

Asymptotic models for the multiple electromagnetic wave scattering problem by small obstacles

Justine Labat, Victor Péron, Sébastien Tordeux

PhD student in Applied Mathematics

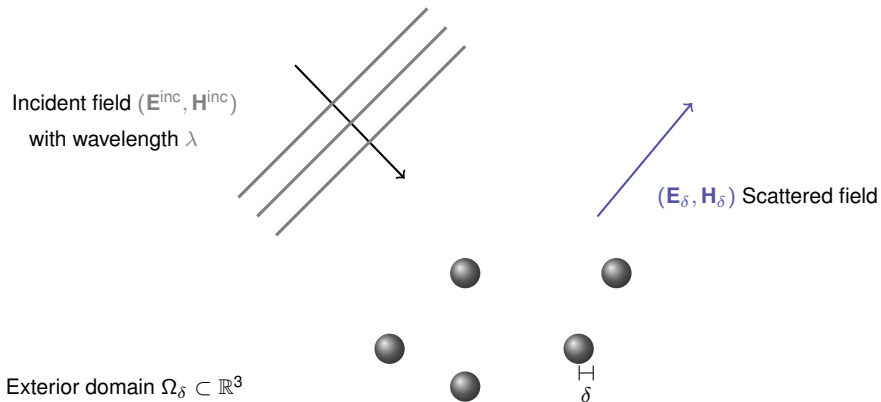
EPC Magique 3D — UPPA-E2S, INRIA Bordeaux Sud-Ouest, LMAP UMR CNRS 5142

Journées Ondes Sud-Ouest

Le Barp, March 13, 2019

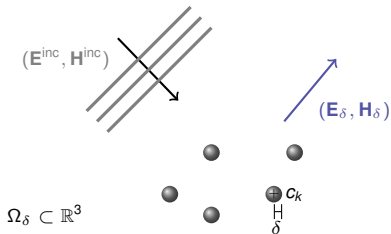


3D Scattering problem by small obstacles



Asymptotic assumption : $\delta \ll \lambda$

Model problem



- Time-harmonic domain
- Homogeneous & isotropic medium Ω_δ
- Perfect conductors of characteristic length δ

Time-harmonic Maxwell equations

$$\begin{cases} \mathbf{curl} \mathbf{E}_\delta - i\kappa \mathbf{H}_\delta = 0 & \text{in } \Omega_\delta \\ \mathbf{curl} \mathbf{H}_\delta + i\kappa \mathbf{E}_\delta = 0 & \text{in } \Omega_\delta \end{cases}$$

$$\text{with } \kappa^2 = \omega^2 \mu \left(\epsilon + \frac{i\sigma}{\omega} \right), \Im(\kappa) \geq 0$$

Boundary condition

$$\mathbf{n} \times \mathbf{E}_\delta = -\mathbf{n} \times \mathbf{E}^{\text{inc}} \quad \text{on } \partial\Omega_\delta$$

Silver-Müller radiation condition

$$r(\mathbf{H}_\delta \times \hat{\mathbf{x}} - \mathbf{E}_\delta) \xrightarrow[r \rightarrow \infty]{} 0 \quad \text{unif. in } \hat{\mathbf{x}} = \frac{\mathbf{x}}{r}$$

Mathematical well-posedness

For any $\mathbf{E}^{\text{inc}} \in \mathbf{H}_{\text{loc}}(\mathbf{curl}, \Omega_\delta)$ there exist a unique solution $(\mathbf{E}_\delta, \mathbf{H}_\delta) \in \mathbf{H}_{\text{loc}}(\mathbf{curl}, \Omega_\delta)^2$ to the exterior Maxwell problem.

Applications

- Medical imaging
- Civil engineering
- Nuclear industry
- Atmospheric science
- Astronomy
- Climatology
- ...

Inverse problem ?

- Locate heterogeneities, cracks, ...
- Characterize them

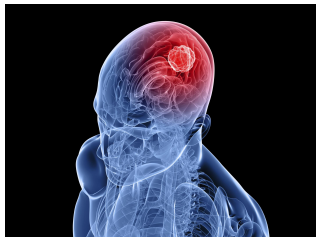


Figure: Small tumor detection¹



Figure: Non-destructive testing²

¹ <https://medcitynews.com/wp-content/uploads/2017/01/GettyImages-94456546-600x450.jpg>

² <https://tri-intl.com/wp-content/uploads/2017/07/NonDestructiveTesting.jpg>

Asymptotic models

- ✗ Restricted to small obstacles
- ✓ Low computational cost
- ✓ Meshless method

Workflow

Asymptotic models

- ✗ Restricted to small obstacles
- ✓ Low computational cost
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Multiple scattering



Foldy-Lax model

- ✓ Interactions taken into account
- ✓ Low computational cost
- ✓ Meshless method

Superposition
principle



Born approximation

- ✗ No interaction between the obstacles
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(Non-exhaustive) alternatives

- Boundary integral equations
 - ▶ Boundary element methods
 - ▶ Spectral-based methods
- Enrichment of approximation spaces
 - ▶ Trefftz-based methods
 - ▶ Extended FEM
 - ▶ Partition of Unity method

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Some references

- Historic references
 - ▶ Rayleigh (1884), Foldy (1945), Lax (1951)
- **Small defect** theory
 - ▶ Il'in (1992), Maz'ya-Nazarov-Plamenevskii (2000)

- Acoustic obstacle
 - ▶ Ammari-Kang (2003), Ramm (2005), Claeys (2008)
- Time-dependent domain
 - ▶ Mattesi (2014), Korikov (2015), Marmorat (2015)
- **Electromagnetic** obstacle
 - ▶ Vogelius-Volkov (2000), Ammari-Vogelius-Volkov (2001), Korikov-Plamenevskii (2017)

- **Foldy** theory
 - ▶ Martin (2004), Cassier-Hazard (2013), Bendali-Cocquet-Tordeux (2014), Challa-Hu-Sini (2014)
- High-order **spectral** algorithms
 - ▶ Xu (1995), Ganesh-Hawkins (2009), Barucq-Chabassier-Pham-Tordeux (2017)
- Inverse problem
 - ▶ Volkov (2001), Ammari-Kang (2004), Challa-Sini (2012)

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Outline

1. Asymptotic models

- Single-scattering
- Application: Born approximation
- Multiple-scattering: Foldy-Lax model
- Numerical results

2. Spectral method: Spherical case

- Discretization
- Numerical convergence
- Comparison with asymptotic models
- Comparison with finite element solutions

3. Conclusions and perspectives

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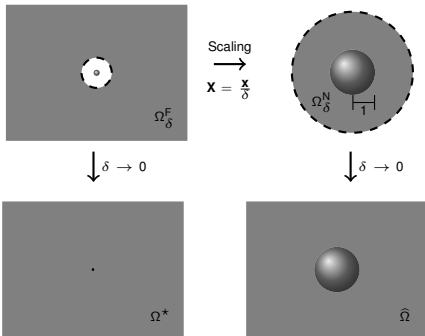
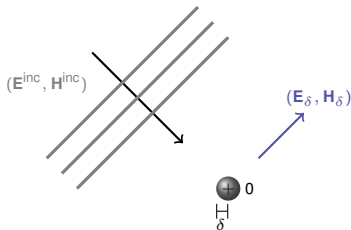
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3. Conclusions and perspectives

Approximation of solution to single scattering

Method of **matched asymptotic expansions**

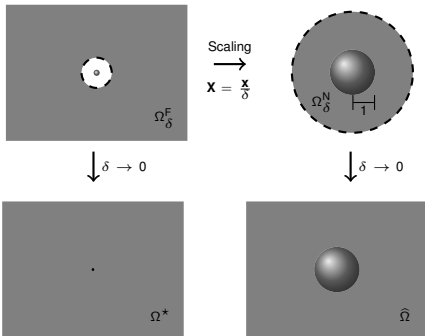
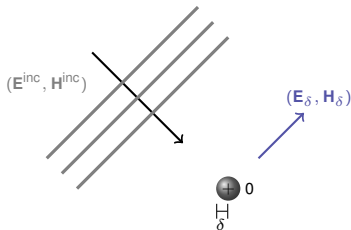
- Domain decomposition
- Local approximations
- Matching procedure



