

Analysis of a Monte-Carlo Nystrom method and well-posedness of the Foldy-Lax approximation

Florian Feppon – Habib Ammari

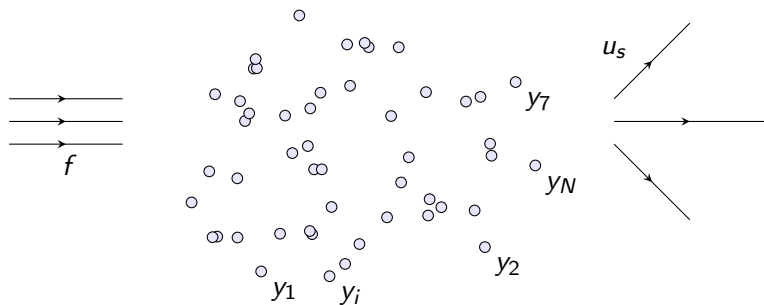
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Seminar for Applied Mathematics

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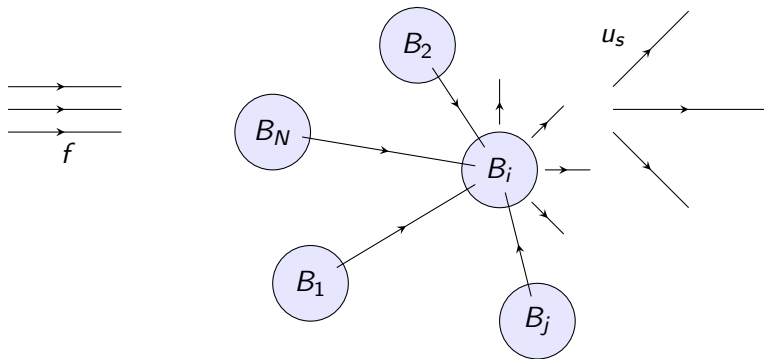
Motivation: the Foldy-Lax approximation

Acoustic scattering of an incident field f through N obstacles $(B_i)_{1 \leq i \leq N}$ located at $(y_i)_{1 \leq i \leq N}$:



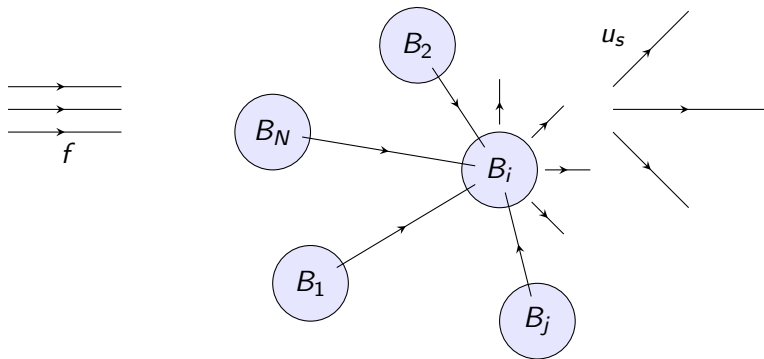
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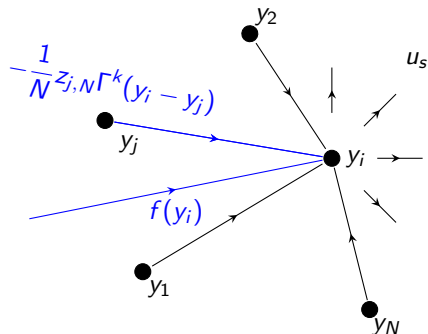
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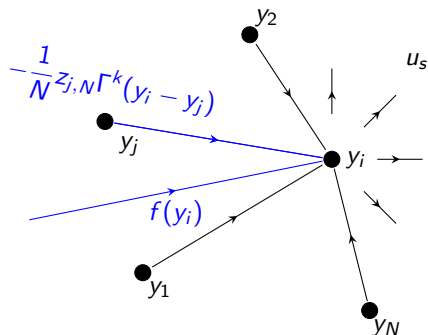


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1. The scattered field can be approximated by the contribution of N point-sources located at the centers $(y_i)_{1 \leq i \leq N}$:

$$u_s(y) \simeq -\frac{1}{N} \sum_{i=1}^N z_{i,N} \Gamma^k(y - y_i)$$

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$\Gamma^k(y)$ is e.g. the (outgoing) fundamental solution to the Helmholtz equation:

$$(\Delta + k^2)\Gamma^k = \delta_0 \text{ in } \mathbb{R}^d,$$

$$\Gamma^k(y) = \begin{cases} -\frac{i}{4} H_0^{(1)}(k|y|) & \text{if } d = 2, \\ -\frac{e^{ik|y|}}{4\pi|y|} & \text{if } d = 3 \end{cases}$$

