

Averaged Steklov eigenvalues as macroscopic indicator functions

Housseem HADDAR

INRIA & ENSTA

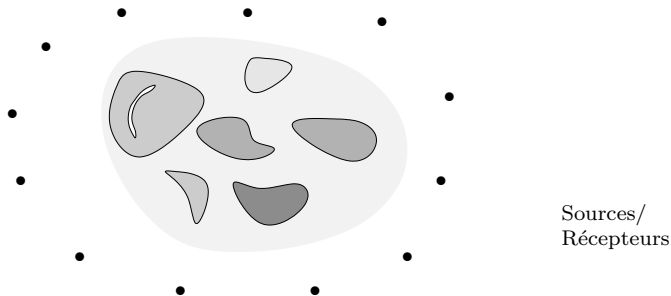
Project team IDEFIX (INRIA, EDF R&D and ENSTA)

Based on a joint work with

L. Audibert and F. Pourre

WICOM 2025

Context

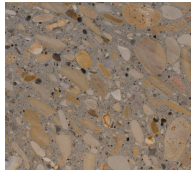


Goal: Construct “qualitative and/or quantitative” indicators on the material properties or defects inside unknown (complex) media from multi-static measurements of scattered waves.

- **Forward solvers** for the exact problem **cannot be used** since either the background cannot be correctly modeled or the defect has a complicated structure (network of cracks).

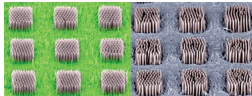
Targeted applications

- ▶ Monitor **defects** or **material quality** in concrete type materials, composite materials, etc, ...



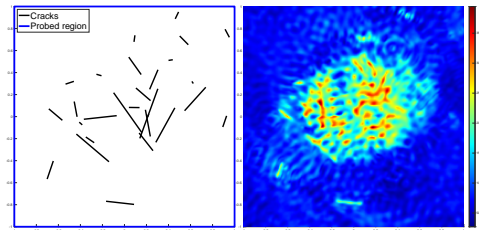
Concrete materials

- ▶ Image defects in **periodic structures** with unknown periodic pattern such as nano-grass structures.



Nano-grass structure

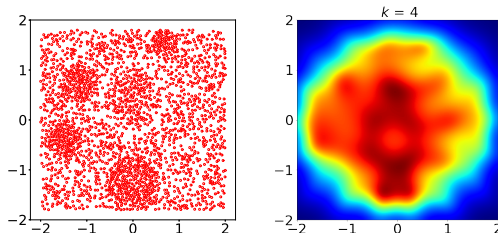
Examples of qualitative reconstructions



Images produced by “linear sampling method” using far field data ^a

^aOther/similar non iterative imaging techniques: Liu-Sini (2010), Ammari-Garnier-Kang-Park-Sølna (2011), Bonnet (2011), Boukari-H. (2013), Bellis-Bonnet (2013), Guo-Wu-Yan (2015), Daimon-Furuya-Saiin (2020), Audibert-Chesnel-Napal-H. (2022) ...

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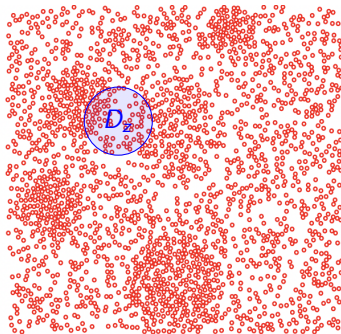


Images produced by “linear sampling method” using far field data ^a

Goal: Can we construct a “better” indicator function that also has a quantitative interpretation?

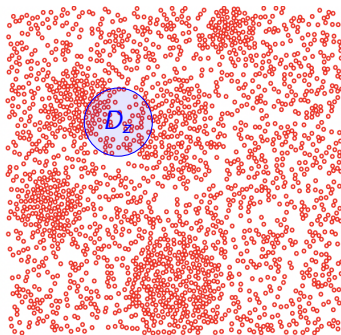
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General scheme for the new algorithm



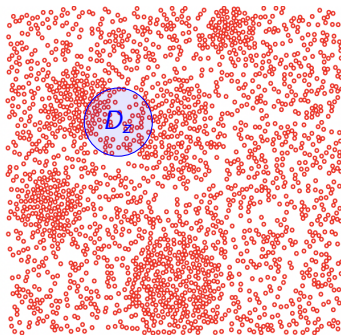
- At a location z , numerically introduce an artificial resonator D_z

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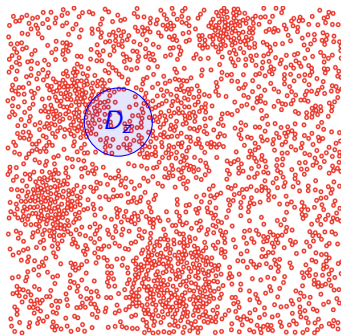
- ▶ At a location z , numerically introduce an artificial resonator D_z
- ▶ Identify the resonance parameter from measurements using the inside-outside duality :
resonant inside \Leftrightarrow silent outside

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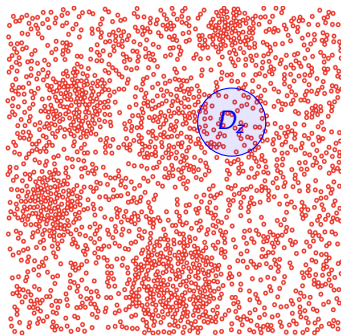
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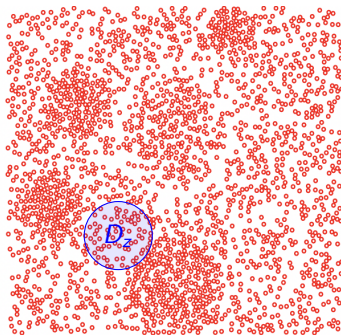
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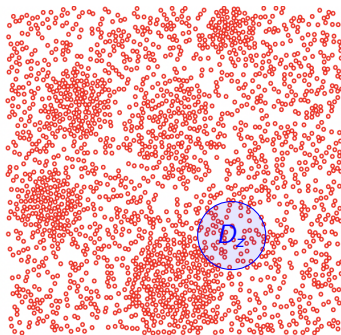
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In this talk:

- ▶ Averaged-Steklov eigenvalues as resonators
- ▶ Where do they come from and why they are good candidates
- ▶ Inside - Outside duality to identify them from measurements at fixed frequency
- ▶ From far field to localized macroscopic indicator functions (link with homogenized properties)
- ▶ Generalization. From far field to a localized DtN operator.