Approximation of solution to single scattering

Method of **matched asymptotic expansions**

- Domain decomposition
- **Local approximations**
- Matching procedure



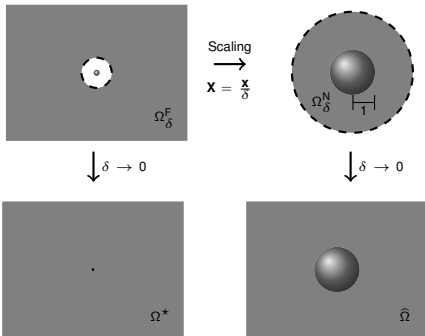
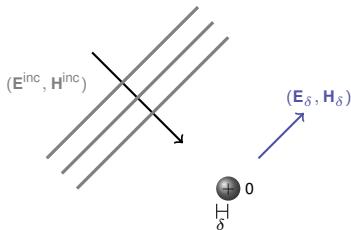
Asymptotic expansions

- **Far field** expansion in $\Omega^* = \mathbb{R}^3 \setminus \{0\}$
- **Near field** expansion in $\hat{\Omega} = \mathbb{R}^3 \setminus \overline{\mathcal{B}(0, 1)}$

Approximation of solution to single scattering

Method of **matched asymptotic expansions**

- Domain decomposition
- Local approximations
- **Matching procedure**



Asymptotic expansions

- **Far field** expansion in $\Omega^* = \mathbb{R}^3 \setminus \{0\}$
- **Near field** expansion in $\hat{\Omega} = \mathbb{R}^3 \setminus \overline{\mathcal{B}(0, 1)}$

related by an asymptotic process

Asymptotic expansions

Near field expansion

$$\mathbf{E}_\delta(\delta\mathbf{X}) \approx \widehat{\mathbf{E}}_0(\mathbf{X}) + \delta\widehat{\mathbf{E}}_1(\mathbf{X}) + \delta^2\widehat{\mathbf{E}}_2(\mathbf{X}) + \dots$$

$$\mathbf{H}_\delta(\delta\mathbf{X}) \approx \widehat{\mathbf{H}}_0(\mathbf{X}) + \delta\widehat{\mathbf{H}}_1(\mathbf{X}) + \delta^2\widehat{\mathbf{H}}_2(\mathbf{X}) + \dots$$

$\mathbf{X} = \frac{\mathbf{x}}{\delta}$: fast variable

Far field expansion

$$\mathbf{E}_\delta(\mathbf{x}) \approx \delta^3\widetilde{\mathbf{E}}_3(\mathbf{x}) + \delta^4\widetilde{\mathbf{E}}_4(\mathbf{x}) + \delta^5\widetilde{\mathbf{E}}_5(\mathbf{x}) + \dots$$

$$\mathbf{H}_\delta(\mathbf{x}) \approx \delta^3\widetilde{\mathbf{H}}_3(\mathbf{x}) + \delta^4\widetilde{\mathbf{H}}_4(\mathbf{x}) + \delta^5\widetilde{\mathbf{H}}_5(\mathbf{x}) + \dots$$

- For an obstacle of arbitrary shape

Numerical solution of elementary problems **independent of δ**

- ▶ Near-field: **quasi-static** problems
- ▶ Far-field: **time-harmonic** problems + equivalent multipole distributions

- For a **spherical** obstacle

- ▶ Analytical solutions of elementary problems
- ▶ Equivalent **multipole distributions**

Near-field approximation: Spherical case

Close to the obstacle : **quasi-static** approximation

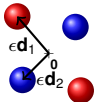
Superposition of electric fields generated by **multipole distributions**:

$$\widehat{\mathbf{E}}_p = \sum_{\substack{n=1 \\ |n-p| \text{ odd}}}^{p+1} \boldsymbol{\varepsilon}_{n,\text{elec}}^{\text{stat}}[\mathbf{u}_{1,n}^{p,E}, \dots, \mathbf{u}_{n,n}^{p,E}] + \sum_{\substack{n=1 \\ |n-p| \text{ even}}}^{p+1} \boldsymbol{\varepsilon}_{n,\text{mag}}^{\text{stat}}[\mathbf{u}_{1,n}^{p,H}, \dots, \mathbf{u}_{n,n}^{p,H}] \quad p = 0, 1, 2, \dots$$

Dipole ($n = 1$)



Quadrupole ($n = 2$)



$\boldsymbol{\varepsilon}_n^{\text{stat}}[\mathbf{d}_1, \dots, \mathbf{d}_n]$ electric fields

- generated by **charges** OR **currents**
- electric/magnetic **2^n -point charges**
 - ▶ electric charge: $\boldsymbol{\varepsilon} = -\nabla \mathcal{V}$
 - ▶ magnetic current: $\boldsymbol{\varepsilon} = -\text{curl } \mathcal{A}$
- of **moments** ($\mathbf{d}_1, \dots, \mathbf{d}_n$)
- defined by an **asymptotic** process

Charges

Currents



Near-field approximation: Spherical case

Close to the obstacle : **quasi-static** approximation

Superposition of electric fields generated by **multipole distributions**:

$$\hat{\mathbf{E}}_p = \sum_{\substack{n=1 \\ |n-p| \text{ odd}}}^{p+1} \boldsymbol{\varepsilon}_{n,\text{elec}}^{\text{stat}} [\mathbf{u}_{1,n}^{p,E}, \dots, \mathbf{u}_{n,n}^{p,E}] + \sum_{\substack{n=1 \\ |n-p| \text{ even}}}^{p+1} \boldsymbol{\varepsilon}_{n,\text{mag}}^{\text{stat}} [\mathbf{u}_{1,n}^{p,H}, \dots, \mathbf{u}_{n,n}^{p,H}] \quad p = 0, 1, 2, \dots$$

	Order	Dipole		Quadrupole		Octupole	
		Electric	Magnetic	Electric	Magnetic	Electric	Magnetic
Near field	0	●	●				
	1	●	●	●	●		
	2	●	●	●	●	●	●

● Electric field

● Magnetic field

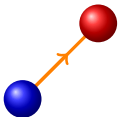
Far-field approximation

Far from the obstacle: **time-harmonic** approximation

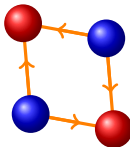
Superposition of electric fields generated by **multipole distributions**:

$$\tilde{\mathbf{E}}_{p+3} = \sum_{n=1}^{p+1} \boldsymbol{\varepsilon}_{n,\text{elec}}[\mathbf{v}_{1,n}^{\rho,E}, \dots, \mathbf{v}_{n,n}^{\rho,E}] + \boldsymbol{\varepsilon}_{n,\text{mag}}[\mathbf{v}_{1,n}^{\rho,H}, \dots, \mathbf{v}_{n,n}^{\rho,H}] \quad \rho = 0, 1, 2, \dots$$

Dipole ($n = 1$)



Quadrupole ($n = 2$)