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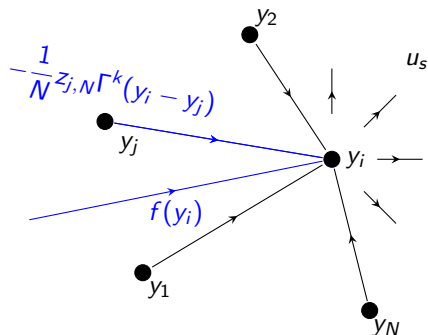
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$\Gamma^k(\cdot - y)$ is the wave pattern generated by a point source located at y .

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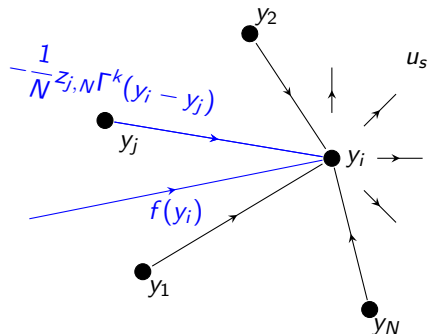


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2. The intensity $z_{i,N}$ of the wave field scattered by the source y_i is the contribution of the field scattered by the other sources $(y_j)_{1 \leq j \neq i \leq N}$ and of the incident field $f(y_i)$:

$$z_{i,N} = f(y_i) - \frac{1}{N} \sum_{j \neq i} z_{j,N} \Gamma^k(y_j - y_i).$$

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We obtain the following linear system for the wave field intensity $(z_{i,N})_{1 \leq i \leq N}$:

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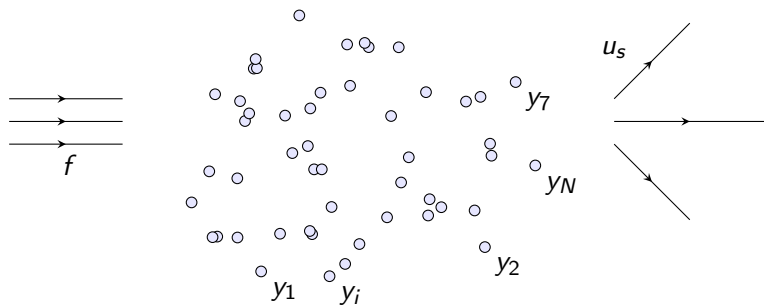
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3. i.e. can we prove a convergence result $z_{i,N} \rightarrow z(y_i)$ as $N \rightarrow +\infty$?
In that case (2) is an equation characterizing the effective medium associated to the random point cloud $(y_i)_{1 \leq i \leq N}$.

A Monte-Carlo Nystrom method

Replace Γ^k with a general kernel $k(y, y')$:

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N, \quad (1)$$

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If $z_{i,N} \rightarrow z(y_i)$ as $N \rightarrow +\infty$, then (1) can also be viewed as a Monte-Carlo method for solving (2).

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2. Numerical illustration in 1D and 2D
3. Sketch of the proofs: random operator theory

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(i) $\Omega \subset \mathbb{R}^d$ is a bounded Lipschitz domain and

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(ii) $(y_i)_{1 \leq i \leq N}$ are independent samples of a probability distribution $\rho(y) dy$ with density $\rho \in L^\infty(\Omega, \mathbb{R}^+)$ (satisfying $\int_{\Omega} \rho(y) dy = 1$).

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(iii) The integral equation (2) is well-posed.

Result 1: well-conditioning

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N \quad \iff (I + A_N)z_N = F \quad (1)$$

with $z_N = (z_{i,N})_{1 \leq i \leq N}$ and $F = (f(y_i))_{1 \leq i \leq N}$ and where $(A_N)_{1 \leq i, j \leq N}$ is the random matrix defined by

$$A_{N,ij} = \begin{cases} \frac{1}{N} k(y_i, y_j) & \text{if } i \neq j, \\ 0 & \text{if } i = j. \end{cases}$$

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Proposition

Assume (i), (ii) and (iii). Then with probability one, there exists $N_0 \in \mathbb{N}$ such that the matrix $\mathbf{I} + \mathbf{A}_N$ is invertible for any $N \geq N_0$, and there exists a constant $C > 0$ independent of N such that

$$\forall N \geq N_0, \|\|(\mathbf{I} + \mathbf{A}_N)^{-1}\|\|_2 \leq C$$

where $\|\| \cdot \|\|_2$ is the operator norm ($\|\|A\|\|_2 := \sup_{\|x\|_2=1} \|Ax\|_2$).

