0^s(\theta, \cdot) = u_0 - u^i(\theta, \cdot) \text{ and } u^s(\theta, \cdot) = u - u^i(\theta, \cdot) \quad \text{in } \mathbb{R}^d,$$

satisfy the Sommerfeld radiation condition.

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Our data is formed by (noisy measurements of) so-called farfield patterns

$$u_0^\infty(\theta, \hat{x}) \text{ and } u^\infty(\theta, \hat{x}) \text{ for all } (\theta, \hat{x}) \in \mathbb{S}^{d-1} \times \mathbb{S}^{d-1}$$

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Recall that with  $\hat{x} := x/|x|$ ,

$$u_0^s(\theta, x) = \frac{e^{ik|x|}}{|x|^{(d-1)/2}} (u_0^\infty(\theta, \hat{x}) + O(1/|x|))$$

$$u^s(\theta, x) = \frac{e^{ik|x|}}{|x|^{(d-1)/2}} (u^\infty(\theta, \hat{x}) + O(1/|x|))$$

as  $|x| \rightarrow \infty$ .

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**Goal:** Assuming that  $D_0 \subset D$  and would like to reconstruct (an approximation of)

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without knowing (or approximating)  $n$  and  $n_0$ .

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**Algorithm:** introduce a **filtered difference** between the indicator functions provided by **a modified version of** the Linear Sampling Method (LSM) applied to each set of data separately.

# A generalized version of LSM

- ▶ A version based on a new **exact characterization** of the scatterer geometry in terms of the farfields.
- ▶ A version capable of answering the imaging problem for differential measure: explicit link with solutions of the **interior transmission problem**.
- ▶ A **flexible setting** that can be generalized to limited aperture or/and near-field data (ongoing).

**Related reference:** L. Audibert - H. Haddar, Inverse Problems, 2014

# Outline of LSM

**Farfield Operator:**  $F : L^2(\mathbb{S}^{d-1}) \rightarrow L^2(\mathbb{S}^{d-1})$ , defined by

$$Fg(\hat{x}) := \int_{\mathbb{S}^{d-1}} u^\infty(\theta, \hat{x}) g(\theta) ds(\theta).$$

Let us define for  $\psi \in L^2(D)$ , the unique function  $w \in H_{\text{loc}}^1(\mathbb{R}^d)$  satisfying

$$\begin{cases} \Delta w + nk^2 w = k^2(1-n)\psi \text{ in } \mathbb{R}^d, \\ \lim_{r \rightarrow \infty} \int_{|x|=r} \left| \frac{\partial w}{\partial r} - ikw \right|^2 ds = 0. \end{cases} \quad (1)$$

**Remark**

$$\psi = u^i(\theta, \cdot) \Rightarrow w = u^s(\theta, \cdot) \Rightarrow w^\infty = u^\infty(\theta, \cdot)$$

$\Rightarrow Fg$  is nothing but  $w^\infty$  for  $w$  solution of (1) with  $\psi = v_g$  in  $D$ , where

$$v_g(x) := \int_{\mathbb{S}^{d-1}} u^i(\theta, x) g(\theta) ds(\theta), \quad g \in L^2(\mathbb{S}^{d-1}), \quad x \in \mathbb{R}^d.$$

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$\Rightarrow$  Considering the (compact) operator  $H : L^2(\mathbb{S}^{d-1}) \rightarrow L^2(D)$  defined by

$$Hg := v_g|_D, \tag{2}$$

and the (compact) operator  $G : \overline{\mathcal{R}(H)} \subset L^2(D) \rightarrow L^2(\mathbb{S}^{d-1})$  defined by

$$G\psi := w^\infty,$$

then clearly:

$$\boxed{F = G \circ H}$$



# Main ingredient of LSM

**Theorem:** Assume that ITP is well posed. With  $\phi_z^\infty(\hat{x}) = e^{-ik\hat{x}\cdot z}$  we have:  $\phi_z^\infty \in \mathcal{R}(G)$  if and only if  $z \in D$ .

Main ingredients of the proof:

- ▶  $\overline{\mathcal{R}(H)} = \{v \in L^2(D); \Delta v + k^2 v = 0 \text{ in } D\}$ .
- ▶  $\phi_z^\infty$  is the farfield of  $\Phi(\cdot, z)$ , radiating solution of  $\Delta \Phi + k^2 \Phi = -\delta_z$ .

