# A Quasi-Trefftz Method for the Iterative Solution of Time-Harmonic Wave Problems based on the Flux Reconstruction Method

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# Context of the work

 Increasing need of 3D simulations of time-harmonic electromagnetic waves: high-frequency, heterogeneous environments, wide domains (in terms of wavelengths)...



Figure: FDTD simulation on a Manhattan mesh of  $700 \times 600 \times 300$ m, Thibault Volpert (DEMR ONERA)

Example of computation challenge: field radiated by two antennas in Manhattan

- Large simulation domain.
- Wide range of frequencies (1-18 GHz).

 $\implies$  Need of efficient methods to solve such ambitious cases!

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# Plan of the talk

1. 3D Maxwell: limitations and (quasi-)Trefftz method

2. Quasi-Trefftz numerical analysis

3. Information transfer optimisation

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# The 3D Maxwell problem in wide domains: current limitations and (quasi-)Trefftz method interests

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### Current issues: memory and iterative resolution

Time-harmonic Maxwell equations for the electromagnetic field  $\mathbf{Y} := (\mathbf{E}, \mathbf{H}) \in [\mathrm{H}(\mathrm{curl}, \Omega)]^6$ :

$$i\kappa \mathbf{M}\mathbf{Y} + \sum_{j=1}^{3} \frac{\partial \mathbf{F}^{j}\mathbf{Y}}{\partial x_{j}} = \mathbf{0} \text{ in } \Omega,$$
 (1)

with impedance Boundary Conditions (BCs)

$$(\mathbf{n}_{\partial\Omega} \times \mathbf{E}) \times \mathbf{n}_{\partial\Omega} + Z_{\partial\Omega} \ \mathbf{n}_{\partial\Omega} \times \mathbf{H} = \mathbf{g} \text{ on } \partial\Omega.$$
(2)

# Current issues: memory and iterative resolution

Many classic methods as Finite Differences [Yee 1966], Finite Elements [Nédélec 1980], Discontinuous Galerkin [Fezoui et al. 2005], but limitations for current applications:

- Wide domains lead to very large linear systems.
- Traditional direct solvers induce a prohibitive memory consumption.
- Usual numerical methods are not well adapted to iterative resolution.

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Figure: Memory consumption of a FEM w.r.t. the domain size (in wavelengths) [Sirdey 2022].

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# Considered numerical method: the Trefftz approach

- 1. Belongs to the Discontinuous Galerkin methods, with the Galerkin space  $\mathbb{X} := \prod_{T \in \mathfrak{T}_h} \mathbb{X}_T$  made up of local solutions (*i.e.* in each mesh cell  $T \in \mathfrak{T}_h$ ) of the Maxwell equations.
- 2. Use of the reciprocity formula verified in each cell

$$\forall T \in \mathfrak{I}_h, \ \int_{\partial T} \gamma_{\times}^T \mathbf{H}^T \cdot \overline{\gamma_t \mathbf{E}'^T} + \gamma_t \mathbf{E}^T \cdot \overline{\gamma_{\times}^T \mathbf{H}'^T} = 0,$$
(3)

in addition to the introduction of numerical traces [Sirdey 2022]:

Find  $\mathbf{Y}=(\mathbf{E},\mathbf{H})\in\mathbb{X}$  such that  $\forall~\mathbf{Y}'=(\mathbf{E}',\mathbf{H}')\in\mathbb{X}$ ,

$$\sum_{T \in \mathfrak{T}_h} \int_{\partial T} \widehat{\gamma_{\times}^T \mathbf{H}^T} \cdot \overline{\gamma_t \mathbf{E}'^T} + \widehat{\gamma_t \mathbf{E}^T} \cdot \overline{\gamma_{\times}^T \mathbf{H}'^T} = \ell(\mathbb{E}'),$$

where  $\hat{\cdot}$  stands for upwind (or Riemann) numerical traces and with

$$\gamma_t \mathbf{u}^T = \mathbf{u}^T - \left(\mathbf{u}^T \cdot \mathbf{n}_T\right) \mathbf{n}_T \text{ and } \gamma_{\times}^T \mathbf{u}^T = \mathbf{n}_T \times \mathbf{u}^T.$$

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# The Trefftz approach: properties and classic limits

- 1. The formulation is posed on the mesh skeleton and adaptable to an iterative resolution (contraction property) [Cessenat, Després 1998].
- 2. A relevant choice of the numerical traces naturally implies formulation coercivity.
- 3. Basis functions have a **physical meaning**: leads to a reduced numerical pollution [Ihlenburg, Babuška 1995].

Limits of the classic Plane Waves (PWs) choice:

- Numerical dependence phenomena: ill-conditioned basis [Congreve et al. 2019].
- No adaptability of the basis to the local properties: expected singularities, complex interference phenomena...