$\boldsymbol{\varepsilon}_n[\mathbf{d}_1, \dots, \mathbf{d}_n]$ electric fields

- generated by charges **AND** currents
 - ▶ related by **charge conservation principle**
- electric/magnetic 2^n -point charges
 - ▶ electric multipole: $\boldsymbol{\varepsilon} = -\nabla \mathcal{V}_E + i\kappa \mathcal{A}_E$
 - ▶ magnetic multipole: $\boldsymbol{\varepsilon} = -\text{curl } \mathcal{A}_H$
- of moments ($\mathbf{d}_1, \dots, \mathbf{d}_n$)
- defined by an asymptotic process

$$\text{div } \mathcal{J} - i\omega \rho = 0$$

Far-field approximation

Far from the obstacle: **time-harmonic** approximation

$$\tilde{\mathbf{E}}_{p+3} = \sum_{n=1}^{p+1} \boldsymbol{\varepsilon}_{n,\text{elec}}[\mathbf{v}_{1,n}^{p,E}, \dots, \mathbf{v}_{n,n}^{p,E}] + \boldsymbol{\varepsilon}_{n,\text{mag}}[\mathbf{v}_{1,n}^{p,H}, \dots, \mathbf{v}_{n,n}^{p,H}]$$

$$\tilde{\mathbf{H}}_{p+3} = \sum_{n=1}^{p+1} \boldsymbol{\mathcal{H}}_{n,\text{mag}}[\mathbf{v}_{1,n}^{p,H}, \dots, \mathbf{v}_{n,n}^{p,H}] + \boldsymbol{\mathcal{H}}_{n,\text{elec}}[\mathbf{v}_{1,n}^{p,E}, \dots, \mathbf{v}_{n,n}^{p,E}]$$

	Order	Dipole		Quadrupole		Octupole	
		Electric	Magnetic	Electric	Magnetic	Electric	Magnetic
Far field	3	●	●	●	●		
	4	●	●	●	●		
	5	●	●	●	●		

● Electric field

● Magnetic field

Far-field approximation

Far from the obstacle: **time-harmonic** approximation

$$\tilde{\mathbf{E}}_{p+3} = \sum_{n=1}^{p+1} \boldsymbol{\varepsilon}_{n,\text{elec}}[\mathbf{v}_{1,n}^{p,E}, \dots, \mathbf{v}_{n,n}^{p,E}] + \boldsymbol{\varepsilon}_{n,\text{mag}}[\mathbf{v}_{1,n}^{p,H}, \dots, \mathbf{v}_{n,n}^{p,H}]$$

$$\tilde{\mathbf{H}}_{p+3} = \sum_{n=1}^{p+1} \boldsymbol{\mathcal{H}}_{n,\text{mag}}[\mathbf{v}_{1,n}^{p,H}, \dots, \mathbf{v}_{n,n}^{p,H}] + \boldsymbol{\mathcal{H}}_{n,\text{elec}}[\mathbf{v}_{1,n}^{p,E}, \dots, \mathbf{v}_{n,n}^{p,E}]$$

- Spherical** case

	Order	Dipole		Quadrupole		Octupole	
		Electric	Magnetic	Electric	Magnetic	Electric	Magnetic
Far field	3	•	•	•	•		
	4						
	5	•	•	•	•	•	•

- Collected dipolar** approximation

	Order	Dipole		Quadrupole		Octupole	
		Electric	Magnetic	Electric	Magnetic	Electric	Magnetic
Far field	3	•	•	•	•		
	4						
	5	•	•	•	•		

Born approximation

- The electromagnetic fields are approximated by the **superposition principle**

$$\mathbf{E}_\delta(\mathbf{x}) \approx \sum_{k=1}^{N_{\text{obs}}} \mathbf{E}_{\delta,k}(\mathbf{x}) \quad \mathbf{H}_\delta(\mathbf{x}) \approx \sum_{k=1}^{N_{\text{obs}}} \mathbf{H}_{\delta,k}(\mathbf{x})$$

- Each obstacle is modeled as a **dipolar source** around c_k

$$\mathbf{E}_{\delta,k}(\mathbf{x}) = \boldsymbol{\varepsilon}_{1,\text{elec}}[\mathbf{d}_{\delta,k}^{\text{E}}](\mathbf{x} - c_k) + \boldsymbol{\varepsilon}_{1,\text{mag}}[\mathbf{d}_{\delta,k}^{\text{H}}](\mathbf{x} - c_k)$$

For **perfectly conducting spheres**:

- ▶ Approximation of order 3

$$\mathbf{d}_{\delta,k}^{\text{E}} = 4\pi\delta^3 \mathbf{E}^{\text{inc}}(c_k) \quad \mathbf{d}_{\delta,k}^{\text{H}} = -2\pi\delta^3 \mathbf{H}^{\text{inc}}(c_k)$$

- ▶ Collected dipolar approximation

$$\mathbf{d}_{\delta,k}^{\text{E}} = 4\pi\delta^3 \left(1 + \frac{3(\kappa\delta)^2}{10}\right) \mathbf{E}^{\text{inc}}(c_k) \quad \mathbf{d}_{\delta,k}^{\text{H}} = -2\pi\delta^3 \left(1 - \frac{3(\kappa\delta)^2}{5}\right) \mathbf{H}^{\text{inc}}(c_k)$$

Foldy-Lax model

- The electromagnetic fields are approximated by the superposition principle

$$\mathbf{E}_\delta(\mathbf{x}) \approx \sum_{k=1}^{N_{\text{obs}}} \mathbf{E}_{\delta,k}(\mathbf{x}) \quad \mathbf{H}_\delta(\mathbf{x}) \approx \sum_{k=1}^{N_{\text{obs}}} \mathbf{H}_{\delta,k}(\mathbf{x})$$

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For **perfectly conducting** spheres:

- Approximation of order 3

$$\mathbf{d}_{\delta,k}^{\text{E}} = 4\pi\delta^3 \left(\mathbf{E}^{\text{inc}}(c_k) + \sum_{\substack{\ell=1 \\ \ell \neq k}}^{N_{\text{obs}}} \mathbf{E}_{\delta,\ell}(c_k) \right) \quad \mathbf{d}_{\delta,k}^{\text{H}} = -2\pi\delta^3 \left(\mathbf{H}^{\text{inc}}(c_k) + \sum_{\substack{\ell=1 \\ \ell \neq k}}^{N_{\text{obs}}} \mathbf{H}_{\delta,\ell}(c_k) \right)$$

- Collected dipolar approximation

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- Vectorial formulation** (3-order approximation)

Find $\mathbf{d} = ((\mathbf{d}_1^{\text{E}}), \dots, (\mathbf{d}_{N_{\text{obs}}}^{\text{E}}), (\mathbf{d}_1^{\text{H}}), \dots, (\mathbf{d}_{N_{\text{obs}}}^{\text{H}}))^\top \in \mathbb{C}^{6N_{\text{obs}}}$ such that

$$\mathbf{d} - \delta^3 \mathbb{A} \mathbf{d} = \delta^3 \mathbf{f}$$

$$\text{with } \mathbf{f} = \begin{pmatrix} 4\pi \mathbf{E}^{\text{inc}}(c_1) \\ \vdots \\ 4\pi \mathbf{E}^{\text{inc}}(c_{N_{\text{obs}}}) \\ -2\pi \mathbf{H}^{\text{inc}}(c_1) \\ \vdots \\ -2\pi \mathbf{H}^{\text{inc}}(c_{N_{\text{obs}}}) \end{pmatrix}$$

and \mathbb{A} the “interaction” matrix

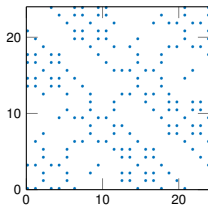


Figure: Skeleton of \mathbb{A} for $N_{\text{obs}} = 4$

Foldy-Lax approximations

- Approximation of order 3

$$\mathbf{d} - \delta^3 \mathbb{A} \mathbf{d} = \delta^3 \mathbf{f}$$

- Collected dipolar approximation

$$\mathbb{D}^{-1} \mathbf{d} - \delta^3 \mathbb{A} \mathbf{d} = \delta^3 \mathbf{f}$$

where

$$\mathbb{D} = \text{diag}(\underbrace{\alpha, \dots, \alpha}_{3N_{\text{obs}}}, \underbrace{\beta, \dots, \beta}_{3N_{\text{obs}}}) \quad \alpha = 1 + \frac{3(\kappa\delta)^2}{10} \quad \beta = 1 - \frac{6(\kappa\delta)^2}{10}$$