**Interior Transmission Problem (ITP):**  $(u, v) \in L^2(D) \times L^2(D)$  such that  $u - v \in H^2(D)$  and

$$\begin{cases} \Delta u + k^2 n u = 0 & \text{in } D, \\ \Delta v + k^2 v = 0 & \text{in } D, \\ (u - v) = f & \text{on } \partial D, \\ \frac{\partial}{\partial \nu}(u - v) = g & \text{on } \partial D, \end{cases} \quad (3)$$

for given  $f \in H^{\frac{3}{2}}(\partial D)$  and  $g \in H^{\frac{1}{2}}(\partial D)$ .

**Remark:** A well posed ITP requires  $n \neq 1$  in any neighborhood of  $\partial D$ .

# Main theorem of LSM

**Farfield Operator:**  $F : L^2(\mathbb{S}^{d-1}) \rightarrow L^2(\mathbb{S}^{d-1})$ , defined by

$$Fg(\hat{x}) := \int_{\mathbb{S}^{d-1}} u^\infty(\theta, \hat{x}) g(\theta) ds(\theta).$$

**Theorem:** Assume that ITP is well posed. Then the operator  $F$  is injective with dense range. Moreover, the following holds.

- ▶ If  $z \in D$  then there exists  $g_z^\epsilon$  such that  $\|Fg_z^\epsilon - \phi_z\|_{L^2(\mathbb{S}^{d-1})} \leq \epsilon$  and  $\limsup_{\epsilon \rightarrow 0} \|Hg_z^\epsilon\|_{L^2(D)} < \infty$ .
- ▶ If  $z \notin D$  then for all  $g_z^\epsilon$  such that  $\|Fg_z^\epsilon - \phi_z\|_{L^2(\mathbb{S}^{d-1})} \leq \epsilon$ ,  $\lim_{\epsilon \rightarrow 0} \|Hg_z^\epsilon\|_{L^2(D)} = \infty$ .

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$\Rightarrow$  Gives a “characterization” of  $D$  in terms of a nearby solutions of

$$Fg_z^\epsilon \simeq \phi_z.$$

**Problems:** This is not constructive...

- ▶ We do not know how to construct  $g_z^\epsilon$ . In practice we use a regularization scheme.
- ▶ We cannot compute  $\|Hg_z^\epsilon\|_{L^2(D)}$ . In practice we use  $\|g_z^\epsilon\|_{L^2(\mathbb{S}^{d-1})}$ .

## A robust formulation of LSM

**Idea:** Reconstruct a nearby solution of the LSM by using a least squares misfit functional with a penalty term that controls  $\|Hg_z^\epsilon\|_{L^2(D)}^2$ .

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We exploit the (second) Factorization:

$$w^\infty(\hat{x}) = - \int_D e^{-iky \cdot \hat{x}} (1-n) k^2 (\psi(y) + w(y)) dy,$$

$\Rightarrow G = H^* T \psi$  where  $H^* : L^2(D) \rightarrow L^2(\mathbb{S}^{d-1})$  is the adjoint of  $H$  given by

$$H^* \varphi(\hat{x}) := \int_D e^{-iky \cdot \hat{x}} \varphi(y) dy, \quad \varphi \in L^2(D), \quad \hat{x} \in \mathbb{S}^{d-1},$$

and where  $T : L^2(D) \rightarrow L^2(D)$  is defined by

$$T\psi := -k^2(1-n)(\psi + w), \tag{4}$$

$$\boxed{F = H^* \circ T \circ H}$$

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**Theorem:** Assume that (ITP) is well posed and there exists  $n_0 > 0$  and  $\alpha > 0$  such that

$$1 - \Re(n(x)) + \alpha \Im(n(x)) \geq n_0 \text{ for a.e. } x \in D$$

or

$$\Re(n(x)) - 1 + \alpha \Im(n(x)) \geq n_0 \text{ for a.e. } x \in D.$$

Then:  $|(T\psi, \psi)_{L^2(D)}| \geq c \|\psi\|_{L^2(D)}^2$  for all  $\psi \in \mathcal{R}(H)$ .

$$\Rightarrow |(Fg, g)_{L^2(\mathbb{S}^{d-1})}| \geq c \|Hg\|_{L^2(D)}^2$$

$$\Rightarrow |(Fg, g)_{L^2(\mathbb{S}^{d-1})}| \text{ is equivalent to } \|Hg\|_{L^2(D)}^2$$

## Abstract setting for a Generalized LSM (GLSM)

We consider two bounded linear operators  $F : X \rightarrow X^*$  and  $B : X \rightarrow X^*$

$$\boxed{F = GH \quad \text{and} \quad B = H^*TH}$$

$H : X \rightarrow Y$ ,  $T : Y \rightarrow Y^*$  and  $G : \overline{\mathcal{R}(H)} \subset Y \rightarrow X^*$  are bounded.