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where $\|\cdot\|_2$ is the operator norm ($\|A\|_2 := \sup_{\|x\|_2=1} \|Ax\|_2$).

So (1) is well-posed if the continuous problem is well-posed.

Result 2: convergence of the Nystrom interpolant

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N, \quad (1)$$

$$z(y) + \int_{\Omega} z(y') k(y, y') \rho(y') dy' = f(y), \quad y \in \Omega. \quad (2)$$

Let $z_N(y)$ be the Nystrom interpolant

$$z_N(y) := f(y) - \frac{1}{N} \sum_{i=1}^N k(\cdot, y_i) z_{N,i}, \quad y \in \Omega.$$

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If $k(y, y') = \Gamma^k(y - y')$, then

$$z_N(y) = f(y) - \frac{1}{N} \sum_{i=1}^N \Gamma^k(y - y_i) z_{N,i} = f(y) + u_s(y), \quad y \in \Omega.$$

is the total wave field, and $z_N - f$ is the scattered field.

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1. (1) is invertible for $N \geq N_0$ when \mathcal{H}_{N_0} is realized
2. z_N converges to z at rate $O(N^{-\frac{1}{2}})$ in a mean-square sense:

$$\mathbb{E}[\|z_N - z\|_{L^2(\Omega)}^2 | \mathcal{H}_{N_0}]^{\frac{1}{2}} \leq CN^{-\frac{1}{2}}.$$

Result 3: point-wise convergence

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2. the vector $(z_{N,i})_{1 \leq i \leq N}$ converges to the point-wise values $(z(y_i))_{1 \leq i \leq N}$ at rate $O(N^{-\frac{1}{2}})$ in a mean-square sense:

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Numerical 1D example

- ▶ We consider $k(y, y') := |y - y'|^{-\alpha}$ with $\alpha = 0.4 < 1/2$ on the interval $\Omega = (0, 1)$ and the integral equation

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- ▶ We draw M times a sample of N random points $(y_i^p)_{1 \leq i \leq N}$ independently from the uniform distribution in $(0, 1)$ for $1 \leq p \leq M$.

Numerical 1D example

- ▶ We consider $k(y, y') := |y - y'|^{-\alpha}$ with $\alpha = 0.4 < 1/2$ on the interval $\Omega = (0, 1)$ and the integral equation

$$z(y) + \int_0^1 k(y, y')z(y')dy' = f(y), \quad y \in (0, 1). \quad (2)$$

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$$z_{N,i}^p + \frac{1}{N} \sum_{j \neq i} k(y_i^p, y_j^p)z_{N,j}^p = f(y_i^p), \quad 1 \leq i \leq N.$$

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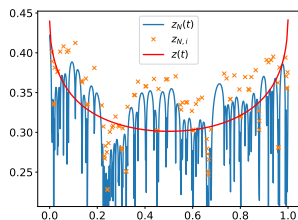
$$z_{N,i}^p + \frac{1}{N} \sum_{j \neq i} k(y_i^p, y_j^p)z_{N,j}^p = f(y_i^p), \quad 1 \leq i \leq N.$$

- ▶ We solve (2) accurately with a Nystrom method on a regular grid and we estimate the mean-square error:

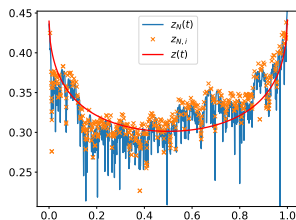
$$\text{MSE} := \mathbb{E} \left[\frac{1}{N} \sum_{i=1}^N |z_{N,i} - z(y_i)|^2 \right]^{\frac{1}{2}} \simeq \sqrt{\frac{1}{MN} \sum_{p=1}^M \sum_{i=1}^N |z_{N,i}^p - z(y_i^p)|^2}.$$