 $\implies$  Possible algebraic adaptations [Barucq et al. 2021].

 $\implies$  Introduction of well-conditioned basis by parameterising the space of local solutions by polynomial boundary conditions.

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### A Quasi-Trefftz approach

### Definition of the local problem

A local solution  $\mathbf{Y}^T$  in T is parameterised by a tangential field  $g^T \in \mathbb{V}^T$  such that

$$\gamma_t \mathbf{E}^T + Z_{\partial T} \ \gamma_{\times}^T \mathbf{H}^T = \mathbf{g}^T \text{ on } \partial T.$$
(4)

 $\implies$  Consider a finite-dimensional subspace  $\mathbb{V}_{\mathbf{h}}^T \subset \mathbb{V}^T$  and the associated Maxwell solutions as local basis [Fure et al. 2020].

#### Construction of $\mathbb{V}_h$ :

- Consider a cell  $T \in \mathfrak{T}_{\mathbf{b}}^T$
- Define a mesh of  $\mathfrak{T}(\partial T)$
- Consider a piecewise polynomial g<sup>T</sup> of degree k<sub>OT</sub> on T(∂T).



Figure: 2D example of the mesh of  $\partial T$  for  $k_{QT} = 1$ .

⇒ Associated Maxwell solutions are unknown: need of an auxiliary solver to compute approximations!

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# Description of the FR local solver

Choice of the local solver: the Flux Reconstruction method [Huynh 2007]

- Based on the first-order strong formulation (solves both  ${\bf E}$  and  ${\bf H}):$  no dependence on a quadrature rule.
- Idea: copy the strong equation in each cell of the mesh, which is solved on piecewise polynomials: the flux is corrected at interfaces thanks to numerical traces and correction polynomial functions.

$$i\kappa\mathbf{y} + \frac{\mathrm{d}\phi}{\mathrm{d}x} = \mathbf{0} \text{ with } \phi = \mathcal{F}(\mathbf{y}) \implies \forall \ n \in [\![1,N]\!], \ i\kappa\mathbf{y}_h + \frac{\mathrm{d}\phi_h}{\mathrm{d}x} = \mathbf{0} \text{ in } [X_{n-1}, X_n].$$

- Natural high-order method and adaptable to unstructured meshes.
- Choice of correction polynomials allows to retrieve usual methods: nodal DG (for Radau polynomials), Spectral Differences (for Lagrange polynomials)...
- Choice of the correction polynomials to the user: possibility to optimise them for wave propagation problems, especially in pre-asymptotic regime [Rivet, Pernet, Tordeux 2024].
- For identical cells, the inversion has to be realised only once!
- $\implies$  Any numerical method, solving the first-order system, may be used!

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# A quasi-Trefftz approach based on a Flux Reconstruction auxiliary solver: numerical analysis

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## Summary of this Quasi-Trefftz approach

Variational formulation posed on  $\mathbb{X} := \prod_{T \in \mathcal{T}_{h}} \mathbb{X}_{T}$  made up of local solutions.

Functions  $(\mathbf{E}^T, \mathbf{H}^T)$  of  $\mathbb{X}_T$  parameterised by  $\mathbf{g}^T \in \mathbb{V}^T$  such that:

$$\gamma_t \mathbf{E}^T + Z_{\partial T} \ \gamma_{\times}^T \mathbf{H}^T = \mathbf{g}^T \text{ on } \partial T.$$
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### ✦

**Polynomial approximation** of the trace space:  $\mathbb{V}_{\mathbf{h}}^{T} = \operatorname{span}(\mathbf{g}_{i}^{T})$ . For all *i*, the associated solutions  $(\mathbb{E}_{i}^{T}, \mathbf{H}_{i}^{T})$  are taken as basis functions of  $\mathbb{X}_{T}$ .

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**FR** auxiliary solver to compute an approximation of  $(\mathbf{E}_i^T, \mathbf{H}_i^T)$ .

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Variational formulation posed on  $\mathbb{X} := \prod_{T \in \mathfrak{T}_b} \mathbb{X}_T$  made up of local solutions.

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### Numerical experiments: basis quality



Figure: Eigenvalues of the mass matrix for 196 PWs (and reduction for  $\epsilon=10^{-6})$  and 192 Quasi-Trefftz basis functions.

#### **Basis quality**

- $\implies$  Possible reduction techniques to restore the condition number.
- → Avoids spurious numerical modes and ill-conditioning of the mass matrix!

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# Approximation properties of the basis for a smooth solution: sum of random PWs



#### $\implies$ Saturation phenomenon due to condition number.

⇒ **Reduction techniques** avoid conditioning issues, but locked convergence because of rounding pollution error for an asked threshold.