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For  $\alpha > 0$  be a given parameter and  $\phi \in X^*$  we consider:

$$J_\alpha(\phi; g) := \alpha |\langle Bg, g \rangle| + \|Fg - \phi\|^2 \quad \forall g \in X.$$

**Remark** This functional has not a minimizer in general!

Assume that  $F$  has dense range. Then for all  $\phi \in X^*$ ,

$$j_\alpha(\phi) := \inf_{g \in X} J_\alpha(\phi; g) \rightarrow 0 \text{ as } \alpha \rightarrow 0.$$



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$\Rightarrow$  Nearby solutions (of the farfield equation) are given by  $g_\alpha \in X$  such that

$$J_\alpha(\phi; g_\alpha) \leq j_\alpha(\phi) + p(\alpha).$$

where  $p(\alpha) > 0$  is such that  $p(\alpha) \rightarrow 0$  as  $\alpha \rightarrow 0$

# Main theorem of GLSM (for noise free)

$$F : X \rightarrow X^*, B : X \rightarrow X^* \text{ and } F = GH \quad \text{and} \quad B = H^*TH$$

$$J_\alpha(\phi; g) := \alpha |\langle Bg, g \rangle| + \|Fg - \phi\|^2 \quad \forall g \in X.$$

**Theorem:** We assume in addition that

- ▶  $G$  is compact and  $F = GH$  has dense range.
- ▶  $T$  satisfies:  $|\langle T\varphi, \varphi \rangle| > \mu \|\varphi\|^2 \quad \forall \varphi \in \mathcal{R}(H).$

Consider for  $\alpha > 0$  and  $\phi \in X^*$ ,  $g_\alpha \in X$  such that

$$J_\alpha(\phi; g_\alpha) \leq j_\alpha(\phi) + p(\alpha) \text{ and } p(\alpha) \leq C\alpha.$$

- ▶  $\phi \in \mathcal{R}(G) \Rightarrow \limsup_{\alpha \rightarrow 0} |\langle Bg_\alpha, g_\alpha \rangle| < \infty.$
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**Application:**  $\mathcal{R}(G)$  characterizes the inclusion  $D \Rightarrow F$  and  $B$  uniquely determine  $D$ .

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Similar asymptotic characterizations: Inf-Criterion ([Nachman-Päivärinta-Teirilä](#) (2007), [Kirsch-Grinberg](#) (2008)), Probe Method ([Ikehata](#) (2005), [Erhard-Potthast](#) (2006)).

## Application with the natural choice: $B = F$

For  $z \in \mathbb{R}^d$  we consider  $g_\alpha^z \in L^2(S^{d-1})$  such that

$$J_\alpha(\phi_z^\infty; g_\alpha^z) \leq j_\alpha(\phi_z^\infty) + p(\alpha) \text{ and } p(\alpha) \leq C\alpha.$$

$$\phi_z^\infty(\hat{x}) = e^{-ik\hat{x} \cdot z}$$

**Theorem:** Assume that there exists  $n_* > 0$  and  $\alpha > 0$  such that

$$1 - \Re(n(x)) + \alpha \Im(n(x)) \geq n_* \text{ for a.e. } x \in D \text{ or}$$

$$\Re(n(x)) - 1 + \alpha \Im(n(x)) \geq n_* \text{ for a.e. } x \in D.$$

Then, except for a countable set of  $k$  (without finite accumulation points),

$$\blacktriangleright z \in D \Rightarrow \limsup_{\alpha \rightarrow 0} |\langle Fg_\alpha^z, g_\alpha^z \rangle| < \infty.$$

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**Theorem:** Assume that there exists  $n_* > 0$  and  $\alpha > 0$  such that

$$1 - \Re(n(x)) + \alpha \Im(n(x)) \geq n_* \text{ for a.e. } x \in D \text{ or}$$

$$\Re(n(x)) - 1 + \alpha \Im(n(x)) \geq n_* \text{ for a.e. } x \in D.$$

Then, except for a countable set of  $k$  (without finite accumulation points),

$$\blacktriangleright z \in D \Rightarrow \limsup_{\alpha \rightarrow 0} |\langle Fg_\alpha^z, g_\alpha^z \rangle| < \infty.$$

$$\blacktriangleright z \notin D \Rightarrow \lim_{\alpha \rightarrow 0} |\langle Fg_\alpha^z, g_\alpha^z \rangle| = \infty.$$