A simple model problem

Scalar acoustic equation for inhomogeneous media

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

The total field: $u \in H_{loc}^1(\mathbb{R}^d)$

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

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

$$u^i(\hat{x}_0, x) := e^{ikx \cdot \hat{x}_0} \quad \hat{x}_0 \in \mathbb{S}^{d-1}$$

The scattered field

$$u^s(\hat{x}_0, \cdot) = u - u^i(\hat{x}_0, \cdot) \quad \text{in } \mathbb{R}^d,$$

satisfies the Sommerfeld radiation condition

$$\lim_{r \rightarrow \infty} \int_{|x|=r} \left| \frac{\partial u^s}{\partial r} - iku^s \right|^2 ds = 0.$$

A simple model problem

Scalar acoustic equation for inhomogeneous media

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

$$u^s(\hat{x}_0, x) = \frac{e^{ik|x|}}{|x|^{(d-1)/2}} (u^\infty(\hat{x}_0, \hat{x}) + O(1/|x|))$$

Our data is formed by (noisy measurements of) so-called **far field patterns**

$$u^\infty(\hat{x}_0, \hat{x}) \text{ for all } (\hat{x}_0, \hat{x}) \in \mathbb{S}^{d-1} \times \mathbb{S}^{d-1}$$

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

$$Fg(\hat{x}) := \int_{\mathbb{S}^{d-1}} u^\infty(\hat{x}_0, \hat{x}) g(\hat{x}_0) ds(\hat{x}_0).$$

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Fg is the far field generated by an incident field $u^i = v_g$ where

$$v_g(x) := \int_{\mathbb{S}^{d-1}} u^i(\hat{x}_0, x) g(\hat{x}_0) ds(\hat{x}_0), \quad g \in L^2(\mathbb{S}^{d-1}), \quad x \in \mathbb{R}^d.$$

Averaged-Steklov eigenvalues and Inside-Outside duality

L. Audibert, F. Cakoni, and H.H., **New sets of eigenvalues in inverse scattering for inhomogeneous media and their determination from scattering data**, (2017)

L. Audibert, H.H. and F. Pourre, **Imaging highly heterogeneous media using modified transmission eigenvalues**, (2022-2024)

Averaged-Steklov eigenvalue problem

Definition Averaged-Steklov eigenvalues for n and a domain D_b are $\mu \in \mathbb{R}$ such that there exists a non trivial solution $w \in H^1(D_b)$ such that

$$\begin{cases} \Delta w + k^2 n w = 0 \text{ in } D_b, \\ \mu w + \int_{\partial D_b} \partial_\nu w ds = 0 \text{ on } \partial D_b, \end{cases} \quad (1)$$

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Where does this problem come from?

$$\begin{cases} \Delta u + k^2 n u = 0 & \text{in } D_b \\ -a \Delta v + k^2 \lambda v = 0 & \text{in } D_b \\ u = v & \text{on } \partial D_b \\ \frac{\partial u}{\partial \nu} = -a \frac{\partial v}{\partial \nu} & \text{on } \partial D_b \end{cases}$$

The largest eigenvalue $\lambda \longrightarrow \frac{\mu}{k^2 |D_b|}$ as $a \rightarrow +\infty$

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Theorem 1: Except for a discrete set of values for k^2 , problem (1) has a **unique (generically) non trivial eigenvalue** $\mu_1(n, D_b)$.

$$\mu_1(n, D_b) = - \int_{\partial D_b} \partial_\nu w_1 ds = k^2 \int_{D_b} n w_1 dx$$

where $w_1 \in H^1(D_b)$ is the unique solution of

$$\begin{cases} \Delta w_1 + k^2 n w_1 = 0 & \text{in } D_b \\ w_1 = 1 & \text{on } \partial D_b. \end{cases}$$

Averaged-Steklov eigenvalue problem

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Theorem 2: Denote by $\Lambda_0(n, D_b)$ the first eigenvalue of

$$\begin{cases} -\Delta w = \Lambda n w & \text{in } D_b \\ w = 0 & \text{on } \partial D_b \end{cases}$$

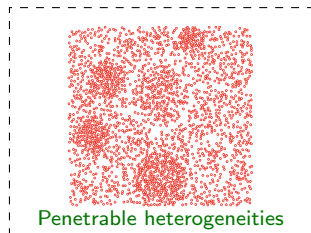
Then $\mu_1(n, D_b) > 0$ and $\mu_1(n, D_b) \rightarrow +\infty$ as $k^2 \rightarrow \Lambda_0(n, D_b)$.

Moreover we have the monotonicity property (for real n)

$$\boxed{n_1 \leq n_2 \implies \mu_1(n_1, D_b) \leq \mu_1(n_2, D_b)}$$

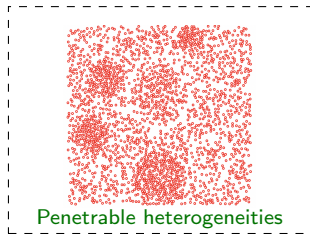
if $k^2 < \Lambda_0(n_i, D_b)$, $i = 1, 2$.

