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Domain decomposition method for harmonic wave propagation: a general presentation

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Dedicated to the memory of Philippe Chevalier who made a contribution to domain decomposition methods for electromagnetic waves before his accidental death in May 1998 prevented his defending his thesis.

Abstract

In this paper we give a general presentation of non overlapping domain decomposition methods for harmonic wave propagation models. Our abstract framework leads to concise convergence proofs and contains some recent developments about the use of non local transmission conditions. It also includes applications to acoustic, electromagnetic or elastic waves, as well as the treatment of space discretization. © 2000 Elsevier Science S.A. All rights reserved.

1. Introduction

Time harmonic wave propagation phenomena appear to be a privileged field of application of domain decomposition methods (DDM): indeed, the solution is in general highly oscillatory (except may be for very low frequency computations) and its approximation requires generally a very thin space discretization, leading to large size linear system which cannot reasonably be solved using direct factorization methods. Moreover, these are linear systems over the complex field which are non hermitian (even if they are symmetric in general), so that usual iterative methods either do not converge or converge only slowly, without any theoretical guarantees of convergence, [8]. In this context, it is natural to try to reduce the solution of the full problem to a succession of smaller local problems that could be handled for instance by direct methods. This is precisely the purpose of the domain decomposition techniques. However, the solution of the Helmholtz type equations which describe wave propagation in the frequency domain has not the same nice properties than the solutions of more standard elliptic equations such as Laplace equation: it does not naturally correspond to the minimum of a quadratic functional, the maximum principle does not hold, ... That is why the ideas of the famous overlapping Schwartz algorithm [26–28] cannot be applied to harmonic wave propagation models. This probably explains why the research on DDM for such models started only recently. The objective of the present paper is to give a survey of some works in this field made by the authors and some of their colleagues and to present some more recent developments.

The methods we are going to speak about are always non overlapping iterative methods: the domains only communicate through their interfaces. As usual all the difficulty lies in the way one makes the different sub-domains communicate between each others: the key points are thus the choice of good transmission

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conditions and the choice of a clever iterative scheme. The starting point of the methods is the choice of suitable transmission conditions which involve an appropriate linear combination of Neumann and Dirichlet data with imaginary coefficients. To indicate the basic ideas, let us consider the very simple example of the 1D Helmholtz equation

$$-\partial_{xx}^2 u - \omega^2 u = f, \quad x \in \,]-a, a[, \tag{1}$$

with transparent boundary conditions at $x = \pm a$,

$$(\partial_x u \pm i\omega u)\,(\pm a) = 0. \tag{2}$$

Let us consider a partition of the interval $]-a, a[$ into sub-intervals $]x_k, x_{k+1}[$, $x_0 = -a < x_1 < \cdots < x_K = +a$. Denoting by $u_{k+(1/2)}$ the restriction of u to $]x_k, x_{k+1}[$, we have

$$-\partial_{xx}^2 u_{k+(1/2)} - \omega^2 u_{k+(1/2)} = f, \quad x \in \,]x_k, x_{k+1}[, \tag{3}$$

with the transmission conditions

$$u_{k+(1/2)}(x_k) = u_{k-(1/2)}(x_k), \qquad 0 < k < K-1,$$
$$\partial_x u_{k+(1/2)}(x_k) = \partial_x u_{k-(1/2)}(x_k), \quad 0 < k < K-1, \tag{4}$$

and the boundary conditions

$$\partial_x u_{(1/2)}(x_0) - i\omega u_{(1/2)}(x_0) = 0,$$
$$\partial_x u_{K-(1/2)}(x_K) + i\omega u_{K-(1/2)}(x_K) = 0. \tag{5}$$

The idea of the basic method, due to Despres [14–16,18] and independently to Shaidurov and Ogorodnikov [31] consists in rewriting (5) in the following form: for $k = 1, \ldots, K-1$

$$\partial_x u_{k+(1/2)}(x_k) + i\omega u_{k+(1/2)}(x_k) = \partial_x u_{k-(1/2)}(x_k) + i\omega u_{k-(1/2)}(x_k),$$
$$-\partial_x u_{k+(1/2)}(x_k) + i\omega u_{k+(1/2)}(x_k) = -\partial_x u_{k-(1/2)}(x_k) + i\omega u_{k-(1/2)}(x_k). \tag{6}$$

Therefore, looking at a single domain $]x_k, x_{k+1}[$, we have in particular, by picking one of the two equations (6) at $x = x_k$ and $x = x_{k+1}$

$$-\partial_{xx}^2 u_{k+(1/2)} - \omega^2 u_{k+(1/2)} = f, \quad x \in \,]x_k, x_{k+1}[,$$
$$-\partial_x u_{k+(1/2)}(x_k) + i\omega u_{k+(1/2)}(x_k) = -\,\partial_x u_{k-(1/2)}(x_k) + i\omega u_{k-(1/2)}(x_k),$$
$$\partial_x u_{k+(1/2)}(x_{k+1}) + i\omega u_{k+(1/2)}(x_{k+1}) = \partial_x u_{k+(3/2)}(x_{k+1}) + i\omega u_{k+(3/2)}(x_{k+1}). \tag{7}$$

This system also holds for $k = 0$ and $k = K-1$ if we use the convention $u_{-(1/2)} = u_{K+(1/2)} = 0$. It can be seen as a problem which defines $u_{k+(1/2)}$, with both $u_{k-(1/2)}$ and $u_{k+(3/2)}$ considered to be known.