- Modified dipolar approximation

$$\tilde{\mathbb{D}}^{-1} \mathbf{d} - \delta^3 \mathbb{A} \mathbf{d} = \delta^3 \mathbf{f}$$

where

$$\tilde{\mathbb{D}} = \text{diag}(\underbrace{\tilde{\alpha}, \dots, \tilde{\alpha}}_{3N_{\text{obs}}}, \underbrace{\tilde{\beta}, \dots, \tilde{\beta}}_{3N_{\text{obs}}}) \quad \tilde{\alpha} = \frac{3i}{2(\kappa\delta)^3} \frac{j_1(\kappa\delta)}{h_1^{(1)}(\kappa\delta)} \quad \tilde{\beta} = -\frac{3i}{(\kappa\delta)^3} \frac{j_1(\kappa\delta) + \kappa\delta j_1'(\kappa\delta)}{h_1^{(1)}(\kappa\delta) + \kappa\delta h_1^{(1)'}(\kappa\delta)}$$

Foldy-Lax approximations

- Approximation of order 3

$$\mathbf{d} - \delta^3 \mathbb{A} \mathbf{d} = \delta^3 \mathbf{f}$$

- Collected dipolar approximation

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- **Modified** dipolar approximation

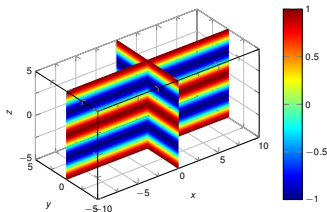
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where

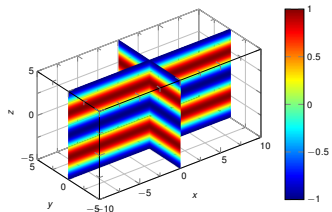
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Numerical validation: Spherical case

- Incident field: Electromagnetic plane wave of wavelength $\lambda = 5$ along z -axis polarized by \mathbf{e}_x

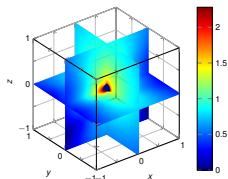


(a) Electric field (x -component)

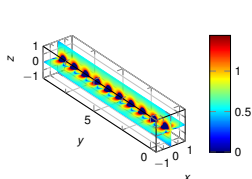


(b) Magnetic field (y -component)

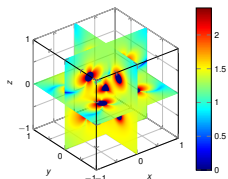
- Reference solution: "Spectral" solution truncated at the order 10



(c) One obstacle



(d) Aligned obstacles



(e) Real 3D-case

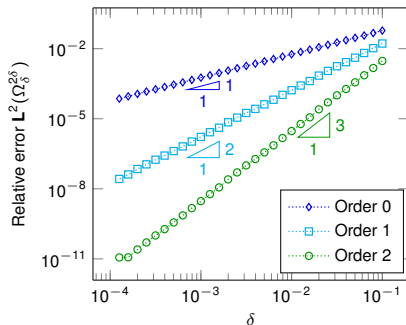
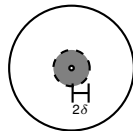
Single-scattering: Validation of asymptotic expansions

- **Near-field** approximations

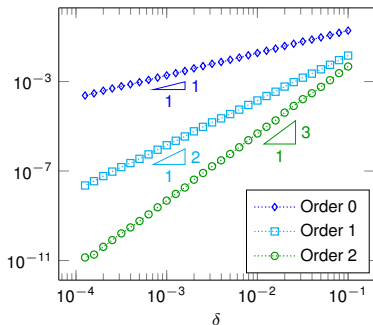
Order 0 : $\|\mathbf{E}_\delta - \widehat{\mathbf{E}}_0(\frac{\cdot}{\delta})\|_{\mathbf{L}^2(\Omega_\delta^{2\delta})} / \|\mathbf{E}_\delta\|_{\mathbf{L}^2(\Omega_\delta^{2\delta})}$

Order 1 : $\|\mathbf{E}_\delta - \widehat{\mathbf{E}}_0(\frac{\cdot}{\delta}) - \delta \widehat{\mathbf{E}}_1(\frac{\cdot}{\delta})\|_{\mathbf{L}^2(\Omega_\delta^{2\delta})} / \|\mathbf{E}_\delta\|_{\mathbf{L}^2(\Omega_\delta^{2\delta})}$

Order 2 : $\|\mathbf{E}_\delta - \widehat{\mathbf{E}}_0(\frac{\cdot}{\delta}) - \delta \widehat{\mathbf{E}}_1(\frac{\cdot}{\delta}) - \delta^2 \widehat{\mathbf{E}}_2(\frac{\cdot}{\delta})\|_{\mathbf{L}^2(\Omega_\delta^{2\delta})} / \|\mathbf{E}_\delta\|_{\mathbf{L}^2(\Omega_\delta^{2\delta})}$



(f) Electric field



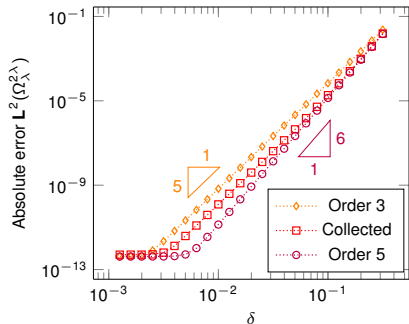
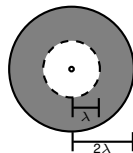
(g) Magnetic Field

Single-scattering: Validation of asymptotic expansions

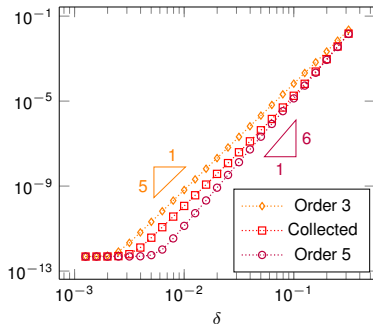
- Far-field approximations

Order 3 : $\|\mathbf{E}_\delta - \delta^3 \tilde{\mathbf{E}}_3\|_{\mathbf{L}^2(\Omega_\lambda^{2\lambda})}$

Order 5 : $\|\mathbf{E}_\delta - \delta^3 \tilde{\mathbf{E}}_3 - \delta^5 \tilde{\mathbf{E}}_5\|_{\mathbf{L}^2(\Omega_\lambda^{2\lambda})}$



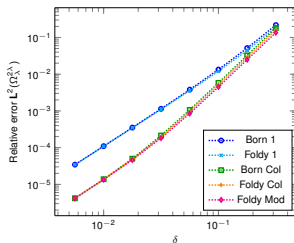
(l) Electric field



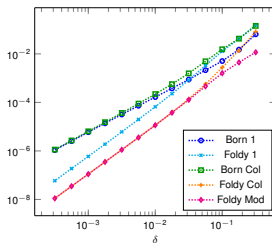
(m) Magnetic field

Multiple-scattering: Validation of Foldy-Lax model

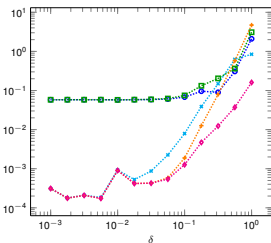
- Order 3: Born and Foldy
- Collected dipole improvement: Collected Born and Collected Foldy
- Spectral improvement: Modified Foldy



(n) Real 3D case with $N_{\text{obs}} = 13$ (fixed distance)



(o) Aligned obstacles with $N_{\text{obs}} = 5$ ($\sqrt{\delta}$ -dependence)



(p) Aligned obstacles with $N_{\text{obs}} = 5$ (δ -dependence)