Dual definition

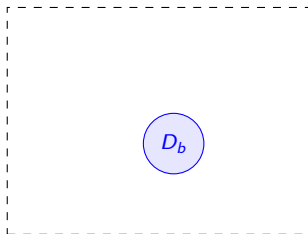


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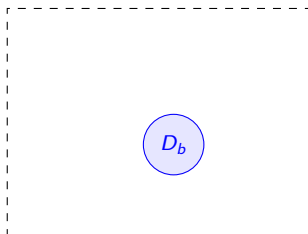
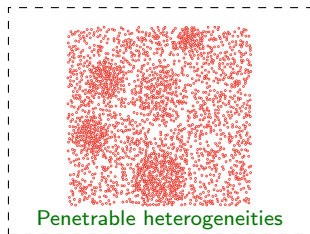


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$$\begin{cases} \Delta u_b + k^2 u_b = 0 & \text{in } \mathbb{R}^d \setminus D_b \\ \mu u_b + \int_{\partial D_b} \partial_\nu u_b ds = 0 \text{ on } \partial D_b, \\ u_b = u_b^s + u^i \text{ in } \mathbb{R}^d \setminus \overline{D_b} \\ \lim_{r=|x| \rightarrow \infty} r \left(\frac{\partial u_b^s}{\partial r} - i k u_b^s \right) = 0 \end{cases}$$

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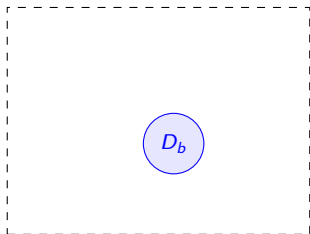
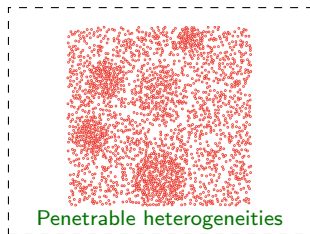
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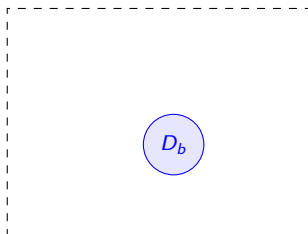
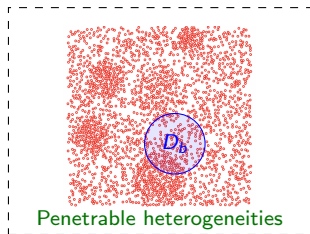
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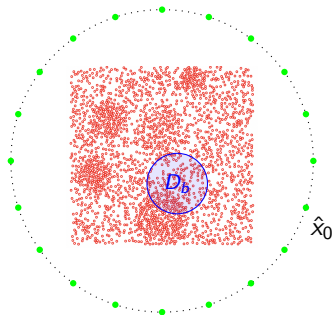
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Dual definition with far field data

$$u^i = u^i(\hat{x}_0, x) := e^{ikx \cdot \hat{x}_0} \quad \hat{x}_0 \in \mathbb{S}^{d-1}$$

$$u^s(\hat{x}_0, \cdot) \longrightarrow u^\infty(\hat{x}_0, \cdot)$$

$$u_b^{\mu, s}(\hat{x}_0, \cdot) \longrightarrow u_b^{\mu, \infty}(\hat{x}_0, \cdot)$$

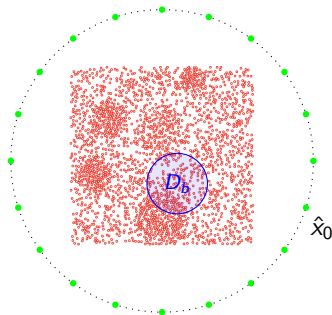


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$$Fg := \int_{\mathbb{S}^{d-1}} u^\infty(\hat{x}_0, \cdot) g(\hat{x}_0) ds(\hat{x}_0) \quad \text{and}$$

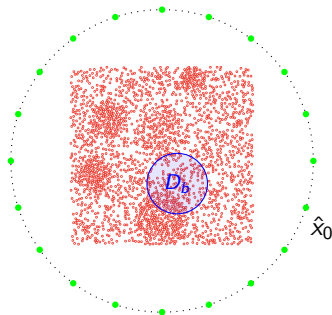
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Theorem: $\mu = \mu_1(n, D_b)$ if and only if there exists a sequence $(g_\epsilon) \in L^2(\mathbb{S}^{d-1})$ such that v_{g_ϵ} converges to some non trivial $v \in H^1(D_b) \cup L^2(D)$ and

$$\|Fg_\epsilon - F_b^\mu g_\epsilon\|_{L^2(\mathbb{S}^{d-1})} \rightarrow 0$$

Determination of TEs from far field data

Difficulty: $F - F_b^\mu$ is a compact operator... Impossible to numerically test the non injectivity of this operator.

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Two methods can be used to overcome this difficulty.

- ▶ **GLSM method:**^a Transform the non-injectivity into the **non solvability** of the far field equation used by the Linear Sampling Method.
 - ▶ Works for limited aperture
 - ▶ Needs a careful tuning of a regularization parameter
 - ▶ Hard to exploit for imaging if the eigenvalue is not isolated

^aCakoni-Colton-H. (2011), Audibert-H. (2014), Audibert-Cakoni-H. (2018), Audibert-H.-Pourre (2023), ...

Determination of TEs from far field data

Difficulty: $F - F_b^\mu$ is a compact operator... Impossible to numerically test the non injectivity of this operator.

Two methods can be used to overcome this difficulty.

- ▶ **Inside-Outside duality:**^a Use the **phase** of the eigenvalues of $F - F_b^\mu$ that accumulate at 0 as μ approaches the averaged Steklov eigenvalue.
 - ▶ Numerically cheaper,
 - ▶ Requires symmetry between emitters/receivers,
 - ▶ Is not fully justified for penetrable inclusions.
 - ▶ Hard to implement if the eigenvalue is not isolated

^aEckmann-Pillet (1995), Kirsch-Lechleiter (2013), [Lechleiter-Peters \(2015\)](#), [Audibert-Chesnel-H. \(2019\)](#), [H.-Khenissi-Mansouri \(2022\)](#),...)

The inside-outside duality method for A.S.E.

Assume that the refractive index n and μ are real valued.