We then propose the following iterative procedure

$$\left\{ u_{k-(1/2)}^n \right\}_{k=1, K} \; \longrightarrow \; \left\{ u_{k-(1/2)}^{n+1} \right\}_{k=1, K}, \tag{8}$$

where we have set

$$-\partial_{xx} u_{k+(1/2)}^{n+1} - \omega^2 u_{k+(1/2)}^{n+1} = f, \quad x \in \,]x_k, x_{k+1}[,$$
$$-\partial_x u_{k+(1/2)}^{n+1}(x_k) + i\omega u_{k+(1/2)}^{n+1}(x_k) = -\,\partial_x u_{k-(1/2)}^n(x_k) + i\omega u_{k-(1/2)}^n(x_k),$$
$$\partial_x u_{k+(1/2)}^{n+1}(x_{k+1}) + i\omega u_{k+(1/2)}^{n+1}(x_{k+1}) = -\,\partial_x u_{k+(3/2)}^n(x_{k+1}) + i\omega u_{k+(3/2)}^n(x_{k+1}). \tag{9}$$

It appears that this method is optimal in the sense that it converges exactly in a finite number of iterations. Indeed, introducing the error defined in each sub-domain by

$$e_{k+(1/2)}^n = u_{k+(1/2)} - u_{k+(1/2)}^n. \tag{10}$$

We know that the $e_{k+(1/2)}^n$ satisfy (9) with $f = 0$. Therefore we can write

$$e_{k+(1/2)}^n = a_{k+(1/2)}^n \exp(i\omega x) + b_{k+(1/2)}^n \exp(-i\omega x). \tag{11}$$

Observing that

$$\partial_x e_{k+(1/2)}^n + i\omega e_{k+(1/2)}^n = 2i\omega a_{k+(1/2)}^n \exp(i\omega x),$$

$$\partial_x e_{k+(1/2)}^n - i\omega e_{k+(1/2)}^n = -2i\omega b_{k+(1/2)}^n \exp(-i\omega x) \tag{12}$$

the transmission conditions lead to

$$a_{k-(1/2)}^{n+1} = a_{k+(1/2)}^n, \quad k = 1, 2, \ldots, K - 1,$$

$$b_{k+(1/2)}^{n+1} = b_{k-(1/2)}^n, \quad k = 1, 2, \ldots, K - 1, \tag{13}$$

while the boundary conditions express that

$$a_{K-(1/2)}^n = 0, \qquad b_{1/2}^n = 0, \quad \forall n. \tag{14}$$

From (13) and (14), one easily deduces that, after K iterations

$$a_{k-(1/2)}^K = b_{k-(1/2)}^K = 0, \quad k = 1, \ldots, K, \tag{15}$$

which means that the process converges in K iterations, where K is the number of sub-domains. In this sense, the iterative procedure we proposed based on the use of the mixed quantities $\pm(\mathrm{d}u/\mathrm{d}x) + i\omega u$ is optimal.

The methods we are going to treat in the paper are essentially extensions and variants of this basic algorithm. The extensions mainly concern higher space dimensions and other types of equations. The variants concern the interface conditions and the iteration procedure. In this paper, we make a particular effort of abstraction in order to give a unified presentation of all the extensions and variants. The outline is as follows.

In Section 2, we present the model problem (the Helmholtz equation) that we consider in the major part of this article. The main objective of this section is to write an equivalent transmission problem on a partition of Ω using, as in 1D, local transmission operators. We reformulate this transmission problem in terms of a new problem whose unknowns are quantities defined at the interfaces between the different sub-domains. We identify an abstract general form of this problem via the introduction of appropriate linear operators. In Section 3, we describe the various iterative domain decomposition algorithms that we have chosen to discuss. In particular, we introduce the notion of Jacobi and Gauss–Seidel algorithms (Section 3.2) as well as their relaxed version. In Section 3.3, we introduce a more general class of algorithms based on the use of a more general class of transmission operators. We show that they fit the general framework that we have constructed in the previous section. Section 4 is devoted to the convergence analysis. Our proof is only based on a priori estimates. We emphasize a particular choice of partition of Ω and of transmission operators which guarantees geometric convergence. In Section 5, we analyze in a more quantitative way the speed of convergence of our algorithms in a very particular case of two domains. This permits us to discuss the choice of some of the parameters of the method. In Section 6, we present various extensions within the abstract formalism. They concern first the space discretization of the model problem then the application of our methods to other physically relevant problems. Finally, we illustrate some aspects of our theoretical analysis by some numerical results.

We cannot treat in a few pages all the aspects of DDM for harmonic wave propagation. Let us conclude this introduction by a rapid review of the literature (a useful complement to the present paper). Historically, we must refer to the pioneering works of Despres about the Helmholtz equation with purely local transmission conditions [14,15], for the theoretical aspects, [17,18] for computational issues, treated also in [32] and Shaidurov and Ogorodnikov [31] (see also [19] for an extension to absorbing media and [20] for a more algebraic point of view). In particular, one can find in [17,18] how to couple the iterative methods we propose in this paper with more sophisticated algorithms such as GMRES and its variants (The algorithms we introduced here can be used as efficient preconditionners). The use of coarse grid preconditioners is

considered in [12]. Benamou extended the ideas therein to treat [17,18] optimal control problems related to harmonic wave propagation (see [1–3] for a review). The case of non local transmission conditions is treated in [21,22], while in [6,7], Chevalier and Nataf propose to use of second order tangential differential operators (an intermediate situation). Overlapping methods are introduced in [5]. For the case of 3D Mawxell's equations, we refer to [9,18].