Numerical 1D example

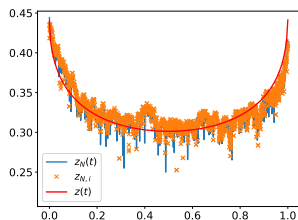
Case 1: $f(y) = 1$



(a) $N = 100$



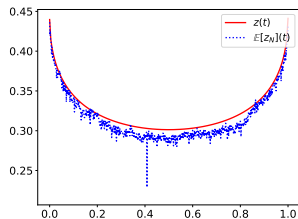
(b) $N = 500$



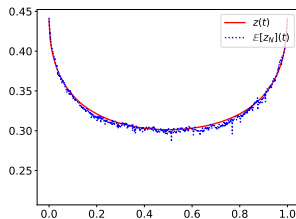
(c) $N = 2,000$

Numerical 1D example

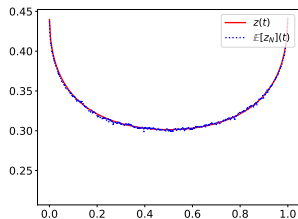
Case 1: $f(y) = 1$



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(c) $N = 2,000$

Figure: Empirical average of the Nystrom interpolant $\mathbb{E}[z_N]$.

Numerical 1D example

Case 1: $f(y) = 1$

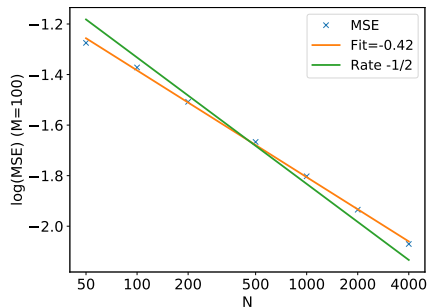
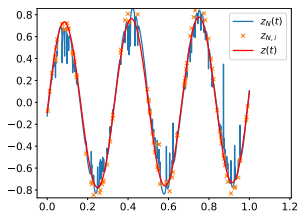


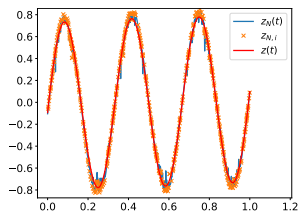
Figure: Mean-square error MSE.

Numerical 1D example

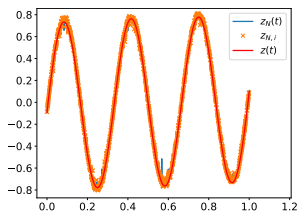
Case 2: $f(y) = \sin(6\pi y)$



(a) $N = 100$



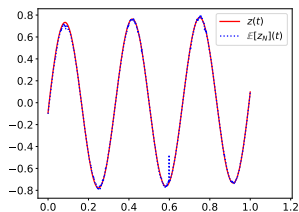
(b) $N = 500$



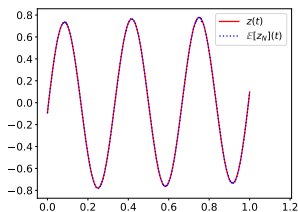
(c) $N = 2,000$

Numerical 1D example

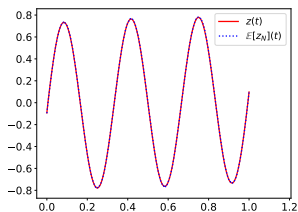
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Numerical 1D example

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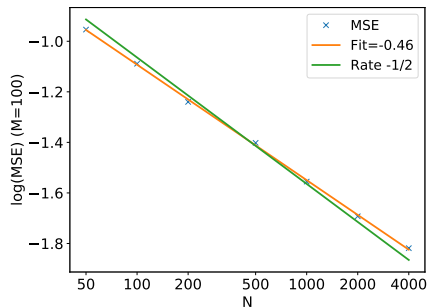


Figure: Mean-square error MSE.

Numerical 2D example

We solve with our Monte-Carlo method the following Lippmann-Schwinger equation:

$$\begin{cases} (\Delta + k^2 n_\Omega)z = 0 \text{ in } \mathbb{R}^2, \\ (\partial_r - ik)(z - u_{in}) = O(|x|^{-2}) \text{ as } r \rightarrow +\infty, \end{cases} \quad (3)$$

whose solution z is the scattered field produced by an incident wave u_{in} propagating through a material with refractive index $n_\Omega(x)$ given by

$$n_\Omega(x) = \begin{cases} m \text{ if } x \in \Omega, \\ 1 \text{ if } x \in \mathbb{R}^2 \setminus \Omega, \end{cases}$$

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The integral formulation of (3) is

$$z(y) + (m - 1)k^2 \int_{\Omega} \Gamma^k(y - y')z(y')dy' = u_{in}(y), \quad y \in \Omega, \quad (2)$$

Numerical 2D example

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¹Aussal and Alouges, *Gypsilab* (2018)

²Averseng, *Fast discrete convolution in \mathbb{R}^2 with radial kernels using non-uniform fast Fourier transform with nonequispaced frequencies* (2020)

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- ▶ We solve (2) with the finite-element method¹.