- ▶ The compact operators F and F_b^μ are normal
- ▶ The scattering operators $S := \frac{2ik}{\gamma} I + F$ and $S_b^\mu := \frac{2ik}{\gamma} I + F_b^\mu$ are unitary.^a

^a $\gamma = 4\pi$ for $d = 3$ and $\gamma = \sqrt{8\pi k} e^{i\frac{\pi}{4}}$ for $d = 2$

The inside-outside duality method for A.S.E.

Assume that the refractive index n and μ are real valued.

- ▶ The compact operators F and F_b^μ are normal
- ▶ The scattering operators $S := \frac{2ik}{\gamma} I + F$ and $S_b^\mu := \frac{2ik}{\gamma} I + F_b^\mu$ are unitary.^a

Define

$$\mathbf{F}^\mu := \overline{\gamma}(S_b^\mu)^*(F - F_b^\mu),$$

- ▶ The compact operator \mathbf{F}^μ is also normal and the associated scattering operator $\mathbf{S}^\mu := \frac{2ik}{|\gamma|^2} I + \mathbf{F}^\mu$ is unitary.
- ▶ Moreover

$$\mathbf{F}^\mu = \mathcal{H}_\mu^* T_\mu \mathcal{H}_\mu.$$

with T_μ a (generically) coercive operator.

^a $\gamma = 4\pi$ for $d = 3$ and $\gamma = \sqrt{8\pi k} e^{i\frac{\pi}{4}}$ for $d = 2$

The inside-outside duality method for A.S.E.

$$\mathbf{F}^{\mu} := \bar{\gamma}(S_b^{\mu})^*(F - F_b^{\mu}) = \mathcal{H}_{\mu}^* T_{\mu} \mathcal{H}_{\mu}.$$

Theorem: Assume in addition that $n - 1 \geq 0$ and $\mu > 0$ and is not a A.S.E, then

$$T_{\mu} = T_0 + K_{\mu}$$

with T_0 positive definite and K_{μ} compact.

The inside-outside duality method for A.S.E.

$$\mathbf{F}^\mu := \bar{\gamma}(S_b^\mu)^*(F - F_b^\mu) = \mathcal{H}_\mu^* T_\mu \mathcal{H}_\mu.$$

Theorem: Assume in addition that $n - 1 \geq 0$ and $\mu > 0$ and is not a A.S.E, then

$$T_\mu = T_0 + K_\mu$$

with T_0 positive definite and K_μ compact. Therefore the eigenvalues \mathbf{F}^μ accumulate at 0 with positive real parts:

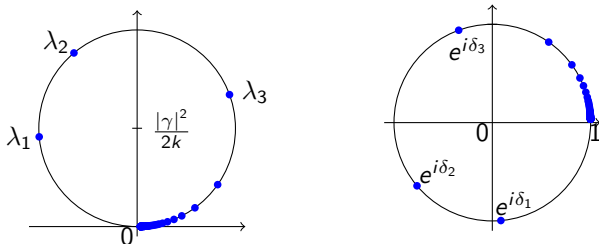


Figure: Left: eigenvalues of $\overline{\mathbf{F}}^\mu$. Right: eigenvalues of \mathbf{S}^μ .

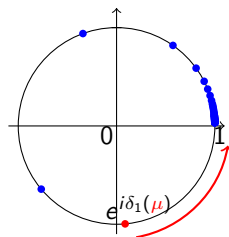
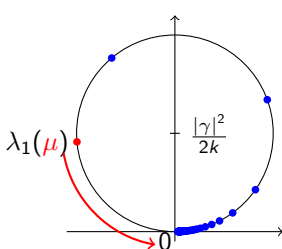
The inside-outside duality method for A.S.E.

$$\mathbf{F}^\mu := \bar{\gamma}(S_b^\mu)^*(F - F_b^\mu) = \mathcal{H}_\mu^* T_\mu \mathcal{H}_\mu.$$

Theorem: Assume in addition that $n - 1 \geq 0$ and $\mu > 0$ and is not a A.S.E, then

$$T_\mu = T_0 + K_\mu$$

with T_0 positive definite and K_μ compact.



Theorem: $\mu \rightarrow \mu_1(n, D_b)$ with $\mu > \mu_1(n, D_b)$ if and only if $\lambda_1(\mu) \rightarrow 0$ or equivalently $\delta_1(\mu) \rightarrow 2\pi$.

A numerical illustration

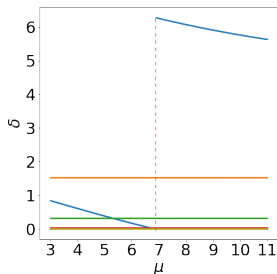


Figure: The curves $\mu \rightarrow \delta_m(\mu)$ for D_b a disc of radius $\rho = 0.4333$, $k = 3$ and $n = 1$. The red dashed line indicates the exact A.S.E.

A numerical illustration

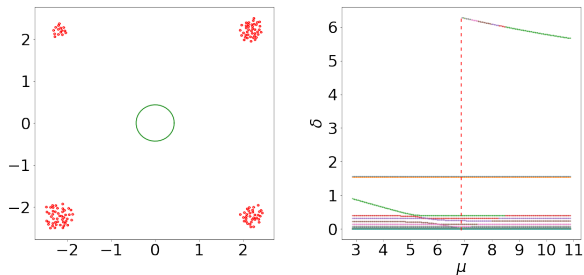


Figure: Left: The domain D (in red) and D_b , the disk of radius $\rho = 0.433$ (in green). Right: Plot of the curves $\mu \rightarrow \delta_m(\mu)$ for $k = 3$ and $n = 2$ inside the red discs. The red dashed line indicates the exact A.S.E.