Outline

1. Asymptotic models

- Single-scattering
- Application: Born approximation
- Multiple-scattering: Foldy-Lax model
- Numerical results

2. Spectral method: Spherical case

- Discretization
- Numerical convergence
- Comparison with asymptotic models
- Comparison with finite element solutions

3. Conclusions and perspectives

Spectral method

- The electromagnetic fields are represented by the **Stratton-Chu formula** for $\mathbf{x} \in \Omega_\delta$

$$\mathbf{E}_\delta(\mathbf{x}) = \sum_{k=1}^{N_{\text{obs}}} \mathbf{curl} \int_{\Gamma_\delta^k} \Phi(\mathbf{x}, \mathbf{y}) \mathbf{p}_k(\mathbf{y}) ds_{\mathbf{y}}$$

where $\Phi(\mathbf{x}, \mathbf{y}) = \frac{\exp(i\kappa|\mathbf{x}-\mathbf{y}|)}{4\pi|\mathbf{x}-\mathbf{y}|}$

Spectral method

- The electromagnetic fields are represented by the **Stratton-Chu formula** for $\mathbf{x} \in \Omega_\delta$

$$\mathbf{E}_\delta(\mathbf{x}) = \sum_{k=1}^{N_{\text{obs}}} \mathbf{curl} \int_{\Gamma_\delta^k} \Phi(\mathbf{x}, \mathbf{y}) \mathbf{p}_k(\mathbf{y}) ds_{\mathbf{y}}$$

- Each $\mathbf{p}_k \in \mathbf{H}_t^{-\frac{1}{2}}(\text{div}_{\Gamma_\delta^k}, \Gamma_\delta^k)$ satisfies the **boundary integral equation**

$$\sum_{j=1}^{N_{\text{obs}}} \mathcal{M}_\Gamma^{kj} \mathbf{p}_j = -\mathbf{n} \times \mathbf{E}^{\text{inc}} \quad \text{on } \Gamma_\delta^k$$

where $\mathcal{M}_\Gamma^{kj} : \mathbf{H}_t^{-\frac{1}{2}}(\text{div}_{\Gamma_\delta^j}, \Gamma_\delta^j) \rightarrow \mathbf{H}_t^{-\frac{1}{2}}(\text{div}_{\Gamma_\delta^k}, \Gamma_\delta^k)$ is the extension of

$$\mathcal{M}_\Gamma^{kj} \boldsymbol{\lambda}(\mathbf{x}_\Gamma) = \mathbf{n}(\mathbf{x}_\Gamma) \times \lim_{\mathbf{x} \rightarrow \mathbf{x}_\Gamma} \mathbf{curl} \int_{\Gamma_\delta^j} \Phi(\mathbf{x}, \mathbf{y}) \boldsymbol{\lambda}(\mathbf{y}) ds_{\mathbf{y}} \quad \boldsymbol{\lambda} \in \mathcal{C}^\infty(\Gamma_\delta^j) \quad \mathbf{x}_\Gamma \in \Gamma_\delta^k$$

Spectral method

- The electromagnetic fields are represented by the **Stratton-Chu formula** for $\mathbf{x} \in \Omega_\delta$

$$\mathbf{E}_\delta(\mathbf{x}) = \sum_{k=1}^{N_{\text{obs}}} \mathbf{curl} \int_{\Gamma_\delta^k} \Phi(\mathbf{x}, \mathbf{y}) \mathbf{p}_k(\mathbf{y}) ds_{\mathbf{y}}$$

- Each $\mathbf{p}_k \in \mathbf{H}_t^{-\frac{1}{2}}(\text{div}_{\Gamma_\delta^k}, \Gamma_\delta^k)$ satisfies the **boundary integral equation**

$$\sum_{j=1}^{N_{\text{obs}}} \mathcal{M}_{\Gamma_\delta^k}^{kj} \mathbf{p}_j = -\mathbf{n} \times \mathbf{E}^{\text{inc}} \quad \text{on } \Gamma_\delta^k$$

- Galerkin discretization** of the BIE on **local spectral basis** with N_{mod} modes

$$\mathbf{p}_j(\mathbf{x}) = \sum_{n=1}^{N_{\text{mod}}} \sum_{m=-n}^n \mathbf{p}_{n,m}^{j,\perp} \nabla_{S^2} Y_{n,m}(\hat{\mathbf{x}}_j) + \mathbf{p}_{n,m}^{j,\times} \mathbf{curl}_{S^2} Y_{n,m}(\hat{\mathbf{x}}_j) \quad \mathbf{x} \in \Gamma_\delta^j$$

with $\hat{\mathbf{x}}_j = \frac{\mathbf{x} - \mathbf{c}_j}{|\mathbf{x} - \mathbf{c}_j|}$ and $\nabla_{S^2} Y_{n,m}$, $\mathbf{curl}_{S^2} Y_{n,m}$: complex-valued vector spherical harmonics

Vectorial formulation

- **Variational** formulation: Find $\mathbf{p}_k \in \mathbf{H}_t^{-\frac{1}{2}}(\text{div}_{\Gamma_\delta^k}, \Gamma_\delta^k)$ such that

$$\sum_{j=1}^{N_{\text{obs}}} \langle \mathcal{M}_{\Gamma}^{kj} \mathbf{p}_j, \mathbf{v}_k \rangle_{\Gamma_\delta^k} = -\langle \mathbf{n} \times \mathbf{E}^{\text{inc}}, \mathbf{v}_k \rangle_{\Gamma_\delta^k} \quad \forall \mathbf{v}_k \in \mathbf{H}_t^{-\frac{1}{2}}(\text{curl}_{\Gamma_\delta^k}, \Gamma_\delta^k)$$

- **Vectorial** formulation: Find $\mathbf{p} = ((\rho_{n,m}^{1,\perp}), \dots, (\rho_{n,m}^{N_{\text{obs}},\perp}), (\rho_{n,m}^{1,\times}), \dots, (\rho_{n,m}^{N_{\text{obs}},\times}))^\top \in \mathbb{C}^{\mathbf{N}}$ s.t.

$$\mathbf{M} \mathbf{p} = \mathbf{f}$$

with $\mathbf{N} = 2N_{\text{mod}}(N_{\text{mod}} + 2)N_{\text{obs}}$ and

$$\mathbf{M} = \begin{pmatrix} \mathbf{M}_{\perp\perp} & \mathbf{M}_{\perp\times} \\ \mathbf{M}_{\times\perp} & \mathbf{M}_{\times\times} \end{pmatrix} \quad \text{with} \quad \mathbf{M}_{\alpha\beta} = \begin{pmatrix} \mathbf{M}_{\alpha\beta}^{11} & \mathbf{M}_{\alpha\beta}^{12} & \dots & \mathbf{M}_{\alpha\beta}^{1N_{\text{obs}}} \\ \mathbf{M}_{\alpha\beta}^{21} & \mathbf{M}_{\alpha\beta}^{22} & \dots & \mathbf{M}_{\alpha\beta}^{2N_{\text{obs}}} \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{M}_{\alpha\beta}^{N_{\text{obs}}1} & \mathbf{M}_{\alpha\beta}^{N_{\text{obs}}2} & \dots & \mathbf{M}_{\alpha\beta}^{N_{\text{obs}}N_{\text{obs}}} \end{pmatrix}$$

$$= \mathbf{M}_{\alpha\beta}(\delta, \mathbf{d}_{jk})$$

Vectorial formulation

- Variational formulation: Find $\mathbf{p}_k \in \mathbf{H}_t^{-\frac{1}{2}}(\text{div}_{\Gamma_\delta^k}, \Gamma_\delta^k)$ such that

$$\sum_{j=1}^{N_{\text{obs}}} \langle \mathcal{M}_{\Gamma}^{kj} \mathbf{p}_j, \mathbf{v}_k \rangle_{\Gamma_\delta^k} = -\langle \mathbf{n} \times \mathbf{E}^{\text{inc}}, \mathbf{v}_k \rangle_{\Gamma_\delta^k} \quad \forall \mathbf{v}_k \in \mathbf{H}_t^{-\frac{1}{2}}(\text{curl}_{\Gamma_\delta^k}, \Gamma_\delta^k)$$