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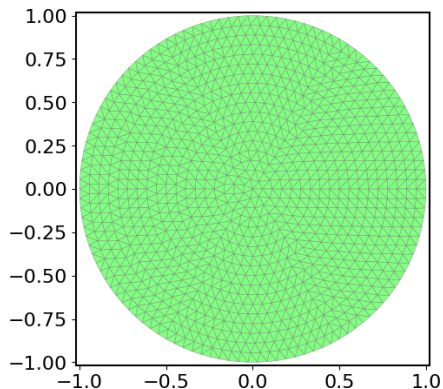
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- ▶ We solve (2) with the finite-element method¹.
- ▶ We solve (1) for $500 \leq N \leq 40,000$ using the Efficient Bessel Decomposition method².

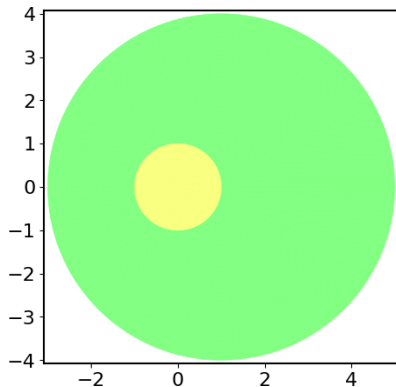
¹Aussal and Alouges, *Gypsilab* (2018)

²Averseng, *Fast discrete convolution in \mathbb{R}^2 with radial kernels using non-uniform fast Fourier transform with nonequispaced frequencies* (2020)

Numerical 2D example

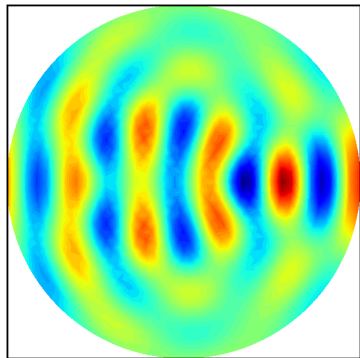


(a) The discretization mesh \mathcal{T} considered for the acoustic obstacle Ω (the unit disk).

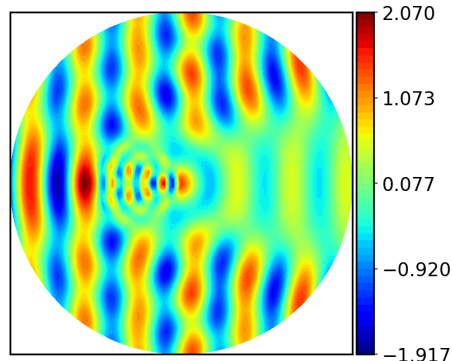


(b) The surrounding disk Ω' (the disk centered at $(1, 0)$ of radius 4, in green) containing the acoustic obstacle Ω (in yellow).

Numerical 2D example



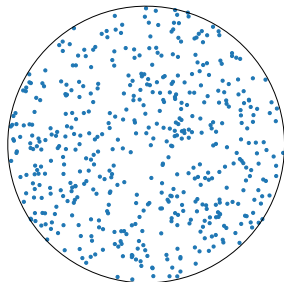
(a) Plot of the solution z in the interior domain Ω .



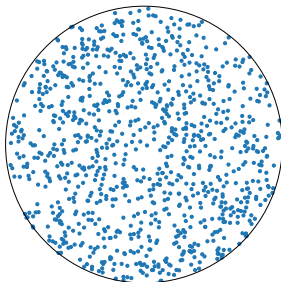
(b) Plot of the solution z in the exterior domain Ω' .

Thanks Martin Averseng and Ignacio Labarca.

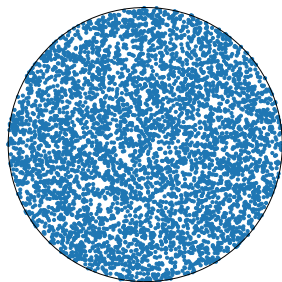
Numerical 2D example



(a) $N = 500$



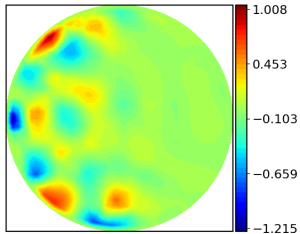
(b) $N = 1,000$



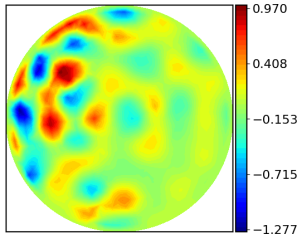
(c) $N = 5,000$

Figure: Samples of N random points drawn randomly and independently from the uniform distribution in the unit disk.