A numerical illustration

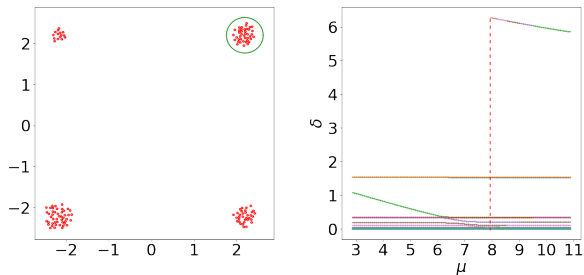
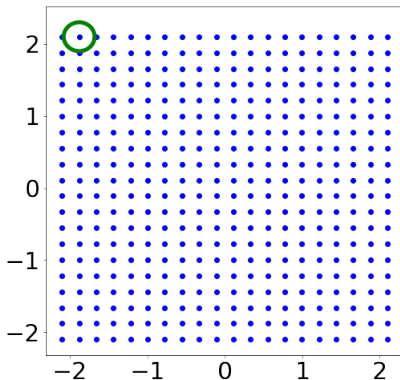


Figure: Left: The domain D (in red) and D_b , the ball of radius $\rho = 0.433$ centered at $(2.2, 2.2)$ (in green). Right: Plot of the curves $\mu \rightarrow \delta_m(\mu)$ for $k = 3$ and $n = 2$ inside the red circles. The red dashed line indicates the A.S.E. numerically evaluated by solving the eigenvalue problem.

A first application to the inverse problem

A qualitative type algorithm:

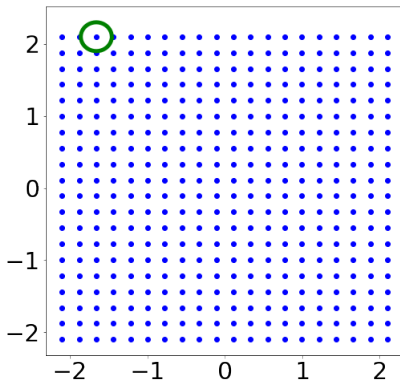
- Sweep a fixed geometry D_b , a ball of radius ρ , over center positions y in a sampling of the probed region.



A first application to the inverse problem

A qualitative type algorithm:

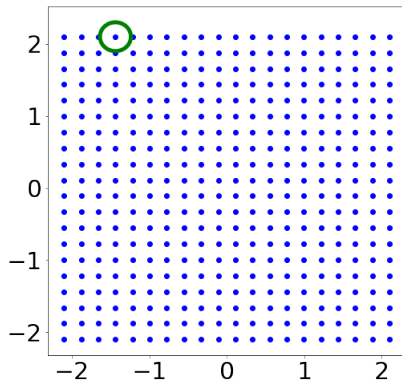
- Sweep a fixed geometry D_b , a ball of radius ρ , over center positions y in a sampling of the probed region.



A first application to the inverse problem

A qualitative type algorithm:

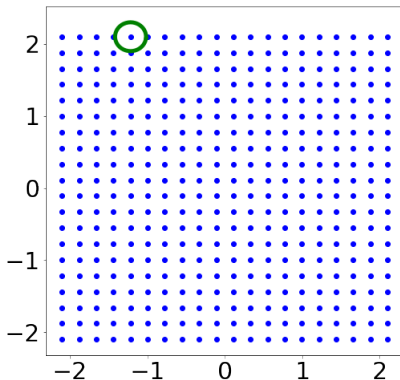
- Sweep a fixed geometry D_b , a ball of radius ρ , over center positions y in a sampling of the probed region.



A first application to the inverse problem

A qualitative type algorithm:

- Sweep a fixed geometry D_b , a ball of radius ρ , over center positions y in a sampling of the probed region.



A first application to the inverse problem

A qualitative type algorithm:

- ▶ Sweep a fixed geometry D_b , a ball of radius ρ , over center positions y in a sampling of the probed region.
- ▶ For each position of D_b , determine $\mu(n, y)$ from measurements.
- ▶ Plot the function $\mathcal{I} : y \rightarrow \mu(n, y) - \mu(1, y)$

Remark

- ▶ If $n|_{D_b} > 1$, then $\mathcal{I}(y) > 0$
- ▶ The larger the $n|_{D_b}$, the larger is $\mathcal{I}(y)$

Numerical illustrations for the first qualitative algorithm

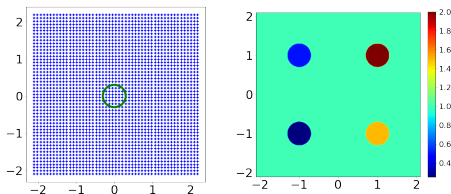


Figure: Four diffracting discs of radius 0.3, $n = 0.25$ (bottom left), $n = 0.5$ (top left), $n = 1.5$ (bottom right) and $n = 2$ (upper right).

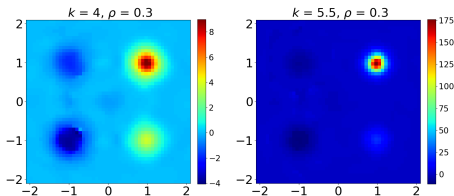


Figure: Indicator function with $\rho = 0.3$ and the noise level 1%.

Numerical illustrations for the first qualitative algorithm

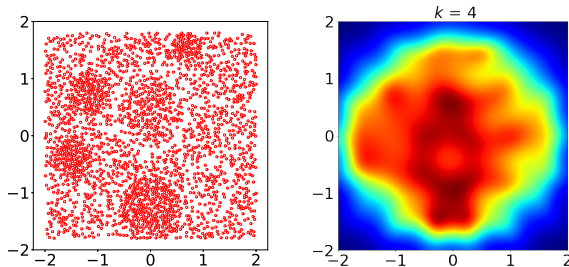


Figure: Reconstruction obtained using the Linear Sampling Method (1% of added noise, $k = 4$).

Numerical illustrations for the first qualitative algorithm

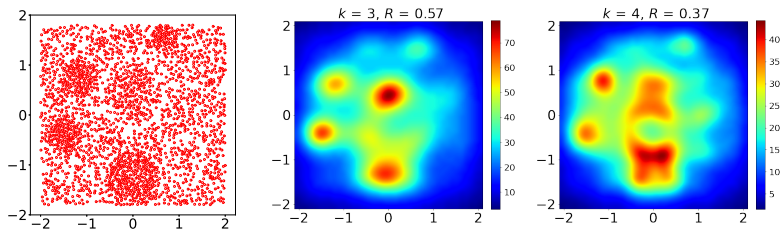


Figure: Reconstructions using the indicator function $y \mapsto I(y)$ (1% of added noise, $k = 3$ (middle) and $k = 4$ (right)).