2. A model problem – reformulation at interfaces

2.1. Description of the model problem

Here, Ω denotes a bounded domain of $\mathbf{R}^N$ with boundary Γ that we assume to be sufficiently smooth (at least Lipschitz continuous). We consider the problem with complex valued unknown function u

$$(P_\Omega^f) \quad \begin{cases} \text{Find } u \in H^1(\Omega) \quad \text{such that} \\ -\Delta u - \omega^2 u = f \quad \text{in } \Omega, \\ \partial_\nu u + \mathrm{i}\omega u = 0 \quad \text{on } \Gamma, \end{cases} \tag{16}$$

(ν denotes the unit outgoing normal vector to Γ). For instance, the boundary condition can be seen as an approximate radiation condition at finite distance (as an approximation of the propagation of a wave in $\mathbf{R}^N$). The right hand side is the source term and $\omega > 0$ stands for the frequency.

Problem (P_Ω^f) enters in a more general class of problems ($\mathcal{O}$ referring to a bounded domain of $\mathbf{R}^N$)

$$(P_\mathcal{O}^{f,g}) \quad \begin{cases} \text{Find } u \in H^1(\mathcal{O}) \quad \text{such that} \\ -\Delta u - \omega^2 u = f \quad \text{in } \mathcal{O}, \\ \partial_\nu u + \mathrm{i}\omega u = g \quad \text{on } \partial\mathcal{O}. \end{cases} \tag{17}$$

Let us recall the following classical result (for a proof, see [10] or [15] for instance).

Theorem 1. *If (f, g) belongs to $L^2(\mathcal{O}) \times L^2(\partial\mathcal{O})$, $(P_\mathcal{O}^{f,g})$ admits a unique solution in $H^1(\mathcal{O})$ which satisfies $\Delta u \in L^2(\mathcal{O})$, and $\partial_\nu u \in L^2(\partial\mathcal{O})$. Moreover there exists a constant $C = C(\mathcal{O}) > 0$ such that*

$$\|u\|_{H^1(\mathcal{O})} + \|\partial_\nu u\|_{L^2(\partial\mathcal{O})} \leqslant C\Big(\|f\|_{L^2\mathcal{O}} + \|g\|_{L^2(\partial\mathcal{O})}\Big). \tag{18}$$

Let us simply recall that the proof uses the Fredholm's alternative which reduces the existence proof to a uniqueness result. For such a result, one first prove that if $f = g = 0$, then $u = \partial_\nu u = 0$ on $\partial\mathcal{O}$ which implies by the unique continuation principle, [11], that u is zero in $\mathcal{O}$. This first step deeply uses the absorption term $\mathrm{i}\omega u$ in the boundary condition.

Remark 1. *We consider the Helmholtz equation with constant coefficients for simplicity. However, the reader will easily convince himself that all what follows can be extended without any difficulty to the following problem (more realistic from the point of view of the applications)*

$$\begin{cases} \text{Find } u \in H^1(\Omega) \quad\quad\quad \text{such that} \\ -\mathrm{div}\left(\frac{1}{\mu}\vec{\nabla}u\right) - \epsilon\omega^2 u = f \quad \text{in } \Omega, \\ \partial_\nu u + \mathrm{i}\omega Z u = g \quad \text{on } \partial\Omega, \end{cases} \tag{19}$$

where $\epsilon(x), \mu(x)$ are bounded with bounded inverse positive functions defined in Ω and $Z(x) : \Gamma \to \mathbb{C}$ satisfies $\Re(Z) \geqslant 0$ and $\Re(Z) \geqslant \delta > 0$ for $x \in \Gamma_0$, where Γ_0 is a subset of Γ with non zero measure.

2.2. An equivalent transmission problem

Let us consider a partition of Ω into K sub-domains Ω_k:

$$\overline{\Omega} = \bigcup_{k=1}^{K} \overline{\Omega}_k,$$

$$\Omega_k \cap \Omega_j = \emptyset \quad \text{for } j \neq k, \tag{20}$$

$$\Omega_k \text{ is Lipschitz continuous.}$$

In what follows, we shall distinguish two notations for $\partial\Omega_k \cap \partial\Omega_j$ (interface between Ω_k and Ω_j).

$$\Sigma_{kj} = \partial\Omega_k \cap \partial\Omega_j, \quad \text{as a part of } \Omega_k,$$

$$\tag{21}$$

$$\Sigma_{jk} = \partial\Omega_k \cap \partial\Omega_j, \quad \text{as a part of } \Omega_j.$$

This is useful for distinguishing the two traces on $\partial\Omega_k \cap \partial\Omega_j$ for a function which can be discontinuous across an interface. See Fig. 1.