- Vectorial formulation: Find $\mathbf{p} = ((\rho_{n,m}^{1,\perp}), \dots, (\rho_{n,m}^{N_{\text{obs}},\perp}), (\rho_{n,m}^{1,\times}), \dots, (\rho_{n,m}^{N_{\text{obs}},\times}))^\top \in \mathbb{C}^{\mathbf{N}}$ s.t.

$$\mathbb{M} \mathbf{p} = \mathbf{f}$$



Computation time
Ill-conditioned system
Memory consumption

- Numerical integration (Gauss-Lobatto)
- Large number of unknowns
- Dense matrix



Mex interface (C++)
Preconditioning (system & matrix)
Smart storage and assembling

- Code speed-up
- Linear algebra tools
- Sub-blocks $\mathbb{M}_{\alpha\beta}^{kj}$ depend on \mathbf{d}_{jk}

Preconditioning

Linear system: **Change of variable** + **Permutation** + **Normalization**

$$\mathbf{E}_\delta = \phi_\delta(\mathbf{p}_\delta) \quad \text{with } \mathbb{M} \mathbf{p}_\delta = \mathbf{f}$$



$$\mathbf{E}_\delta = \phi(\tilde{\mathbf{p}}_\delta) \quad \text{with } \tilde{\mathbb{M}} \tilde{\mathbf{p}}_\delta = \tilde{\mathbf{f}}$$



Dense matrix $\tilde{\mathbb{M}}$



Error of approximation
Analytic preconditionner (e.g. dipole)
Algebraic preconditionner (small criterion)

Preconditioning

Linear system: Change of variable + Permutation + Normalization

$$\mathbf{E}_\delta = \phi_\delta(\mathbf{p}_\delta) \quad \text{with } \mathbb{M} \mathbf{p}_\delta = \mathbf{f}$$



$$\mathbf{E}_\delta = \phi(\tilde{\mathbf{p}}_\delta) \quad \text{with } \tilde{\mathbb{M}} \tilde{\mathbf{p}}_\delta = \tilde{\mathbf{f}}$$



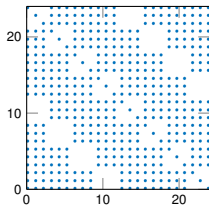
Dense matrix $\tilde{\mathbb{M}}$



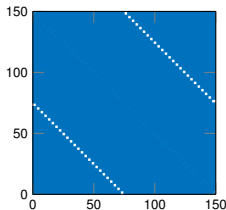
Error of approximation

Analytic preconditionner (e.g. dipole)

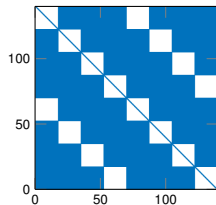
Algebraic preconditionner (small criterion)



(a) $N_{\text{obs}} = 4$ and $N_{\text{mod}} = 1$



(b) $N_{\text{obs}} = 25$ and $N_{\text{mod}} = 1$



(c) $N_{\text{obs}} = 4$ and $N_{\text{mod}} = 5$

Figure: Skeleton of $\tilde{\mathbb{M}}$

Preconditioning

Linear system: Change of variable + Permutation + Normalization

$$\mathbf{E}_\delta = \phi_\delta(\mathbf{p}_\delta) \quad \text{with } \mathbb{M} \mathbf{p}_\delta = \mathbf{f}$$



$$\mathbf{E}_\delta = \phi(\tilde{\mathbf{p}}_\delta) \quad \text{with } \tilde{\mathbb{M}} \tilde{\mathbf{p}}_\delta = \tilde{\mathbf{f}}$$



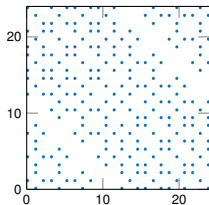
Dense matrix $\tilde{\mathbb{M}}$



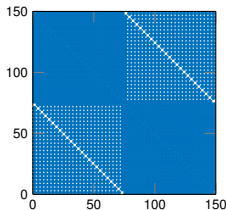
Error of approximation

Analytic preconditionner (e.g. dipole)

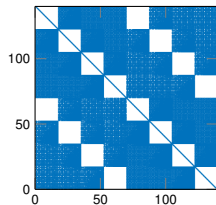
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Figure: Skeleton of $\tilde{\mathbb{M}}$

Preconditioning

Linear system: Change of variable + Permutation + Normalization

$$\mathbf{E}_\delta = \phi_\delta(\mathbf{p}_\delta) \quad \text{with } \mathbb{M} \mathbf{p}_\delta = \mathbf{f}$$



$$\mathbf{E}_\delta = \phi(\tilde{\mathbf{p}}_\delta) \quad \text{with } \tilde{\mathbb{M}} \tilde{\mathbf{p}}_\delta = \tilde{\mathbf{f}}$$



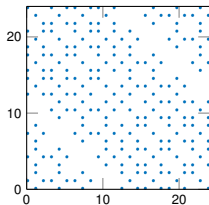
Dense matrix $\tilde{\mathbb{M}}$



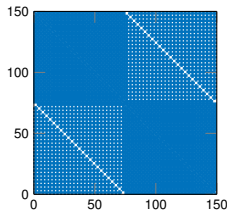
Error of approximation

Analytic preconditionner (e.g. dipole)

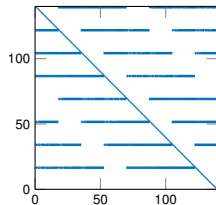
Algebraic preconditionner (small criterion)



(a) $N_{\text{obs}} = 4$ and $N_{\text{mod}} = 1$



(b) $N_{\text{obs}} = 25$ and $N_{\text{mod}} = 1$



(c) $N_{\text{obs}} = 4$ and $N_{\text{mod}} = 5$

Figure: Skeleton of $\text{Precond}(\tilde{\mathbb{M}})$

Preconditioning

Linear system: Change of variable + Permutation + Normalization

$$\mathbf{E}_\delta = \phi_\delta(\mathbf{p}_\delta) \quad \text{with } \mathbb{M} \mathbf{p}_\delta = \mathbf{f}$$



$$\mathbf{E}_\delta = \phi(\tilde{\mathbf{p}}_\delta) \quad \text{with } \tilde{\mathbb{M}} \tilde{\mathbf{p}}_\delta = \tilde{\mathbf{f}}$$



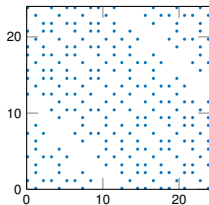
Dense matrix $\tilde{\mathbb{M}}$



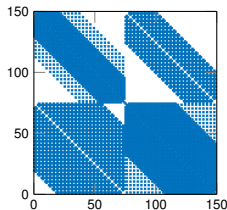
Error of approximation

Analytic preconditionner (e.g. dipole)

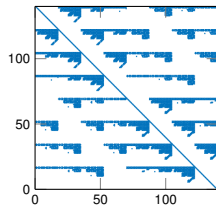
Algebraic preconditionner (small criterion)



(a) $N_{\text{obs}} = 4$ and $N_{\text{mod}} = 1$



(b) $N_{\text{obs}} = 25$ and $N_{\text{mod}} = 1$



(c) $N_{\text{obs}} = 4$ and $N_{\text{mod}} = 5$

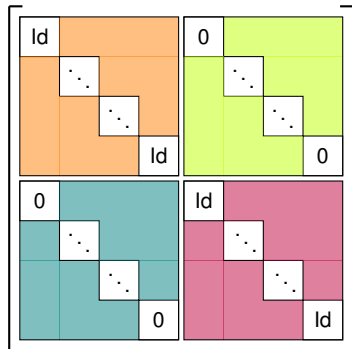
Figure: Skeleton of $\text{Precond}(\tilde{\mathbb{M}})$