Numerical illustrations for the first qualitative algorithm

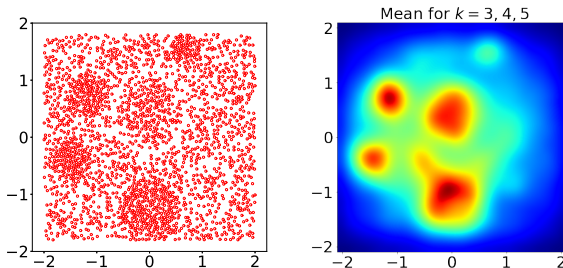


Figure: Mean values of indicator functions for three wavenumbers:
 $k = 3, 4, 5$

From far field to local macroscopic properties

$$\mu_1(n, D_b) = - \int_{\partial D_b} \partial_\nu w_1 ds = k^2 \int_{D_b} n w_1 dx$$

where $w_1 \in H^1(D_b)$ is the unique solution of

$$\begin{cases} \Delta w_1 + k^2 n w_1 = 0 & \text{in } D_b \\ w_1 = 1 & \text{on } \partial D_b. \end{cases}$$

A classical result from homogenization theory:

Assume that $n = n_\delta \rightarrow \bar{n}$ as $\delta \rightarrow 0$ weak star in $L^\infty(D_b)$. Then $\mu_1(n_\delta, D_b) \rightarrow \mu_1(\bar{n}, D_b)$ as $\delta \rightarrow 0$.

From far field to local macroscopic properties

$$\mu_1(n, D_b) = - \int_{\partial D_b} \partial_\nu w_1 ds = k^2 \int_{D_b} n w_1 dx$$

where $w_1 \in H^1(D_b)$ is the unique solution of

$$\begin{cases} \Delta w_1 + k^2 n w_1 = 0 & \text{in } D_b \\ w_1 = 1 & \text{on } \partial D_b. \end{cases}$$

Application: If D_b is a disc of radius ρ and \bar{n} is a constant

$$\mu_1(\bar{n}, D_b) = 2\pi \rho k \sqrt{\bar{n}} \frac{J_1(k \rho \sqrt{\bar{n}})}{J_0(k \rho \sqrt{\bar{n}})}. \quad (2)$$

$$\bar{n} \mapsto \mu_1(\bar{n}, D_b)$$

is a bijection if $\boxed{\bar{n} \leq \gamma_0 / (k \rho)^2}$ where γ_0 is the first zero of J_0 .

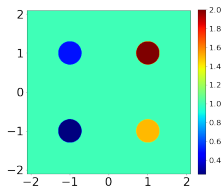
A second application to the inverse problem

A quantitative inversion algorithm:

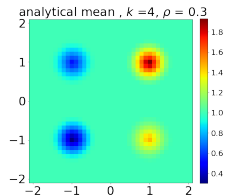
- ▶ Choose D_b to be a disc of center y and radius ρ .
- ▶ Compute $\mu_1(n, D_b)$ and deduce the constant $\bar{n}(y)$ such that $\mu_1(\bar{n}, D_b) = \mu_1(n, D_b)$

\bar{n} is expected to be an approximation of $\frac{1}{|D_b|} \int_{D_b} n(x) dx$

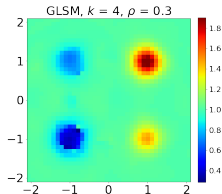
A second application to the inverse problem



Four diffracting discs of radius 0.3



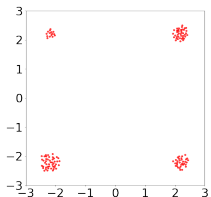
$$y \mapsto \frac{1}{|D_b|} \int_{D_b} n(x) dx$$



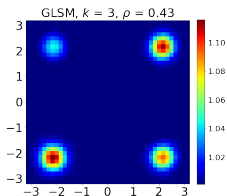
$$y \mapsto \bar{n}(y)$$

$\mu_1(n, D_b)$ determined from far fields

A second application to the inverse problem

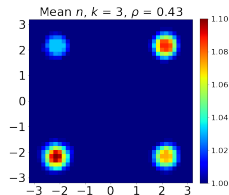


$n = 2$ inside the discs of radii 0.02

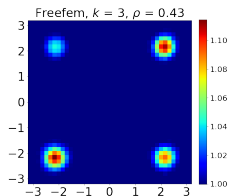


$$y \mapsto \bar{n}(y)$$

$\mu_1(n, D_b)$ determined from far fields



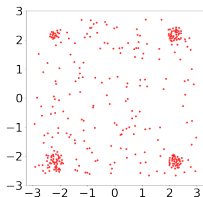
$$y \mapsto \frac{1}{|D_b|} \int_{D_b} n(x) dx$$



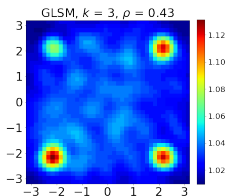
$$y \mapsto \bar{n}(y)$$

$\mu_1(n, D_b)$ computed numerically

A second application to the inverse problem

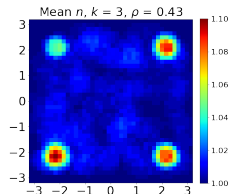


$n = 2$ inside the discs of radii 0.02

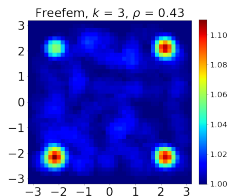


$$y \mapsto \bar{n}(y)$$

$\mu_1(n, D_b)$ determined from far fields



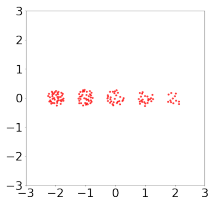
$$y \mapsto \frac{1}{|D_b|} \int_{D_b} n(x) dx$$