Remark 2. *If the distinction is not needed, we shall use indifferently the notation Σ_{kj}, Σ_{jk} or $\partial\Omega_k \cap \partial\Omega_k$ without mentioning it explicitly.*

Finally, we shall set

$$\Sigma_{kk} = \partial\Omega \cap \partial\Omega_k. \tag{22}$$

On $\partial\Omega_k$, we shall denote by v_k the unit normal vector which is outgoing with respect to Ω_k. Obviously, $v_k + v_j = 0$ on $\Sigma_{kj} = \Sigma_{jk}$. Finally, if u is a function of $L^2(\Omega)$, its restriction to Ω_k will be denoted by u_k. With this notation, it is clear that u is solution of (P_Ω^f) if and only if $\{u_k, \ k = 1, K\}$ is a solution of the transmission problem

$$(\mathscr{P})_t \quad \begin{cases} -\Delta u_k - \omega^2 u_k = f_k & \text{in } \Omega_k, \ k = 1, \ldots, K, \\[2mm] \partial_{v_k} u_k + i\omega u_k = 0 & \text{on } \Sigma_{kk}, \\[2mm] u_k = u_j & \text{on } \partial\Omega_k \cap \partial\Omega_j, \ k \neq j, \\[2mm] \partial_{v_k} u_k = -\partial_{v_j} u_j & \text{on } \partial\Omega_k \cap \partial\Omega_j, \ k \neq j. \end{cases} \tag{23}$$
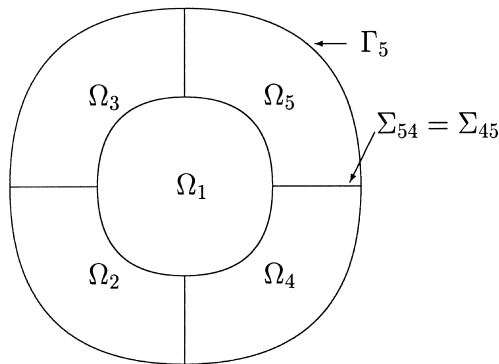

Fig. 1. The domain Ω and its decomposition.

The two transmission conditions can be rewritten in an equivalent way as ($k \neq j$)

$$\partial_{v_k} u_k + i\omega u_k = -\partial_{v_j} u_j + i\omega u_j \quad \text{on } \Sigma_{kj},$$

$$\partial_{v_j} u_j + i\omega u_j = -\partial_{v_k} u_k + i\omega u_k \quad \text{on } \Sigma_{jk}. \tag{24}$$

Therefore, $(\mathscr{P})_t$ can be seen as a collection of local problems of the form $(P_{\mathcal{O}}^{f,g})$

$$-\Delta u_k - \omega^2 u_k = f_k \quad \text{in } \Omega_k,$$

$$\partial_{v_k} u_k + i\omega u_k = -\partial_{v_j} u_j + i\omega u_j \quad \text{on } \Sigma_{kj},\ k \neq j \tag{25}$$

$$\partial_{v_k} u_k + i\omega u_k = 0 \quad \text{on } \Sigma_{kk}.$$

2.3. Some operators

Our goal in this section will be to reformulate the transmission problems as problems whose unknown will be functions defined at the boundaries $\partial\Omega_k$, namely the traces

$$x_k = (-\partial_{v_k} u_k + i\omega u_k)_{/\partial\Omega_k}. \tag{26}$$

For what follows, we are led to make the following regularity assumptions

$$\text{(H)}: \ x = (x_k) \in \prod_{k=1}^{K} V_k, \quad V_k = L^2(\partial\Omega_k), \tag{27}$$

which is not very restrictive in practice (Note that it is true if Ω_k is smooth enough).

Note that the space V can also be decomposed according to the splitting of the interfaces $\partial\Omega_k$:

$$V = \prod_{j,k=1}^{K} V_{kj}, \quad V_{kj} = L^2(\Sigma_{kj}), \tag{28}$$

so that any element y of V can be written as

$$y \equiv (y_k)_{k=1,K} \equiv (y_{kj})_{k,j=1,K}. \tag{29}$$

In order to characterize x as a solution of a linear equation V, we are led to introduce some operators.

Operators E and S. We first introduce

$$E \in \mathscr{L}\left(V; \prod_{k=1}^{K} H^1(\Omega_k)\right), \tag{30}$$

defined by

$$e = (e_k) = E(y) \iff \begin{vmatrix} -\Delta e_k - \Delta e_k = 0 & \text{in } \Omega_k, \\ \partial_{vk} e_k + i\omega e_k = y_k & \text{on } \partial\Omega_k. \end{vmatrix} \tag{31}$$

The fact that E is well defined, linear and continuous is a direct consequence of Theorem 1. Note in particular that the boundary condition implies

$$\partial_{v_k} e_k \in L^2(\partial\Omega_k), \quad k = 1, \dots K. \tag{32}$$

This permits us to define another operator

$$S \in \mathscr{L}(V; V), \tag{33}$$

by

$$x = S(y) \Longleftrightarrow x_k = -\partial_{v_k} e_k + i\omega e_k, \quad \text{where } e = E(y). \tag{34}$$

A fundamental property of operator S is the following one:

Lemma 1. *The operator S is unitary in V.*

Proof. The proof is classical. Let (x, y) in $V \times V$, $u = E(x)$ and $v = E(y)$. By definition

$$(x, y) = \sum_k \int_{\partial\Omega_k} (\partial_{v_k} u_k + i\omega u_k)\overline{(\partial_{v_k} v_k + i\omega v_k)} \, d\sigma,$$

$$(Sx, Sy) = \sum_k \int_{\partial\Omega_k} (-\partial_{v_k} u_k + i\omega u_k)\overline{(-\partial_{v_k} v_k + i\omega v_k)} \, d\sigma, \tag{35}$$

so that

$$(Sx, Sy) - (x, y) = -2i\omega \sum_k \int_{\partial\Omega_k} (\partial_{v_k} \bar{v}_k u_k - \partial_{v_k} u_k \bar{v}_k) \, d\sigma. \tag{36}$$