Smart storage and assembling

- Example: 4 aligned obstacles



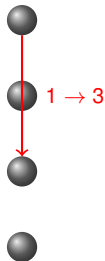
$$\tilde{\mathbf{M}} =$$



The sub-blocks $M_{\alpha\beta}^{kj}$ depend only on δ and \mathbf{d}_{jk}

Smart storage and assembling

- Example: 4 aligned obstacles



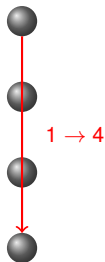
$$\tilde{\mathbf{M}} = \begin{bmatrix} \text{Id} & \begin{matrix} 1 & -2 & 1 & -3 \\ \vdots & & & \end{matrix} & \begin{matrix} 0 & & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & 0 \end{matrix} & \begin{matrix} 0 & & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & 0 \end{matrix} \\ \begin{matrix} 0 & \text{Id} & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & 0 \end{matrix} & \begin{matrix} \text{Id} & & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & \text{Id} \end{matrix} \end{bmatrix}$$

- Storage of sub-blocks

$$\tilde{\mathbf{M}}_{\text{Block}} = \begin{bmatrix} \begin{matrix} 1 & 2 & 1 & 3 \\ \vdots & & & \end{matrix} & \begin{matrix} \text{Id} & & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & 0 \end{matrix} & \begin{matrix} 0 & & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & 0 \end{matrix} \\ \begin{matrix} 0 & \text{Id} & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & 0 \end{matrix} & \begin{matrix} \text{Id} & & & \\ & \text{Id} & & \\ & & \text{Id} & \\ & & & \text{Id} \end{matrix} \end{bmatrix}$$

Smart storage and assembling

- Example: 4 aligned obstacles



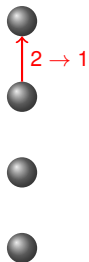
$$\tilde{\mathbf{M}} = \begin{bmatrix} \text{Id} & \begin{matrix} 1 & -2 & 1 & -3 & 1 & -4 \end{matrix} & 0 & \begin{matrix} \text{light green} & \text{green} & \text{dark green} \end{matrix} \\ & \ddots & & \\ & & \ddots & \\ & & & \text{Id} & & 0 \\ 0 & \begin{matrix} \text{teal} & \text{dark teal} & \text{blue} \end{matrix} & \text{Id} & \begin{matrix} \text{pink} & \text{red} & \text{purple} \end{matrix} \\ & \ddots & & \ddots & \\ & & \ddots & & \ddots & \\ & & & 0 & & \text{Id} \end{bmatrix}$$

- Storage of sub-blocks

$$\tilde{\mathbf{M}}_{\text{Block}} = \begin{bmatrix} \begin{matrix} 1 & 2 & 1 & 3 & 1 & 4 \end{matrix} & & & \begin{matrix} \text{light green} & \text{green} & \text{dark green} \end{matrix} & & & \begin{matrix} \text{teal} & \text{dark teal} & \text{blue} \end{matrix} & & & \begin{matrix} \text{pink} & \text{red} & \text{purple} \end{matrix} & & & \end{bmatrix}$$

Smart storage and assembling

- Example: 4 aligned obstacles



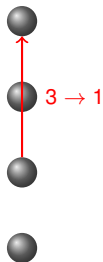
$$\tilde{\mathbf{M}} = \begin{bmatrix} \text{Id} & \begin{matrix} 1 & - & 2 & 1 & - & 3 & 1 & - & 4 \end{matrix} & 0 & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{green} \end{matrix} \\ \begin{matrix} 2 & - & 1 \end{matrix} & \begin{matrix} \text{orange} & \text{orange} & \text{orange} & \text{orange} \end{matrix} & \begin{matrix} \text{green} & \text{green} & \text{green} & \text{green} \end{matrix} & \begin{matrix} \text{yellow} & \text{yellow} & \text{yellow} & \text{yellow} \end{matrix} \\ & \begin{matrix} \text{pink} & \text{pink} & \text{pink} & \text{pink} \end{matrix} & \begin{matrix} \text{green} & \text{green} & \text{green} & \text{green} \end{matrix} & \begin{matrix} \text{yellow} & \text{yellow} & \text{yellow} & \text{yellow} \end{matrix} \\ & & \text{Id} & 0 \end{bmatrix}$$

- Storage of sub-blocks

$$\tilde{\mathbf{M}}_{\text{Block}} = \begin{bmatrix} \begin{matrix} 1 & 2 & 1 & 3 & 1 & 4 & 2 & 1 \end{matrix} & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{green} \end{matrix} & \begin{matrix} \text{teal} & \text{teal} & \text{blue} & \text{light blue} \end{matrix} & \begin{matrix} \text{pink} & \text{red} & \text{purple} & \text{purple} \end{matrix} \end{bmatrix}$$

Smart storage and assembling

- Example: 4 aligned obstacles



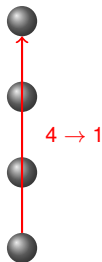
$$\tilde{\mathbf{M}} = \begin{bmatrix} \text{Id} & \begin{matrix} 1 & - & 2 & 1 & - & 3 & 1 & - & 4 \end{matrix} & 0 & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{yellow} & \text{green} & \text{yellow} & \text{green} & \text{yellow} \end{matrix} \\ \begin{matrix} 2 & - & 1 \end{matrix} & \begin{matrix} \text{orange} & \text{orange} & \text{orange} & \text{orange} \end{matrix} & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{yellow} \end{matrix} & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{yellow} \end{matrix} \\ \begin{matrix} 3 & - & 1 \end{matrix} & \begin{matrix} \text{orange} & \text{orange} & \text{orange} & \text{orange} \end{matrix} & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{yellow} \end{matrix} & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{yellow} \end{matrix} \\ & \text{Id} & 0 & \begin{matrix} \text{green} & \text{yellow} & \text{green} & \text{yellow} \end{matrix} \\ 0 & \begin{matrix} \text{teal} & \text{teal} & \text{teal} & \text{blue} \end{matrix} & \text{Id} & \begin{matrix} \text{pink} & \text{red} & \text{purple} & \text{pink} \end{matrix} \\ \begin{matrix} \text{light blue} & \text{light blue} & \text{light blue} & \text{light blue} \end{matrix} & \begin{matrix} \text{teal} & \text{teal} & \text{teal} & \text{blue} \end{matrix} & \begin{matrix} \text{pink} & \text{red} & \text{purple} & \text{pink} \end{matrix} & \begin{matrix} \text{pink} & \text{red} & \text{purple} & \text{pink} \end{matrix} \\ \begin{matrix} \text{light blue} & \text{light blue} & \text{light blue} & \text{light blue} \end{matrix} & \begin{matrix} \text{teal} & \text{teal} & \text{teal} & \text{blue} \end{matrix} & \begin{matrix} \text{pink} & \text{red} & \text{purple} & \text{pink} \end{matrix} & \begin{matrix} \text{pink} & \text{red} & \text{purple} & \text{pink} \end{matrix} \\ & 0 & \text{Id} & \begin{matrix} \text{pink} & \text{red} & \text{purple} & \text{pink} \end{matrix} \end{bmatrix}$$

- Storage of sub-blocks

$$\tilde{\mathbf{M}}_{\text{Block}} = \begin{bmatrix} \text{orange} & \text{orange} & \text{red} & \text{pink} & \text{pink} & \text{white} & \text{green} & \text{yellow} & \text{green} & \text{yellow} & \text{green} & \text{yellow} & \text{white} & \text{teal} & \text{teal} & \text{blue} & \text{light blue} & \text{light blue} & \text{white} & \text{pink} & \text{red} & \text{purple} & \text{pink} & \text{pink} & \text{white} \end{bmatrix}$$