$\Rightarrow$  An **indicator** of some approximation of  $D$  is given by

$$z \rightarrow 1/|\langle Fg_\alpha^z, g_\alpha^z \rangle|.$$

## Application with the natural choice: $B = F$

$\Rightarrow$  An **indicator** of some approximation of  $D$  is given by

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### Remarks

- In this case

$$\lim_{\alpha \rightarrow 0} |\langle Fg_\alpha^z, g_\alpha^z \rangle| = \lim_{\alpha \rightarrow 0} |\langle \phi_z^\infty, g_\alpha^z \rangle| = \lim_{\alpha \rightarrow 0} |v_{g_\alpha^z}(z)|$$

$\Rightarrow$  We obtain a similar indicator function as the one proposed by [Arens \(2004\)](#), [Arens-Lechleiter \(2009\)](#), to justify LSM using the  $(F^*F)^{1/4}$  of [Kirsch \(1997\)](#) in the case  $\Im n = 0$ .

- However this turns out to be a bad indicator function for noisy measurements.

# On other possible choices for $B$

Under more restrictive assumptions on the refractive index

- ▶ If  $\Im n > n_0 > 0$  in  $D$  then we can use

$$B = \Im(F) = \frac{1}{2i}(F - F^*)$$

- ▶ If  $\operatorname{Re}(e^{it}(n-1)) > n_0|n-1| > n_1 > 0$  in  $D$  for some  $t$

$$B = F_{\#} := |e^{it}F + e^{-it}F^*| + \Im(F)$$

(Using the Factorization theorem of [Kirsch-Grinberg](#))

## Remarks

- ▶ For these cases the functional  $J_{\alpha}$  is convex.
- ▶ In these cases we also have ([Kirsch-Grinberg](#))

$$z \in D \text{ iff } \phi_z \in \mathcal{R}(B^{1/2}).$$



# Main theorem of GLSM for noisy operators

$B^\delta$  and  $F^\delta$  compact operators corresponding with noisy measurements

$$\|F^\delta - F\| \leq \delta \|F^\delta\| \quad \text{and} \quad \|B^\delta - B\| \leq \delta \|B^\delta\|$$

for some  $\delta > 0$ .

**Remark:**

$$|\langle Bg, g \rangle| \leq |\langle B^\delta g, g \rangle| + \delta \|B^\delta\| \|g\|^2$$

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**Remark:**

$$|\langle Bg, g \rangle| \leq |\langle B^\delta g, g \rangle| + \delta \|B^\delta\| \|g\|^2$$

$\Rightarrow$  We consider the functional:

$$J_\alpha^\delta(\phi; g) := \alpha(|\langle B^\delta g, g \rangle| + \delta \|B^\delta\| \|g\|^2) + \|F^\delta g - \phi\|^2 \quad \forall g \in X,$$

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**Theorem:** Let  $g_\alpha^\delta$  be the minimizer of  $J_\alpha^\delta(\phi; \cdot)$  for  $\alpha > 0$ ,  $\delta > 0$  and  $\phi \in X^*$ . Then

$$\blacktriangleright \phi \in \mathcal{R}(G) \Rightarrow \limsup_{\alpha \rightarrow 0} \limsup_{\delta \rightarrow 0} \left( |\langle B^\delta g_\alpha^\delta, g_\alpha^\delta \rangle| + \delta \|B^\delta\| \|g_\alpha^\delta\|^2 \right) < \infty$$

$$\blacktriangleright \phi \notin \mathcal{R}(G) \Rightarrow \lim_{\alpha \rightarrow 0} \liminf_{\delta \rightarrow 0} \left( |\langle B^\delta g_\alpha^\delta, g_\alpha^\delta \rangle| + \delta \|B^\delta\| \|g_\alpha^\delta\|^2 \right) = \infty$$

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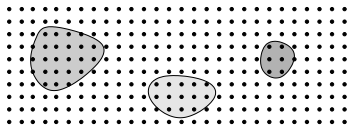
$\Rightarrow$  From the numerical perspective this theorem indicates that a criterion to localize the object would be

$$1 / \left( |\langle B^\delta g_\alpha^\delta, g_\alpha^\delta \rangle| + \delta \|B^\delta\| \|g_\alpha^\delta\|^2 \right)$$

for small values of  $\alpha$ .