On the other hand by definition of E

$$\Delta u_k + \omega^2 u_k = 0 \quad \text{(i)},$$

$$-\Delta \bar{v}_k - \omega^2 \bar{v}_k = 0 \quad \text{(ii)}. \tag{37}$$

We multiply (37)-(i) by $\bar{v}_k$, (37)-(ii) by u_k, add the two equations and integrate over Ω_k, we get

$$\int_{\Omega_k} (\Delta u_k \bar{v}_k - u_k \Delta \bar{v}_k) \, dx = 0. \tag{38}$$

Using Green's theorem, we obtain

$$\int_{\partial\Omega_k} (\partial_{v_k} u_k \bar{v}_k - \partial_{v_k} \bar{v}_k u_k) \, d\sigma = 0, \tag{39}$$

which proves, using (36), that

$$\forall (x, y) \in V \times V, \quad (Sx, Sy) = (x, y). \tag{40}$$

and S is unitary. $\square$

Remark 3. *We have used the notation S in relation to those used in the scattering theory, [24,33] (tome 2, ch. 9, pp. 145–240). In some sense, S can be seen as a local scattering operator.*

The exchange operators Π, Π^+, Π^-. We introduce

$$\Pi \in \mathcal{L}(V; V), \tag{41}$$

defined by

$$z = \Pi y \Longleftrightarrow \begin{vmatrix} z_{kj} \equiv z_{k/\Sigma_{kj}} = y_{j/\Sigma_{jk}} \equiv y_{jk}, & k \neq j, \\ z_{kk} \equiv z_{k/\Sigma_{kk}} = 0. \end{vmatrix} \tag{42}$$

The main properties of the operator Π are summarized in the following obvious lemma.

Lemma 2. *The operator Π is a symmetric contraction in V and $\Pi^2 = \Pi * \Pi = P_\Sigma$ where P_Σ is the orthogonal projector onto the subspace $V_\Sigma = \{x \in V, \ x_{/\Gamma} = 0\}$.*

In the following, we will use the following decomposition of Π (which depends of the numbering of the sub-domains)

$$\Pi = \Pi^+ + \Pi^-, \tag{43}$$

where

$$
z = \Pi^+ y \Longleftrightarrow
\left|
\begin{array}{l}
z_{kj} = z_{k/\Sigma_{kj}} = y_{j/\Sigma_{jk}} = y_{jk}, \quad k < j, \\[6pt]
z_{kj} = z_{k/\Sigma_{kj}} = 0, \quad k > j, \\[6pt]
z_k = z_{k/\Sigma_{kk}} = 0.
\end{array}
\right.
\tag{44}
$$

The following properties of $\Pi^\pm$ are straightforward.

Lemma 3. *The operators $\Pi^\pm$ are mutually adjoint and $P_{\Sigma^+} = \Pi^+\Pi^-$ (resp. $P_{\Sigma^-} = \Pi^-\Pi^+$) is the orthogonal projector onto the subspace $V_\Sigma^+ = \{x \in V, \; x_{/\Sigma_{kj}} = 0, \; \forall k \geqslant j\}$ (resp. $V_\Sigma^- = \{x \in V, \; x_{/\Sigma_{kj}} = 0, \; \forall k \leqslant j\}$).*

In Section 4, we shall also use the operator P_Γ, namely the orthogonal projector on V_Γ, the orthogonal complement of

$$
V_\Sigma = \{x \in V, x_{/\Sigma_{kj}} = 0, \quad \forall k \neq j\}.
$$

Operator A. We introduce

$$
A = S\Pi \in \mathscr{L}(V, V),
\tag{45}
$$

Lemma 4. *The operator A is contractive in V and $I - A$ is injective.*

Proof. Only the injectivity of $I - A$ needs to be clarified. It is in fact a consequence of the uniqueness result for $(P_0^{f,g})$. Indeed, setting $z = (I - A)x$ and $u = E(\Pi x)$, one easily computes that

$$
z_{kj} = (\partial_{v_k} u_k + \partial_{v_j} u_j) - i\omega(u_k - u_j).
\tag{46}
$$

If $z = 0$, writing that $z_{kj} = z_{jk} = 0$, we easily get that

$$
\begin{aligned}
\partial_{v_k} u_k + \partial_{v_j} u_j &= 0, \quad \text{on } \partial\Omega_k \cap \partial\Omega_j, \\
u_k - u_j &= 0, \quad \text{on } \partial\Omega_k \cap \partial\Omega_j,
\end{aligned}
\tag{47}
$$

which means that u is solution of $\Delta u + \omega^2 u = 0$ in Ω. Writing that $z_{kk} = 0$, we get $\partial_v u + i\omega u = 0$ on each Σ_{kk}. Therefore $u = 0$ and $x = 0$. $\square$

2.4. Reformulation of the transmission problem at interfaces

Let us introduce $v_k \in H^1(\Omega_k)$ as the solution of the local problem

$$
\begin{aligned}
-\Delta v_k - \omega^2 v_k &= f_k, \\[6pt]
\partial_{v_k} v_k + i\omega v_k &= 0 \quad \text{on } \partial\Omega_k,
\end{aligned}
\tag{48}
$$

and let us set

$$
g_k = -\partial_{v_k} v_k + i\omega v_{k/\partial\Omega_k},
\tag{49}
$$

so that

$$
g = (g_k) \in V.
\tag{50}
$$

We have:

Theorem 2.
  (i) *If $u = (u_k)$ is solution of $(\mathscr{P}_t)$, then $x \in V$ defined by (26) is solution of*
$$
(\mathscr{P}_g) : (I - A)x = g.
\tag{51}
$$
  (ii) *Reciprocally, if x is solution of $(\mathscr{P}_g)$, then $u = v + E(\Pi x)$ is solution of $\mathscr{P}_t$*

Proof. Let us set

$$u_k = v_k + w_k. \tag{52}$$

As $\partial_{v_k} v_k + i\omega v_k = 0$, we have

$$\partial_{v_k} w_k + i\omega w_k = \partial_{v_k} u_k + i\omega u_k = \begin{vmatrix} -\partial_{v_j} u_j + i\omega u_j & \text{on } \Sigma_{kj}, \ k \neq j \\ \\ 0 & \text{on } \Sigma_{kk}. \end{vmatrix} \tag{53}$$

Finally, w_k satisfies:

$$-\Delta w_k - \omega^2 w_k = 0,$$
$$\partial_{v_k} w_k + i\omega w_k = (\Pi x)_k \quad \text{on } \partial\Omega_k, \tag{54}$$

by definition of E, this implies $w = E(\Pi x)$ and:

$$-\partial_{v_k} w_k + i\omega w_k = S\Pi x = Ax, \tag{55}$$

that is to say, using (26) and (49),

$$x - g = Ax. \tag{56}$$

This proves (i). The proof of (ii) is quite similar and will be omitted there. □

3. The iterative domain decomposition algorithms

3.1. The basic iterative algorithms and its interpretations

We introduce in this section the algorithm initially proposed by Despres in [13,15]. It consists of constructing a sequence

$$u^n = (u_k^n) \in \prod_{k=1}^{K} H^1(\Omega_k), \tag{57}$$

by the following process:
Assuming that u^n is known and satisfies

$$-\partial_{v_k} u_k^n + i\omega u_k^n \in L^2(\partial\Omega_k), \quad k = 1, \ldots, K, \tag{58}$$

we define $u^{n+1} = (u_k^{n+1})$, where u_k^{n+1} is the solution of the non homogeneous boundary value problem:

$$-\Delta u_k^{n+1} - \omega^2 u_k^{n+1} = f \quad \text{in } \Omega_k,$$
$$\partial_{v_k} u_k^{n+1} + i\omega u_k^{n+1} = -\partial_{v_j} u_j^n + i\omega u_j^n \quad \text{on } \Sigma_{kj}, \ k \neq j, \tag{59}$$
$$\partial_{v_k} u_k^{n+1} + i\omega u_k^n = 0 \quad \text{on } \Sigma_{kk}.$$

If the initial value of u_k^0 is chosen such that $-\partial_{v_k} u_k^0 + i\omega u_k^0 \in L^2(\partial\Omega_k)$ (for instance if $u_k^0 = 0$, $\forall k$), it is clear via Theorem 1 that this process is well posed in $\prod_k H^1(\Omega_k)$. Note also that at each step n, u^n will be generally discontinuous across the various interfaces. However if u^n converges in some sense to $u \in H^1(\Omega_k)$, then taking formally the limit in (59), we see that the function u will satisfy to problem $(\mathscr{P}_t)$ equivalent to problem $(\mathscr{P}_\Omega^f)$.

It is now interesting to reinterpret algorithm (59) by using the formalism introduced in the previous section. Let us set

$$u_k^n = v_k + w_k^n, \tag{60}$$

we see that

$$-\Delta w_k^{n+1} - \omega^2 w_k^{n+1} = 0 \quad \text{in } \Omega_k,$$

$$\partial_{v_k} w_k^{n+1} + i\omega w_k^{n+1} = -\partial_{v_j} u_j^n + i\omega u_j^n \quad \text{on } \Sigma_{kj}, \; k \neq j, \tag{61}$$

$$\partial_{vk} w_k^{n+1} + i\omega w_k^{n+1} = 0 \quad \text{on } \Sigma_{kk}.$$

If we set

$$x_k^n = (-\partial_{v_k} u_k^n + i\omega u_k^n)_{/\partial\Omega_k} = (-\partial_{v_k} w_k^n + i\omega w_k^n)_{/\partial\Omega_k} + g_k, \tag{62}$$

system (61) is equivalent to $w_k^{n+1} = E(\Pi x^n)$, hence, by definition of S:

$$x_k^{n+1} = (-\partial_{v_k} w_k^{n+1} + i\omega w_k^{n+1}) + g_k = (S\Pi x^n + g)_k, \tag{63}$$

or equivalently:

$$x^{n+1} = Ax^n + g. \tag{64}$$

One easily checks that, if $x^0 = 0$ for instance,

$$x^n = (I + A + A^2 + \cdots + A^n)g, \tag{65}$$

so that the convergence of the process is formally linked to the convergence of the Neumann series:

$$(I - A)^{-1} = I + A + A^2 + \cdots + A^n + \cdots \tag{66}$$

Remark 4. *The convergence proof shows that the convergence of the series* (66) *only holds in a weak sense, as we will see later.*