Smart storage and assembling

- Example: 4 aligned obstacles



$$\tilde{\mathbf{M}} =$$

Id	1 - 2	1 - 3	1 - 4	0			
2 - 1
3 - 1
4 - 1	.	.	.	Id			0
0				Id			

	.	.	.	0			Id

- Storage of sub-blocks



Smart storage and assembling

- Example: 4 aligned obstacles



$$\tilde{\mathbf{M}} =$$

Id	1 - 2	1 - 3	1 - 4	0			
2 - 1
3 - 1
4 - 1	.	.	.	Id			0
0				Id			

	.	.	.	0			Id

- Storage of sub-blocks

$$\tilde{\mathbf{M}}_{\text{Block}} =$$

1 2	1 3	1 4	2 1	3 1	4 1															
-----	-----	-----	-----	-----	-----	--	--	--	--	--	--	--	--	--	--	--	--	--	--	--

- Iterative solvers: define the **action** of $\tilde{\mathbf{M}}_{\text{Block}}$ on \mathbf{u}

$$\mathcal{A}(\tilde{\mathbf{M}}_{\text{Block}}, \mathbf{u}) := \tilde{\mathbf{M}} \mathbf{u}$$

Smart storage and assembling

- Example: 4 aligned obstacles



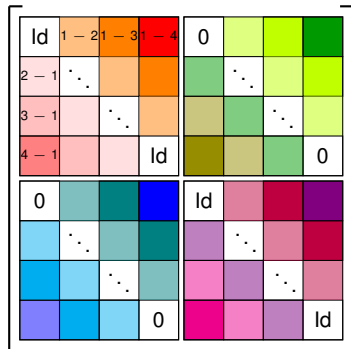
$8(N_{\text{obs}} - 1)$ blocks



instead of $4(N_{\text{obs}})^2$



$$\tilde{\mathbf{M}} =$$



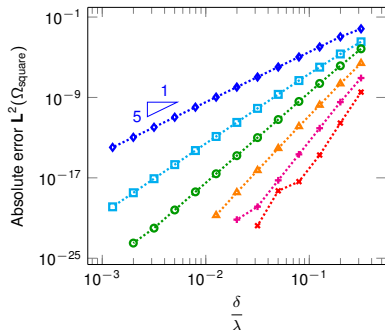
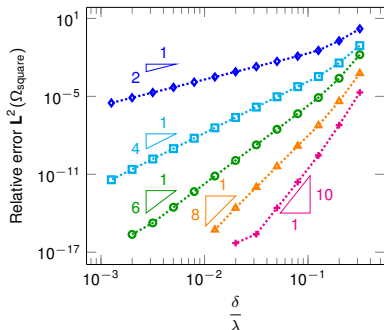
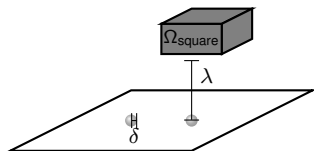
- Storage of sub-blocks



N_{obs}	$4(N_{\text{obs}})^2$	$8(N_{\text{obs}} - 1)$	Ratio (%)	$\tilde{\mathbf{M}}$ (GB)	$\tilde{\mathbf{M}}_{\text{Block}}$ (GB)
100	40 000	792	1.98	$5.76 \cdot 10^{-3}$	$1.15 \cdot 10^{-4}$
1 000	4 000 000	7 992	$1.998 \cdot 10^{-3}$	$5.76 \cdot 10^{-1}$	$1.15 \cdot 10^{-3}$
3 000	36 000 000	23 992	$6.664 \cdot 10^{-4}$	4.01	$3.45 \cdot 10^{-3}$
10 000	400 000 000	79 992	$1.9998 \cdot 10^{-4}$	--	$1.15 \cdot 10^{-2}$

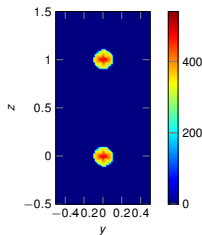
Numerical validation of spectral solutions

- Incident field: Electromagnetic plane wave along z-axis polarized by \mathbf{e}_x
- Reference solution: Spectral solution with $N_{\text{mod}} = 10$

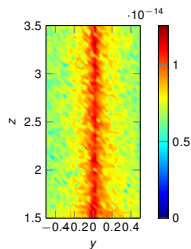


Comparison with asymptotic models

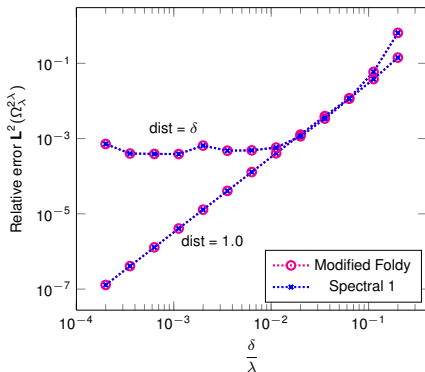
Difference between Spectral 1 and modified Foldy



(c) Close to the obstacles



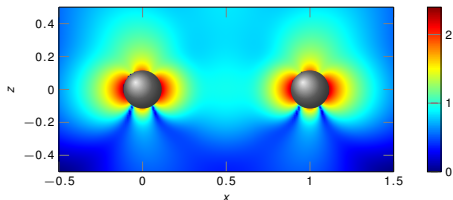
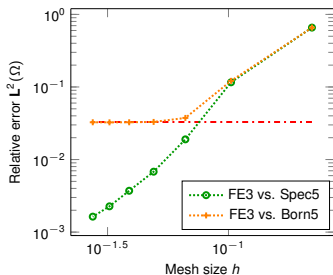
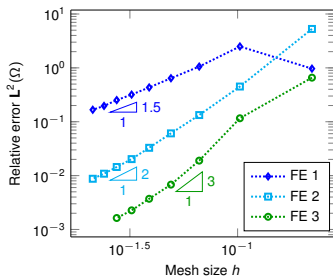
(d) In the far-field domain



	Modified Foldy				Spectral 1			
N_{obs}	100	1000	5000	10000	100	1000	5000	10000
Linear system	0.17	17.71	34.01	43.49	2.66	40.74	352.47	--
Post-processing	5.25	47.84	254.34	459.90	12.17	126.81	> 3600	--

Table: Time comparison in seconds

Comparison with finite element solutions (Montjoie)



Spectral 3			
N_{obs}	1	2	4
Size of linear system	30	60	120
L^2 -Accuracy ($\times 10^{-6}$)	4.77	4.48	3.66
Time (s)	15.46	65.55	262.33
Memory (GB)	0.3	2.36	9.3

Finite Element 3			
N_{obs}	1	2	4
Size of linear system	925 422	1 972 800	3 942 936
L^2 -Accuracy ($\times 10^{-3}$)	0.28	2.27	3.31
Time (min)	26.5	54.25	147.25
Memory (GB)	21.11	36.97	75

Table: Computational costs

Outline

1. Asymptotic models

- Single-scattering
- Application: Born approximation
- Multiple-scattering: Foldy-Lax model
- Numerical results

2. Spectral method: Spherical case

- Discretization
- Numerical convergence
- Comparison with asymptotic models
- Comparison with finite element solutions

3. Conclusions and perspectives

Conclusions and Perspectives

Conclusion

- ✓ Derivation and numerical validation of the matched asymptotic expansions
- ✓ Extension to Born and Foldy-Lax models
- ✓ Derivation and implementation at any order spectral method + Numerical validation

On-going work

- ~ Comparison of preconditionners and iterative solvers
- ~ Smart assembling for particular configurations (plane, cubic volume)
- ~ Contribution to mathematical justification of the asymptotic expansions

Perspectives

- ✗ Existence of electromagnetic centers?
- ✗ Extension to obstacles of arbitrary shape
- ✗ Extension to time-dependent domain

Thank you for your attention