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⇒ Robust approximation by FR basis.

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Approximation properties of the basis for a 'non-smooth' solution: sum of random dipoles



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### Recovery of the Trefftz properties under local solver convergence

Trefftz approach rests on the use of exact solutions of the Maxwell equations:

Is using quasi-solutions enough?

#### Quasi-Trefftz properties at convergence

**Hypothesis :** the local solver verifies the convergence property for  $\gamma_T \mathbf{E}^T + Z_{\partial T} \gamma_{\times}^T \mathbf{H}^T = g^T$  on  $\partial T$ :

 $\|\gamma_t(\mathbf{E}^T - \mathbf{E}_{\mathbf{h}}^T)\|_{\mathbf{L}^2(\partial T)} + \|\gamma_{\times}^T(\mathbf{H}^T - \mathbf{H}_{\mathbf{h}}^T)\|_{\mathbf{L}^2(\partial T)} \leq \varepsilon_{ls}^T \|\mathbf{g}^T\|_{\mathbf{L}^2(\partial T)} \text{ with } \varepsilon_{ls}^T \to 0,$ 

when the local solver mesh is refined.

If the local solver mesh is sufficiently refined, the quasi-Trefftz formulation verifies the

- weak-coercivity,
- contraction properties,

of the original Trefftz formulation.

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# Calibration of the FR auxiliary solver

Trefftz approach rests on the use of exact solutions of the Maxwell equations:

#### Is using quasi-solutions enough? How to adapt the FR order to the Trefftz one?

- Uniform Cartesian mesh  $\mathcal{T}_H(\Omega)$  of the domain  $\Omega$ : N macro-cells per direction.
- Uniform Cartesian mesh  $\mathfrak{T}_h(\partial T)$  of  $\partial T$ : M micro-faces per direction.
- Piecewise polynomial BCs of degree  $k_{QT} = 1$ .



Figure: Mesh  $\mathcal{T}_H(\Omega)$  for N = 3.



Figure: Mesh  $\mathfrak{T}_h(\partial T)$  for M=2 and  $k_{QT}=1.$ 

 $\implies$  No theoretical *a priori* error estimates in  $L^2(\Omega)$ -norm: **numerical convergence in** mesh and number of basis functions [Fure et al. 2020].

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# Calibration of the FR auxiliary solver (2D Helmholtz)

#### Is using quasi-solutions enough? How to adapt the FR order to the Trefftz one?

 $\implies$  Mesh-convergence: refinement of the macro-mesh  $\mathfrak{T}_H(\Omega)$  for M=2.



- Regime 1 ('imperfect' local solver): quasi-optimal for  $k_{FR} \ge k_{QT} + 2$  and the higher the order and the more refined the nano-mesh is, the later the transition.
- Regime 2 ('perfect' local solver): common super-convergence for  $k_{FR} \ge k_{QT} + 1$ .

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- Regime 1 ('imperfect' local solver): quasi-optimal for  $k_{FR} \ge k_{QT} + 2$  and the higher the order and the more refined the nano-mesh is, the later the transition.
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# Calibration of the FR auxiliary solver

#### Is using quasi-solutions enough? How to adapt the FR order to the Trefftz one?

 $\implies$  Mesh-convergence: refinement of the macro-mesh  $\mathcal{T}_H(\Omega)$  for M = 2.



• Need of  $k_{FR} \ge k_{QT} + 2$  for quasi-optimal orders, but no interest in over-resolving!

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### Numerical experiments: convergence in the number of basis functions

Is using quasi-solutions enough? How to adapt the FR order to the Trefftz one?

 $\implies$  Local basis enrichment through refinement of the micro-mesh  $\Im_h(\partial T)$  for N=2.



• Need of  $k_{FR} \ge k_{QT}$  to be sufficiently 'quasi-solution'.

• Improvement for  $k_{FR} \ge k_{QT} + 1$ , but limited interest in over-resolving!

# Conclusions on the quasi-Trefftz approach

- Classic numerical schemes are not well adapted to wide domain simulations, contrary to the Trefftz method.
- The classic choice of **Plane Waves** leads to limitations: we introduce a **Quasi-Trefftz approach**, in which a FR solver computes approximate Maxwell solutions associated to polynomial BCs in each cell.
- Good numerical independence properties of the basis functions.
- Robust approximation properties of the basis, even for complex local solutions.
- Numerical calibration of the local solver:  $k_{FR} \ge k_{QT} + 2$  for asymptotic quasi-optimality, but no need to over-resolve!
- 2 convergence regimes according to the local solver resolution: as refined as possible!

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# Improvement of the transfer of information: DtN approximation and local optimisation

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### Problematic: reduce the number of iterations to convergence

Iterative solution of the linear system thanks to a Krylov method, as GMRES: improving the information transfer would allow to reduce the number of iterations to convergence.