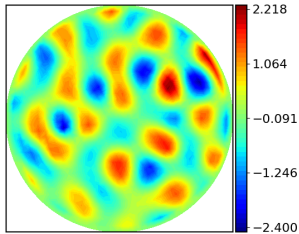
Numerical 2D example



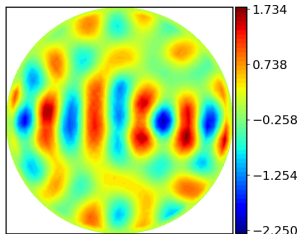
(a) $N = 500$



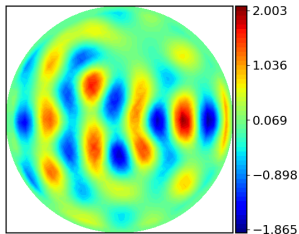
(b) $N = 1,000$



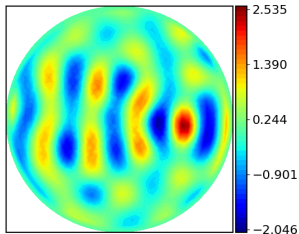
(c) $N = 5,000$



(d) $N = 10,000$



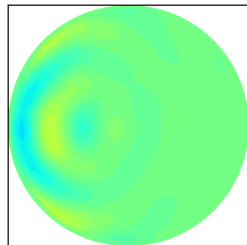
(e) $N = 20,000$



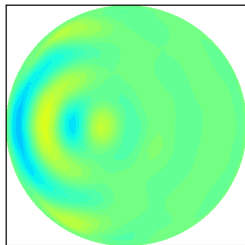
(f) $N = 40,000$

Figure: Monte-Carlo solutions $(z_i^P)_{1 \leq i \leq N}$

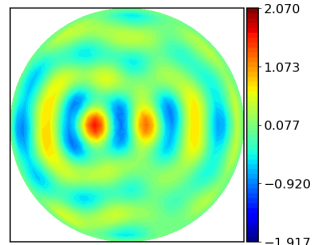
Numerical 2D example



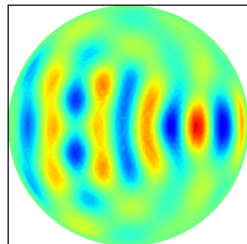
(a) $N = 500$



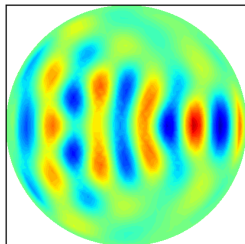
(b) $N = 1,000$



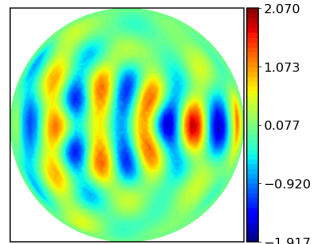
(c) $N = 5,000$



(d) $N = 10,000$



(e) $N = 20,000$



(f) $N = 40,000$

Figure: Averaged field $\mathbb{E}[(\hat{z}_i^P)]$ at the vertices of the mesh \mathcal{T} .

Numerical 2D example

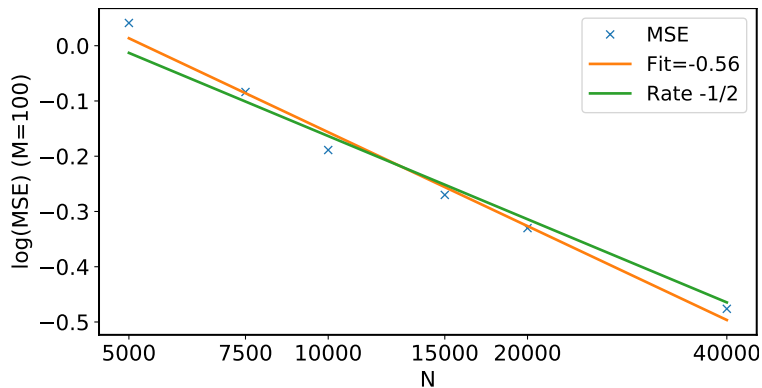


Figure: Mean-square error MSE.