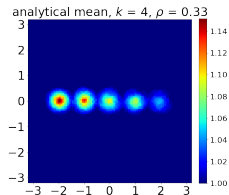
$$y \mapsto \bar{n}(y)$$

$\mu_1(n, D_b)$ computed numerically

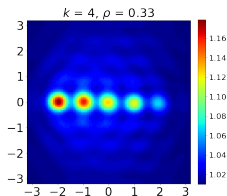
A second application to the inverse problem



Four diffracting discs of radius 0.3



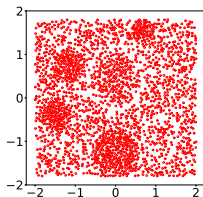
$$y \mapsto \frac{1}{|D_b|} \int_{D_b} n(x) dx$$



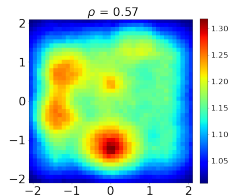
$$y \mapsto \bar{n}(y)$$

$\mu_1(n, D_b)$ determined from far fields

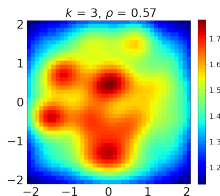
A second application to the inverse problem



$n = 2$ inside the discs of radii 0.02



$$y \mapsto \frac{1}{|D_b|} \int_{D_b} n(x) dx$$



$$y \mapsto \bar{n}(y)$$

$\mu_1(n, D_b)$ determined from far fields

A generalization

Let \bar{f} , be some given function in $H^{1/2}(\partial D_b)$. One can repeat the same arguments replacing the boundary conditions on ∂D_b by

$$\mu u_b + \bar{f} \int_{\partial D_b} \partial_\nu u_b \cdot \bar{f} \, ds = 0 \text{ on } \partial D_b,$$

The corresponding \bar{f} -averaged Steklov eigenvalue is $\mu = \mu(n, D_b, \bar{f})$ the unique non trivial eigenvalue of

$$\begin{cases} \Delta w + k^2 n w = 0 \text{ in } D_b, \\ \mu w + \bar{f} \int_{\partial D_b} \partial_\nu w \cdot \bar{f} \, ds = 0 \text{ on } \partial D_b, \end{cases} \quad (3)$$

A generalization

$$\begin{cases} \Delta w + k^2 n w = 0 & \text{in } D_b, \\ \mu w + \mathbf{f} \int_{\partial D_b} \partial_\nu w \cdot \bar{\mathbf{f}} \, ds = 0 & \text{on } \partial D_b, \end{cases} \quad (3)$$

Theorem: Assume that $k^2 < \eta(n, D_b)$. Then problem (3) has a unique eigenvalue $\mu(n, D_b, \mathbf{f})$.

$$\mu(n, D_b, \mathbf{f}) = - \int_{\partial D_b} \partial_\nu w_{\mathbf{f}} \cdot \bar{\mathbf{f}} \, ds$$

where $w_{\mathbf{f}} \in H^1(D_b)$ is the unique solution of

$$\begin{cases} \Delta w_{\mathbf{f}} + k^2 n w_{\mathbf{f}} = 0 & \text{in } D_b \\ w_{\mathbf{f}} = \mathbf{f} & \text{on } \partial D_b. \end{cases}$$

A generalization

$$\begin{cases} \Delta w + k^2 n w = 0 \text{ in } D_b, \\ \mu w + \mathbf{f} \int_{\partial D_b} \partial_\nu w \cdot \bar{\mathbf{f}} \, ds = 0 \text{ on } \partial D_b, \end{cases} \quad (3)$$

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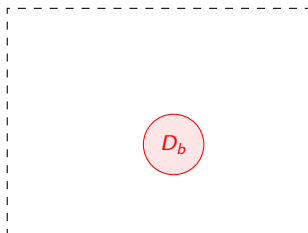
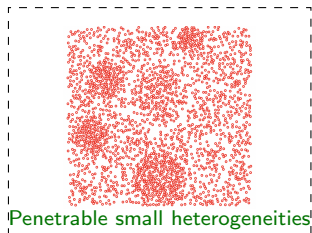
$$\begin{cases} \Delta w_{\mathbf{f}} + k^2 n w_{\mathbf{f}} = 0 & \text{in } D_b \\ w_{\mathbf{f}} = \mathbf{f} & \text{on } \partial D_b. \end{cases}$$

- ▶ $\mu(n, D_b, \mathbf{f})$ can be determined from far field data
- ▶ Knowing $\mu(n, D_b, \mathbf{f})$ for all \mathbf{f} is equivalent to knowing the DtN map $\mathbf{f} \mapsto \partial_\nu w_{\mathbf{f}}$

Conclusion / Some perspectives

- We proposed a new imaging algorithm based on the use of averaged Steklov eigenvalues
- We designed two versions of the algorithm: a qualitative one and a quantitative one
- Validation for penetrable scatterers using synthetic data
- Further explore the potentialities of knowing $\mu(n, D_b, \textcolor{red}{f})$
- Extend to qualitative/quantitative evaluation of anisotropy

Modified Transmission Eigenvalues



$$\begin{cases} \Delta u + k^2 n u = 0 & \text{in } \mathbb{R}^d \\ n \neq 1 \text{ in } D \text{ and } n = 1 \text{ in } \mathbb{R}^d \setminus D \\ u = u^s + u^i \text{ in } \mathbb{R}^d \setminus \Gamma \\ \lim_{r=|x| \rightarrow \infty} r \left(\frac{\partial u^s}{\partial r} - i k u^s \right) = 0 \end{cases}$$

$$\begin{cases} \Delta u_b + k^2 n_b u_b = 0 & \text{in } \mathbb{R}^d \\ n_b = \lambda \text{ in } D_b \text{ and } n_b = 1 \text{ in } \mathbb{R}^d \setminus D_b \\ u_b = u_b^s + u^i \text{ in } \mathbb{R}^d \setminus \overline{D_b} \\ \lim_{r=|x| \rightarrow \infty} r \left(\frac{\partial u_b^s}{\partial r} - i k u_b^s \right) = 0 \end{cases}$$