> (i) Solution evolution (ii) Solution modification Figure: Iterative convergence: information propagation through iterations.

# Trefftz formulation generalisation

#### Principle of the proposed generalisation for the 2D Helmholtz equation:

• Introduction of the outgoing trace from a cell T ( $0^T = i\kappa Id$  classically) :

$$\gamma_{out}^{T} \mathbf{y}^{T} = -v^{T} + \mathbf{O}^{T} \left( u^{T} \right) \quad \text{avec} \quad v^{T} = \frac{\kappa}{\kappa_{T}} \partial_{\mathbf{n}_{T}} u^{T}.$$

• Consistent numerical traces  $\hat{u}$  and  $\hat{v^T}$  defined from  $\gamma_{out}^T \mathbf{y}^T$  and  $\gamma_{out}^K \mathbf{y}^K$ .

 $\implies$  Need to have  $O^T$  in a set of operators C which ensures the formulation conserves classic properties (coercivity and contraction):

 $\mathcal{C} \subset \{\mathcal{O}, \text{ there exists a definite linear operator } \Lambda, \ \mathcal{O} = i\Lambda\Lambda^* \text{ for each face}\}.$ 

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### Approximation of the Dirichlet-to-Neumann operator

Trefftz method can be seen as the variational formulation of

$$v^T + \mathbb{O}^K(u^T) = -v^K + \mathbb{O}^K(u^K) \quad \text{on} \quad \partial T \cap \partial K.$$

The theory of Domain Decomposition Methods (DDM) ensures the **optimal** operator  $O^T$  is the **exterior Dirichlet-to-Neumann** (DtN) one.

3 approximation types :

• Exact operator for classic outgoing waves with respect to the cell centre x<sub>0</sub>:

$$\mathbb{O}^T = i\kappa \mathbf{d} \cdot \mathbf{n}_{\partial\Omega} \text{ (directive PWs)} \text{ and } \mathbb{O}^T = -\kappa \frac{\mathrm{H}_1^{(1)}(\kappa r)}{\mathrm{H}_0^{(1)}(\kappa r)} \text{ (Green kernel)},$$

for  $\mathbf{d} = (\mathbf{x} - \mathbf{x}_0)/r$  and  $r = |\mathbf{x} - \mathbf{x}_0|$ .

 Operator with same principal symbol as the classic approximation [El Bouajaji et al. 2014]

$$\mathbb{O}^T = i\kappa \sqrt{Id + \frac{\Delta_F}{\kappa^2}},$$

as Padé-type approximations [Després et al. 2021]

$$\mathcal{O}^T = i\kappa \left( Id - \frac{\Delta_F}{2\kappa^2} \right)^{-1}$$

 $\rightarrow$  Face point of view: BCs for the Laplace-Beltrami operator?

• Approximation thanks to neural networks (NN): allows to enforce  $\mathbb{O}^T \in \mathbb{C}$ .

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# Local optimisation

Being given local parameters P, we want to optimise the operators  $\mathcal{O}^T$  and  $\mathcal{O}^K$  with respect to an information transfer measure  $\rho_P$ .

$$Z_{1} \begin{array}{c|c} Z_{6} & Z_{5} \\ \hline \kappa_{T} & \kappa_{K} \\ \hline Z_{2} & Z_{3} \end{array} \xrightarrow{} \mathcal{O}^{T} \xrightarrow{} \mathcal{O}^{K}$$

$$\mathcal{O}_{P}^{T,*}, \mathcal{O}_{P}^{K,*} = \underset{\mathcal{O}^{T}, \mathcal{O}^{K} \in \mathcal{C}}{\operatorname{argmin}} \rho_{P}(\mathcal{O}^{T}, \mathcal{O}^{K}).$$

Different problematics:

- Size of the local parameters *P*.
- Possibility to develop interpolation approximations of ρ<sub>P</sub> for efficient optimisation.
- Which measure  $\rho_P$  ?
  - Reduce the distance of the eigenvalues to 1 (fixed point vision).
  - Increase the distance of the eigenvalues to 0 (GMRES point of view).
- Does local optimisation lead to a similar behaviour at the global level?

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#### Preliminary results and conclusion



Figure: GMRES convergence for different generalisations: analytic, NN-based and optimisation-based.

- We proposed a generalisation of the Trefftz method, thanks to general numerical fluxes keeping the original properties of the classic formulation.
- Trefftz approach can be interpreted as a DDM one: we introduced approximations of the Dirichlet-to-Neumann operator, which is supposed to be optimal.
- Introduction of local optimisation problems for the operators, depending on the surrounding parameters.
- Promising preliminary results to reduce the number of iterations to convergence: about 25% reduction by using NNs.

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