Outline

1. Well-posedness and convergence results
2. Numerical illustration in 1D and 2D
3. Sketch of the proofs: random operator theory

Sketch of the proof

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N, \quad (1)$$

► (1) is equivalent to finding a function $z_N \in L^2(\Omega)$ such that

$$z_N(y) + \frac{1}{N} \sum_{j=1}^N k(y, y_j) z_N(y_j) = f(y), \quad \forall y \in \Omega. \quad (a)$$

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- (a) rewrites

$$\left(I + \frac{1}{N} \sum_{i=1}^N A_i \right) z_N = f \text{ with } \begin{array}{l} A_i : L^2(\Omega, \mathbb{C}) \rightarrow L^2(\Omega, \mathbb{C}) \\ z \quad \mapsto k(\cdot, y_i) z(y_i). \end{array}$$

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$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N, \quad (1)$$

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- A_i are independent realizations of the random operator

$$\begin{array}{l} A : \Omega \times L^2(\Omega, \mathbb{C}) \rightarrow L^2(\Omega, \mathbb{C}) \\ (y, z) \mapsto k(\cdot, y) z(y). \end{array} \quad (0.1)$$

Sketch of the proof

(a) rewrites

$$\left(I + \frac{1}{N} \sum_{i=1}^N A_i \right) z_N = f$$

Proposition

Let $(A_i)_{i \in \mathbb{N}}$ be a family of independent realizations of a given bounded random operator $A : L^2(\Omega, \mathbb{C}) \rightarrow L^2(\Omega, \mathbb{C})$. Then as $N \rightarrow +\infty$,

$$\frac{1}{N} \sum_{i=1}^N A_i \longrightarrow \mathbb{E}[A],$$

where the convergence holds at the rate $O(N^{-\frac{1}{2}})$ in the following mean-square sense:

$$\mathbb{E} \left[\left\| \frac{1}{N} \sum_{i=1}^N A_i - \mathbb{E}[A] \right\|^2 \right]^{\frac{1}{2}} \leq \frac{\mathbb{E}[\|A - \mathbb{E}[A]\|^2]^{\frac{1}{2}}}{\sqrt{N}} \text{ for any } N \in \mathbb{N}.$$

Sketch of the proof

For the random operator

$$\begin{aligned} A &: \Omega \times L^2(\Omega, \mathbb{C}) \rightarrow L^2(\Omega, \mathbb{C}) \\ (y, z) &\mapsto k(\cdot, y)z(y), \end{aligned}$$

the expectation $\mathbb{E}[A]$ is given by

$$\mathbb{E}[A] : z \mapsto \int_{\Omega} k(\cdot, y)z(y)\rho(y)dy.$$

Sketch of the proof

Proposition

Let A be a bounded random operator and $(A_i)_{i \in \mathbb{N}}$ be a sequence of independent realizations of A . Then for any $\epsilon > 0$ sufficiently small, with probability one, any

$\lambda \in B(-1, \epsilon)$ belongs to the resolvent set of $\frac{1}{N} \sum_{i=1}^N A_i$ for N large enough:

$$\left(\lambda I - \frac{1}{N} \sum_{i=1}^N A_i \right)^{-1} \rightarrow (\lambda I - \mathbb{E}[A])^{-1}$$

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- ▶ In particular ($\lambda = -1$), $I + \frac{1}{N} \sum_{i=1}^N A_i$ is invertible for N large enough.

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- ▶ In particular ($\lambda = -1$), $I + \frac{1}{N} \sum_{i=1}^N A_i$ is invertible for N large enough.
- ▶ The convergence holds at rate $O(N^{-1/2})$ in the operator norm $\| \cdot \|$ of $L^2(\Omega)$; it yields

$$\mathbb{E}[\|z_N - z\|_{L^2(\Omega)}^2 | \mathcal{H}_{N_0}]^{\frac{1}{2}} \leq CN^{-\frac{1}{2}}.$$

Result 2: convergence of the Nystrom interpolant

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N, \quad (1)$$

$$z(y) + \int_{\Omega} z(y') k(y, y') \rho(y') dy' = f(y), \quad y \in \Omega. \quad (2)$$

Let $z_N(y)$ be the Nystrom interpolant

$$z_N(y) := f(y) - \frac{1}{N} \sum_{i=1}^N k(\cdot, y_i) z_{N,i}, \quad y \in \Omega.$$