## On the numerical implementation

$$J_{\alpha}^{\delta}(\phi; g) := \alpha(|\langle B^{\delta} g, g \rangle| + \delta \|B^{\delta}\| \|g\|^2) + \|F^{\delta} g - \phi\|^2$$



For each  $z$  in the sampling grid, compute

$$g_z = \arg \min J_{\alpha}^{\delta}(\phi_z^{\infty}; g),$$

then plot:

$$z \mapsto 1 / \left( |\langle B^{\delta} g_z, g_z \rangle| + \delta \|B^{\delta}\| \|g_z\|^2 \right)$$

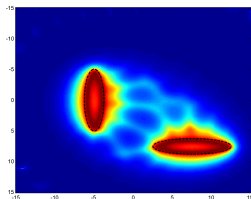
**Initialization:** we use the Tikhonov-Morozov regularized solution

$$(\eta(\delta) + (F^{\delta})^* F^{\delta}) g_z^0 = (F^{\delta})^* \phi_z^{\infty}$$

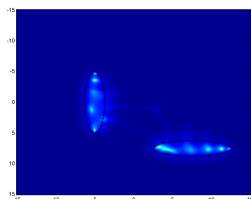
We choose:  $\alpha = \eta(\delta) / (\|F^{\delta}\| + \delta)$ .

# Numerical results

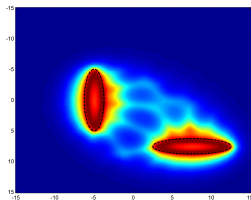
Without optimization, Noise  $\delta = 0\%$



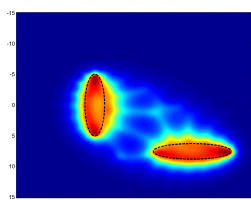
$$1/|\langle B^\delta g_z^0, g_z^0 \rangle|$$



$$1/\|g_z^0\|^2$$



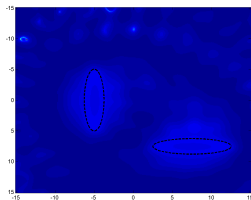
$$1/(|\langle B^\delta g_z^0, g_z^0 \rangle| + \delta \|B^\delta\| \|g_z^0\|^2)$$



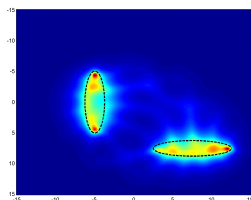
$F_\#$  method

# Numerical results

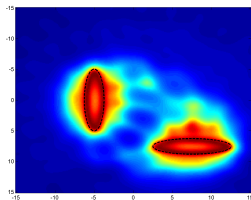
Without optimization, Noise  $\delta = 1\%$



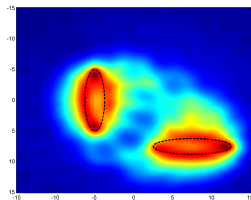
$$1/|\langle B^\delta g_z^0, g_z^0 \rangle|$$



$$1/\|g_z^0\|^2$$



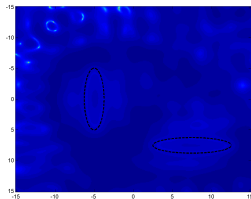
$$1/(|\langle B^\delta g_z^0, g_z^0 \rangle| + \delta \|B^\delta\| \|g_z^0\|^2)$$



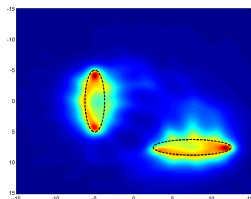
$F_\#$  method

# Numerical results

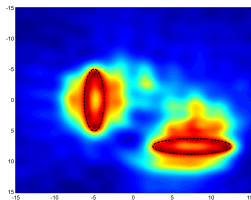
Without optimization, Noise  $\delta = 5\%$



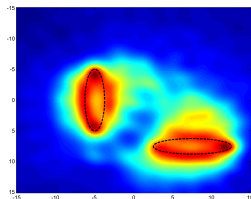
$$1/|\langle B^\delta g_z^0, g_z^0 \rangle|$$



$$1/\|g_z^0\|^2$$



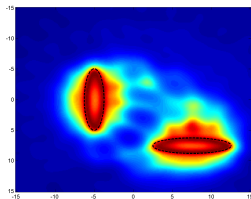
$$1/\left(|\langle B^\delta g_z^0, g_z^0 \rangle| + \delta \|B^\delta\| \|g_z^0\|^2\right)$$



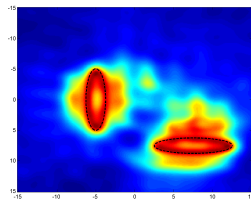
$F_{\#}$  method



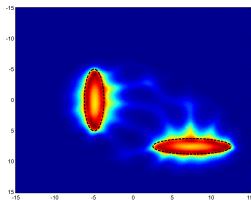
# Numerical results Optim GLSM



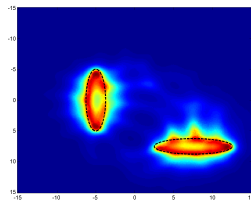
GLSM without optim,  $\delta = 1\%$



GLSM without optim,  $\delta = 5\%$



GLSM with optim,  $\delta = 1\%$



GLSM with optim,  $\delta = 5\%$

# Towards applications to differential measurements

**Main ingredient:** exploit the link between GLSM and ITP.

# Towards applications to differential measurements

**Main ingredient:** exploit the link between GLSM and ITP.

$$F : X \rightarrow X^* \text{ and } B : X \rightarrow X^*$$

$$F = GH \quad \text{and} \quad B = H^*TH$$

$$J_\alpha(\phi; g) := \alpha |\langle Bg, g \rangle| + \|Fg - \phi\|^2 \quad \forall g \in X. \quad (5)$$