The Modified TE are the $\lambda \in \mathbb{C}$ such that there exists a non trivial incident field $u^i = \mathbf{v} \in L^2(D_b \cup D)$ such that

$$u_b^s(\mathbf{v}) \equiv u^s(\mathbf{v}) \text{ in } \mathbb{R}^d \setminus \{D_b \cup D\}.$$

Modified Transmission Eigenvalues

$$u_b^s \equiv u^s \text{ in } \mathbb{R}^d \setminus \{D_b \cup D\}?$$

if and only if $w = u - u_b \in H_0^2(D_b \cup D)$ and $v = u_b \in L^2(D_b \cup D)$

$$\begin{cases} \Delta w + k^2 w = k^2(n_b(\lambda) - n)v & \text{in } D_b \cup D, \\ \Delta v + k^2 n_b(\lambda)v = 0 & \text{in } D_b \cup D \end{cases} \quad (4)$$

In the connected component D_b (assuming that $D_b \cap D = \emptyset$),
 $w \in H_0^2(D_b)$ and $v \in L^2(D_b)$,

$$\begin{cases} \Delta w + k^2 w = k^2(\lambda - n)v & \text{in } D_b, \\ \Delta v + k^2 \lambda v = 0 & \text{in } D_b \end{cases} \quad (5)$$

Modified Transmission Eigenvalues

$$u_b^s \equiv u^s \text{ in } \mathbb{R}^d \setminus \{D_b \cup D\}?$$

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$$\begin{cases} \Delta w + k^2 w = k^2(\lambda - n)v & \text{in } D_b, \\ \Delta v + k^2 \lambda v = 0 & \text{in } D_b \end{cases} \quad (5)$$

This is (still) a **non self-adjoint eigenvalue problem**: $w \in H_0^2(D_b)$

$$(\Delta + k^2 \lambda) \frac{1}{\lambda - n} (\Delta + k^2 n) w = 0 \text{ in } D_b$$

A "better" choice for the background

One can use of an **artificial metamaterial** for the background (as proposed in [Audibert-Cakoni-H. \(2017\)](#))

The background total field verifies

$$\nabla \cdot a_b \nabla u_b + k^2 n_b u_b = 0 \quad \text{in } \mathbb{R}^d$$

$$a_b = n_b = 1 \quad \text{in } \mathbb{R}^d \setminus D_b$$

$$a_b = -a < -1 \text{ and } n_b = \lambda \in \mathbb{R} \quad \text{in } D_b$$

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$a_b = -a < -1$ and $n_b = \lambda \in \mathbb{R}$	$\text{in } D_b$

The Modified transmission eigenvalue problem becomes: $u \in H^1(D_b)$,
and $v \equiv u_b \in H^1(D_b)$,

$$\begin{cases} \Delta u + k^2 u = 0 & \text{in } D_b \\ -a \Delta v + k^2 \lambda v = 0 & \text{in } D_b \\ u = v & \text{on } \partial D_b \\ \frac{\partial u}{\partial \nu} = -a \frac{\partial v}{\partial \nu} & \text{on } \partial D_b \end{cases}$$

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$a_b = n_b = 1$	$\text{in } \mathbb{R}^d \setminus D_b$
$a_b = -a < -1$ and $n_b = \lambda \in \mathbb{R}$	$\text{in } D_b$

If n is real valued \Rightarrow Eigenvalue problem for a **selfadjoint compact operator**: $(u, v) \in \mathcal{H}(D_b)$

$$\int_{D_b} \nabla u \cdot \nabla \bar{u}' \, dx + a \int_{D_b} \nabla v \cdot \nabla \bar{v}' \, dx - k^2 \int_{D_b} n u \bar{u}' \, dx = -k^2 \lambda \int_{D_b} v \bar{v}' \, dx$$

for all $(u', v') \in \mathcal{H}(D_b)$ where

$$\mathcal{H}(D_b) := \{(u, v) \in H^1(D_b) \times H^1(D_b) \text{ such that } u = v \text{ on } \partial D_b\}.$$

A "better" choice for the background

Theorem: There is at least one positive eigenvalue $\lambda_a(k, n)$. Assume that $k^2 < \Lambda_0(n, D_b)$ the first eigenvalue of

$$\begin{cases} -\Delta w = \Lambda n w & \text{in } D_b \\ w = 0 & \text{on } \partial D_b \end{cases}$$

Then the largest positive eigenvalue $\lambda_a(k, n)$ satisfies

$$\lambda_a(k, n) = \sup_{(u,v) \in \mathcal{H}(D_b), v \neq 0} \frac{k^2 \int_{D_b} n |u|^2 dx - \int_{D_b} (|\nabla u|^2 + a |\nabla v|^2) dx}{k^2 \int_{D_b} |v|^2 dx}.$$

- ▶ $\lambda_a(k, n) \rightarrow \infty$ as $k^2 \rightarrow \Lambda_0(n, D_b)$.
- ▶ $n_1 \leq n_2 \implies \lambda_a(k, n_1) \leq \lambda_a(k, n_2)$

Limiting problem as $a \rightarrow \infty$

Theorem: Assume that $k^2 < \Lambda_0(n, D_b)$. Then,

$$\lambda_a(k, n) \longrightarrow \frac{\mu}{k^2 |D_b|} \text{ as } a \rightarrow +\infty$$

where $\mu > 0$ is an eigenvalue of the problem $w \in H^1(D_b)$,

$$\begin{cases} \Delta w + k^2 n w = 0 \text{ in } D_b, \\ \mu w + \int_{\partial D_b} \partial_\nu w ds = 0 \text{ on } \partial D_b, \end{cases} \quad (6)$$