Proposition

There exists an event \mathcal{H}_{N_0} satisfying $\mathbb{P}(\mathcal{H}_{N_0}) \rightarrow 1$ as $N_0 \rightarrow +\infty$ such that

1. (1) is invertible for $N \geq N_0$ when \mathcal{H}_{N_0} is realized
2. z_N converges to z at rate $O(N^{-\frac{1}{2}})$ in a mean-square sense:

$$\mathbb{E}[\|z_N - z\|_{L^2(\Omega)}^2 | \mathcal{H}_{N_0}]^{\frac{1}{2}} \leq CN^{-\frac{1}{2}}.$$

From operators to matrices

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N, \quad (1)$$

$$z(y) + \int_{\Omega} z(y') k(y, y') \rho(y') dy' = f(y), \quad y \in \Omega. \quad (2)$$

It remains to obtain

- ▶ The well-conditioning of the linear system (1)

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It remains to obtain

- ▶ The well-conditioning of the linear system (1)
- ▶ The point-wise convergence $z_{i,N} \rightarrow z(y_i)$.

Result 3: point-wise convergence

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N, \quad (1)$$

$$z(y) + \int_{\Omega} z(y') k(y, y') \rho(y') dy' = f(y), \quad y \in \Omega. \quad (2)$$

Proposition

For the same event \mathcal{H}_{N_0} satisfying $\mathbb{P}(\mathcal{H}_{N_0}) \rightarrow 1$ as $N_0 \rightarrow +\infty$:

1. (1) is invertible for $N \geq N_0$ when \mathcal{H}_{N_0} is realized
2. the vector $(z_{N,i})_{1 \leq i \leq N}$ converges to the point-wise values $(z(y_i))_{1 \leq i \leq N}$ at rate $O(N^{-\frac{1}{2}})$ in a mean-square sense:

$$\mathbb{E} \left[\frac{1}{N} \sum_{i=1}^N |z_{N,i} - z(y_i)|^2 \mid \mathcal{H}_{N_0} \right]^{\frac{1}{2}} \leq CN^{-\frac{1}{2}}.$$

Result 1: well-conditioning

$$z_{i,N} + \frac{1}{N} \sum_{j=1}^N z_{j,N} k(y_i, y_j) = f(y_i), \quad 1 \leq i \leq N \iff (\mathbf{I} + A_N)z_N = F \quad (1)$$

with $z_N = (z_{i,N})_{1 \leq i \leq N}$ and $F = (f(y_i))_{1 \leq i \leq N}$ and where $(A_N)_{1 \leq i, j \leq N}$ is the random matrix defined by

$$A_{N,ij} = \begin{cases} \frac{1}{N} k(y_i, y_j) & \text{if } i \neq j, \\ 0 & \text{if } i = j. \end{cases}$$

Proposition

Assume (i), (ii) and (iii). Then with probability one, there exists $N_0 \in \mathbb{N}$ such that the matrix $\mathbf{I} + A_N$ is invertible for any $N \geq N_0$, and there exists a constant $C > 0$ independent of N such that

$$\forall N \geq N_0, \|\|(\mathbf{I} + A_N)^{-1}\|\|_2 \leq C$$

where $\|\| \cdot \|\|_2$ is the operator norm ($\|\|A\|\|_2 := \sup_{\|x\|_2=1} \|Ax\|_2$).

So (1) is well-posed if the continuous problem is well-posed.

Result 1: well-conditioning

- ▶ We know that $B(-1, \epsilon)$ belongs to the resolvent set of $(I + A_N)$.

³Bandtlow, *Estimates for norms of resolvents and an application to the perturbation of spectra* (2004)

Result 1: well-conditioning

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- ▶ We use the following resolvent estimate from ³:

$$\| (I + A_N)^{-1} \|_2 \leq \frac{1}{d(-1, \sigma(A_N))} \exp \left(\frac{1}{2} \frac{\text{Tr}(\overline{A_N^T} A_N)}{d(-1, \sigma(A_N))} + \frac{1}{2} \right).$$

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- ▶ Since the vector $v_N := (v_{N,i})_{1 \leq i \leq N}$ defined by $v_{N,i} := z_{N,i} - z(y_i)$ satisfies

$$(I + A_N)v_N = -r_N$$

with $\mathbb{E}[|r_N|_2^2] = O(N^{-1/2})$, we obtain the point-wise bound.

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The full details have been submitted in the preprint

Feppon F. and Ammari H., *Analysis of a Monte-Carlo Nystrom Method*.
Submitted. (2021).

Thank you for your attention.