**Theorem:** We assume in addition that

$$\varphi \mapsto |\langle T\varphi, \varphi \rangle| \text{ is uniformly convex}$$

$$J_\alpha(\phi; g_\alpha) \leq j_\alpha(\phi) + p(\alpha) \text{ with } \frac{p(\alpha)}{\alpha} \rightarrow 0 \text{ as } \alpha \rightarrow 0. \quad (6)$$

If  $\phi \in \mathcal{R}(G)$  then  $Hg_\alpha \rightarrow \varphi$  such that  $G(\varphi) = \phi$ .

This is a consequence of Tikhonov applied to  $G(\varphi) = \phi$ .

# Towards application to differential measurements

**Main ingredient:** exploit the link between GLSM and ITP.

**Corollary:** with  $F = F$  and  $B = F_{\sharp}$

$$J_{\alpha}(\phi_z^{\infty}; g_z^{\alpha}) \leq j_{\alpha}(\phi_z^{\infty}) + p(\alpha) \text{ with } \frac{p(\alpha)}{\alpha} \rightarrow 0 \text{ as } \alpha \rightarrow 0.$$

If  $z \in D$  then  $Hg_z^{\alpha} \rightarrow v_z$  strongly in  $L^2(D)$  where  $v_z$  is such that there exists  $u_z \in L^2(D)$  for which  $(u_z, v_z)$  is a solution of ITP with  $(f, g) = (\Phi(z, \cdot), \partial_{\nu}\Phi(z, \cdot))$

**Notation** for  $\text{ITP}(D, n, f, g)$ :  $(u, v) \in L^2(D) \times L^2(D)$  such that  $u - v \in H^2(D)$  and

$$\begin{cases} \Delta u + k^2 n u = 0 & \text{in } D, \\ \Delta v + k^2 v = 0 & \text{in } D, \\ (u - v) = f & \text{on } \partial D, \\ \frac{\partial}{\partial \nu}(u - v) = g & \text{on } \partial D. \end{cases}$$

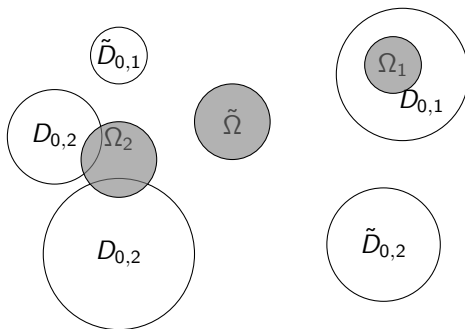
# Application to differential measurements

## Assumptions on the geometry:

$$D_0 \subset D \quad \overline{D} = \overline{\Omega} \cup \overline{D}_0 \quad \text{and } n = n_0 \text{ in } D_0 \setminus \Omega$$

$$D_0 = \bigcup_i \tilde{D}_{0,i} \cup \bigcup_i D_{0,i}.$$

$D_{0,i}$ ,  $i = 1, \dots, M$  the components of  $D_0$  that intersect with  $\Omega$ .



# Application to differential measurements

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## Comparison of ITP solutions:

**Theorem:** Assume that  $\Re(n) > \Re(n_0) > 1$  or  $\Re(n) < \Re(n_0) < 1$  in  $\Omega$ . Let  $z \in D$  and consider  $(u, v) \in L^2(D) \times L^2(D)$  (resp.  $(u_0, v_0) \in L^2(D_0) \times L^2(D_0)$ ) solutions of  $\text{ITP}(D, n, \Phi_z, \frac{\partial \Phi_z}{\partial \nu})$  (resp.  $\text{ITP}(D_0, n_0, \Phi_z, \frac{\partial \Phi_z}{\partial \nu})$ ). Then, except for a countable set of values of  $k$ ,

- ▶ If  $z \in \tilde{D}_{0,i}$  then  $v = v_0$  in  $D_0$ .
- ▶ If  $z \in D_{0,i}$ , then  $v \neq v_0$  in  $D_{0,i}$  and  $v = v_0$  in  $D_0 \setminus D_{0,i}$ .