It is also instructive to reinterpret the algorithm in a matrix form in order to establish an analogy with usual iterative methods in linear algebra. In what follows we shall represent $x \in V$ as a column vector whose kth component is $x_k \in L^2(\partial\Omega_k)$. With this representation, it is clear that the operators S and Π have the following block structure. (Π_{jk} is simply the multiplication by the characteristic function of Σ_{kj})

$$\Pi = \begin{bmatrix} 0 & \Pi_{12} & \Pi_{13} & \cdots & & \Pi_{1K} \\ \Pi_{21} & 0 & \Pi_{23} & & & \Pi_{2K} \\ \Pi_{31} & \Pi_{32} & \ddots & \ddots & & \Pi_{3K} \\ \vdots & & \ddots & \ddots & & \Pi_{K-1K} \\ \Pi_{K1} & \cdots & \cdots & \Pi_{KK-1} & & 0 \end{bmatrix}, \tag{67}$$

$$\Pi_{kj} \in \mathcal{L}(L^2(\partial\Omega_k), L^2(\partial\Omega_j)), \quad \Pi_{kj}^* = \Pi_{jk}, \tag{68}$$

$$S = \begin{bmatrix} S_1 & 0 & 0 & 0 \\ 0 & S_2 & \cdots & 0 \\ \vdots & 0 & \ddots & \\ 0 & 0 & & S_K \end{bmatrix}, \quad S_k \in \mathcal{L}(L^2(\partial\Omega_k)), \tag{69}$$

which illustrates that S is purely local while Π is an exchange operator. Performing the product $S\Pi$, we see that we can write (with $A_{ij} = S_i\Pi_{ij}$)

$$A = \begin{bmatrix} 0 & A_{12} & A_{13} & \dots & A_{1K} \\ A_{21} & 0 & A_{23} & & A_{2K} \\ A_{31} & A_{32} & \ddots & \ddots & A_{3K} \\ \vdots & & \ddots & \ddots & \vdots \\ A_{K1} & \dots & \dots & A_{KK-1} & 0 \end{bmatrix}. \tag{70}$$

Thus, the operator to invert, $I - A$ is written

$$I - A = \begin{bmatrix} I & -A_{12} & -A_{13} & \dots & -A_{1K} \\ -A_{21} & I & -A_{23} & & -A_{2K} \\ -A_{31} & -A_{32} & \ddots & \ddots & -A_{3K} \\ \vdots & & \ddots & \ddots & \vdots \\ -A_{K1} & \dots & \dots & -A_{KK-1} & I \end{bmatrix}, \tag{71}$$

and can be split as the sum of its diagonal part, in fact the identity operator and its anti-diagonal part namely $-A$. In that sense, algorithm $x^{n+1} = Ax^n + g$ is a Jocobi's algorithm.

3.2. Two variants of the basic algorithm using local transmission problems

3.2.1. The relaxed Jacobi's algorithm

The idea is to modify the transmission condition (59) by replacing the in-going quantity $\partial_{v_k} u_k^{n+1} + i\omega u_k^{n+1}$ by a linear combination of the same quantity at steps n and $n + 1$. In other words, the local problem to be solved in Ω_k writes

$$-\Delta u_k^{n+1} - \omega^2 u_k^{n+1} = f \quad \text{in } \Omega_k,$$

$$\alpha\big(\partial_{v_k} u_k^{n+1} + i\omega u_k^{n+1}\big) + (1 - \alpha)\big(\partial_{v_k} u_k^n + i\omega u_k^n\big) = -\partial_{v_j} u_j^n + i\omega u_j^n \quad \text{on } \Sigma_{kj},\ k \neq j, \tag{72}$$

$$\partial_{v_k} u_k^{n+1} + i\omega u_k^{n+1} = 0 \quad \text{on } \Sigma_{kk},$$

where α denotes some positive parameter. Note that one comes back to the original Jacobi's algorithm by taking $\alpha = 1$. Note also that the local problem to be solved in each Ω_k is of type ($\mathscr{P}_0^{f,g}$), which ensures that the sequence $u^n = (u_k^n)$ is well defined in $H^1(\Omega_k)$. To reinterpret this new algorithm at interfaces, we introduce, as in Section 3.2, $w_k^n = u_k^n - v_k$ and

$$x_k^n = \big(-\partial_{v_k} u_k^n + i\omega u_k^n\big)_{/\partial\Omega_k} = \big(-\partial_{v_k} w_k^n + i\omega w_k^n\big)_{/\partial\Omega_k} + g_k,$$

$$y_k^n = \big(\partial_{v_k} u_k^n + i\omega u_k^n\big)_{/\partial\Omega_k} = \big(\partial_{v_k} w_k^n + i\omega w_k^n\big)_{/\partial\Omega_k}. \tag{73}$$

As w_k^n satisfies $\omega^2 w_k^n + \Delta w_k^n = 0$, by definition of S, we get

$$x^n - g = Sy^n \quad \text{for each } n. \tag{74}$$

In addition, the boundary condition relating the x^n and y^n's reads

$$\alpha y^{n+1} + (1 - \alpha)y^n = \Pi x^n. \tag{75}$$

After multiplication by S, we eliminate y^{n+1} and y^n. Setting $r = 1/\alpha$, we obtain

$$x^{n+1} = (1 - r)x^n + r(Ax^n + g), \tag{76}$$

which is our relaxed Jacobi's algorithm. We shall we see that for ensuring convergence we shall need to impose $r \in \,]0, 1]$. It corresponds to apply a fixed point method to the equation $\mathscr{P}_g$ rewritten in the form

$$x = x - r((I - A)x - g). \tag{77}$$

A simple calculation, that we do not detail here, shows that u^n can be reconstructed from x^n by the formula

$$u^{n+1} = (1 - r)E(S^*(x^n - g)) + rE(\Pi x^n) + v, \tag{78}$$

or equivalently $(S^*g = S^*x - \Pi x)$

$$u^{n+1} = (1 - r)E(S^*(x^n - x)) + (1 - r)E(\Pi x) + rE(\Pi x^n) + v, \tag{79}$$