# Application to differential measurements

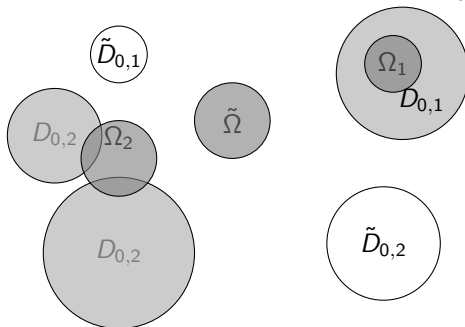
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$$D_0 = \bigcup_i \tilde{D}_{0,i} \cup \bigcup_i D_{0,i}.$$

$D_{0,i}$ ,  $i = 1, \dots, M$  the components of  $D_0$  that intersect with  $\Omega$ .

We use this to obtain characterizations of  $\tilde{\Omega}$  and  $\Omega_0 := \Omega \cup_i D_{0,i}$ .



# Characterization of $\Omega_0$ in terms of $F$ and $F_0$

$F$  farfield associated with  $D$  and  $n$ .  $B = F_{\#}$

$F_0$  farfield associated with  $D_0$  and  $n_0$ .  $B_0 = F_{0,\#}$ .

We introduce

$$\mathcal{D}(g, g_0) := |\langle B_0(g - g_0), g - g_0 \rangle|.$$

**Corollary:** Under previous assumptions on  $D$ ,  $D_0$ ,  $n$ ,  $n_0$  and  $k$  and for  $g_z^\alpha$  and  $g_{0,z}^\alpha$  the minimizing sequences associated resp with  $(F, B)$  and  $(F_0, B_0)$

- ▶ If  $z \in \bigcup_i \tilde{D}_{0,i}$  then  $\lim_{\alpha \rightarrow 0} \mathcal{D}(g_z^\alpha, g_{0,z}^\alpha) = 0$ .
- ▶ If  $z \in \tilde{\Omega}$  then  $\lim_{\alpha \rightarrow 0} \mathcal{D}(g_z^\alpha, g_{0,z}^\alpha) = \infty$ .
- ▶ If  $z \in \bigcup_i D_{0,i}$  then  $\lim_{\alpha \rightarrow 0} \mathcal{D}(g_z^\alpha, g_{0,z}^\alpha) < \infty$ .



# Characterization of $\Omega_0$ in terms of $F$ and $F_0$

The noise free case:

$$\mathcal{I}(g, g_0) := |\langle Bg, g \rangle| (1 + |\langle Bg, g \rangle| / |\langle B_0(g - g_0), g - g_0 \rangle|).$$

**Corollary:** Under previous assumptions on  $D$ ,  $D_0$ ,  $n$ ,  $n_0$  and  $k$ . For  $g_z^\alpha$  and  $g_{0,z}^\alpha$  the minimizing sequences associated resp with  $(F, B)$  and  $(F_0, B_0)$

$$\text{If } z \notin \Omega_0 \text{ then } \lim_{\alpha \rightarrow 0} \mathcal{I}(g_z^\alpha, g_{0,z}^\alpha) = \infty.$$

$$\text{If } z \in \Omega_0 \text{ then } \lim_{\alpha \rightarrow 0} \mathcal{I}(g_z^\alpha, g_{0,z}^\alpha) < \infty.$$

**Therefore**, the limit as  $\alpha \rightarrow 0$  of

$$z \mapsto 1/\mathcal{I}(g_z^\alpha, g_{0,z}^\alpha) \text{ is an indicator for } \Omega_0 = \tilde{\Omega} \cup \bigcup_i D_{0,i}$$

# Characterization of $\Omega_0$ in terms of $F$ and $F_0$

The noisy case:

For a fixed parameter  $\eta \in (0, 1)$ , we define

$$g_{0,z}^{\alpha,\delta} = \arg \min_g \alpha \left( \langle B_0^\delta g, g \rangle + \alpha^{-\eta} \delta \|B_0^\delta\| \|g\|^2 \right) + \|F_0^\delta g - \phi_z^\infty\|^2$$

$$g_z^{\alpha,\delta} = \arg \min_g \alpha \left( \langle B^\delta g, g \rangle + \alpha^{-\eta} \delta \|B^\delta\| \|g\|^2 \right) + \|F^\delta g - \phi_z^\infty\|^2$$

We then consider

$$D(\alpha, z) = \liminf_{\delta \rightarrow 0} \left\langle B_0^\delta (g_{0,z}^{\alpha,\delta} - g_z^{\alpha,\delta}), g_{0,z}^{\alpha,\delta} - g_z^{\alpha,\delta} \right\rangle$$

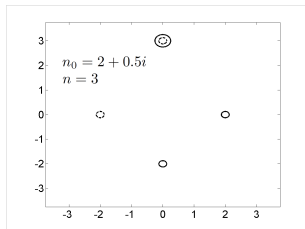
$$A(\alpha, z) = \liminf_{\delta \rightarrow 0} \left( \langle B^\delta g_z^{\alpha,\delta}, g_z^{\alpha,\delta} \rangle + \alpha^{-\eta} \delta \|B^\delta\| \|g_z^{\alpha,\delta}\|^2 \right)$$

$$\mathcal{I}(\alpha, z) := A(\alpha, z) (1 + A(\alpha, z)/D(\alpha, z)).$$

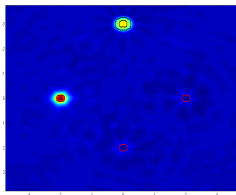
**Theorem:** Under previous assumptions on  $D$ ,  $D_0$ ,  $n$ ,  $n_0$  and  $k$ ,

$$z \in \Omega_0 \text{ iff } \lim_{\alpha \rightarrow 0} \mathcal{I}(\alpha, z) < +\infty.$$

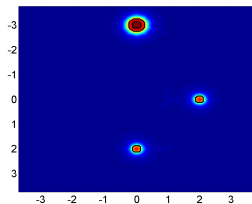
# Some numerical results



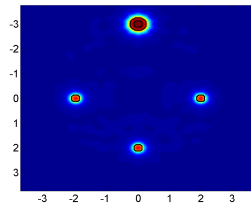
Exact configuration



Differential GLSM

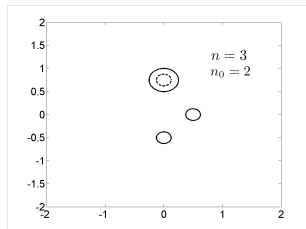


Background Reconstruction

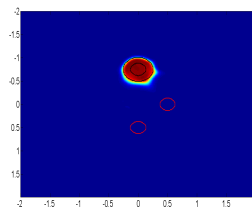


Medium Reconstruction

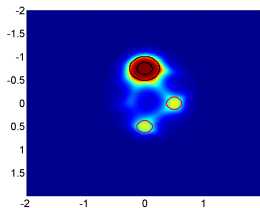
# Some numerical results



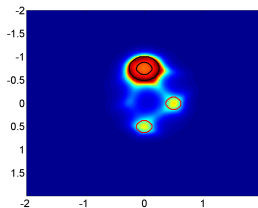
Exact configuration



Differential GLSM



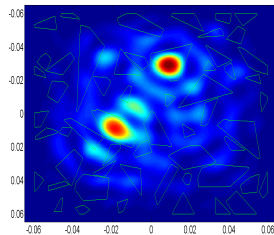
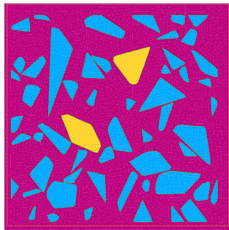
Background Reconstruction



Medium Reconstruction

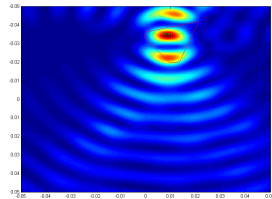
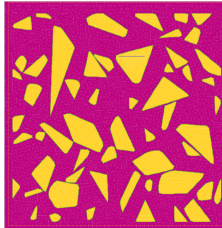
# Some numerical results for scattered background

## Two defects of type inhomogeneities

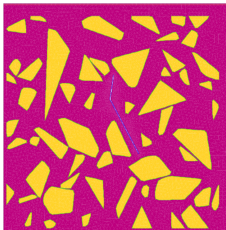


# Some numerical results for scattered background

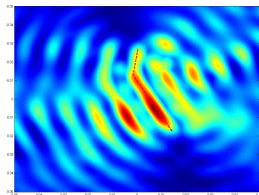
## One defect of type internal crack



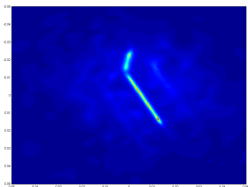
# Some numerical results for scattered background



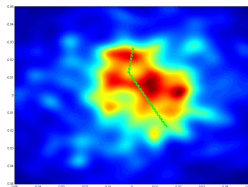
A medium size external crack



Differential GLSM



Reconstructed crack using  
the exact background



Reconstructed crack using  
a homogeneous background

**Thank you**