identity which is clearly consistent with the limit equation

$$u = E(\Pi x) + v. \tag{80}$$

Eq. (80) suggests us another candidate as an approximation of u

$$\tilde{u}^{n+1} = E(\Pi x^n) + v. \tag{81}$$

One easily shows that u^n and $\tilde{u}^n$ are related by

$$u^{n+1} = (1 - r)u^n + r\tilde{u}^n. \tag{82}$$

which shows that $\tilde{u}^n$ converges if and only if u^n converges. It is thus useful to indicate how the sequence $\tilde{u}^n$ can be computed directly. In a first step, one solves:

$$\begin{aligned}
-\Delta\tilde{u}_k^{n+1} - \omega^2\tilde{u}_k^{n+1} &= f \quad \text{in } \Omega_k, \\
\partial_{v_k}\tilde{u}_k^{n+1} + i\omega\tilde{u}_k^{n+1} &= x_{jk}^n \quad \text{on } \Sigma_{kj},\ k \neq j, \\
\partial_{v_k}\tilde{u}_k^{n+1} + i\omega\tilde{u}_k^{n+1} &= 0 \quad \text{on } \Sigma_{kk},
\end{aligned} \tag{83}$$

and, in a second step, the x_{kj} are updated according to

$$x_{kj}^{n+1} = (1 - r)x_{kj}^n + r(-\partial_{v_k}\tilde{u}_k^{n+1} + i\omega\tilde{u}_k^{n+1}). \tag{84}$$

3.2.2. The Gauss–Seidel algorithm

This variant exploits the ordering of the domains and consists in taking profit of the subdomains Ω_j in which the solution at step $n + 1$ has already been calculated. The relaxed version of this algorithm of this algorithm can be written:

$$\begin{aligned}
-\Delta u_k^{n+1} - \omega^2 u_k^{n+1} &= f \quad \text{in } \Omega_k, \\
\alpha(\partial_{v_k}u_k^{n+1} + i\omega u_k^{n+1}) + (1 - \alpha)(\partial_{v_k}u_k^n + i\omega u_k^n) &= -\partial_{v_j}u_j^{n+1} + i\omega u_j^{n+1} \quad \text{on } \Sigma_{kj},\ j < k, \\
\alpha(\partial_{v_k}u_k^{n+1} + i\omega u_k^{n+1}) + (1 - \alpha)(\partial_{v_k}u_k^n + i\omega u_k^n) &= -\partial_{v_j}u_j^n + i\omega u_j^n \quad \text{on } \Sigma_{kj},\ j > k, \\
\partial_{v_k}u_k^{n+1} + i\omega u_k^{n+1} &= 0 \quad \text{on } \Sigma_{kk}.
\end{aligned} \tag{85}$$

Once again each local problem to be solved is of type $(P_{\mathcal{O}}^{f,g})$ so that the sequence $u^n = (u_k^n)$ is well defined in $\prod_k H^1(\Omega_k)$, provided that (27) holds and $-\partial_{v_k}u_k^0 + i\omega u_k^0 \in L^2(\partial\Omega_k)$ for each k.

In order to reinterpret this algorithm at interfaces, we use the decomposition $\Pi = \Pi^+ + \Pi^-$ and simply remark that the transmission condition on the Σ_{kj}'s are equivalent to

$$\alpha y^{n+1} + (1 - \alpha)y^n = \Pi^- x^{n+1} + \Pi^+ x^n + g. \tag{86}$$

Setting once again $r = 1/\alpha$ and after elimination of the y^n's, we obtain

$$x^{n+1} = (1 - r)x^n + r(S\Pi^- x^{n+1} + S\Pi^+ x^n - g), \tag{87}$$

which corresponds to a fixed point algorithm applied to the equation $\mathscr{P}_g$, rewritten as

$$x - rS\Pi^- x = x - r(x - S\Pi^+ x - g). \tag{88}$$

In the case $r = 1$, we recover the non relaxed version Gauss–Seidel algorithm

$$(I - A^-)x^{n+1} = A^+ x^n + g, \tag{89}$$

where $A^- = S\Pi^-, A^+ = S\Pi^+$ and $A = A^+ + A^-$. Remark that, going back to the matrix interpretation of Section 3.1, the structure of the matrices A^+, A^- are of the form

$$A^+ = \begin{bmatrix} 0 & A_{12} & A_{13} & \dots & A_{1K} \\ 0 & 0 & A_{23} & & A_{2K} \\ 0 & 0 & \ddots & \ddots & A_{3K} \\ \vdots & & \ddots & \ddots & \vdots \\ 0 & \dots & \dots & 0 & 0 \end{bmatrix}, \quad A^- = \begin{bmatrix} 0 & 0 & 0 & \dots & 0 \\ A_{21} & 0 & 0 & & 0 \\ A_{31} & A_{32} & \ddots & \ddots & 0 \\ \vdots & & \ddots & \ddots & \vdots \\ A_{K1} & \dots & \dots & A_{KK-1} & 0 \end{bmatrix}. \tag{90}$$

Matrix A^+ is upper triangular while matrix A^- is lower triangular. This establishes a clear analogy between the usual Gauss–Seidel method in linear algebra.

In the same way of what has been done for the Jacobi's algorithm we can substitute to u^n another lifting of x^n, let $\tilde{u}^n$, defined formally by (81). The explicit computation of the x^n_{jk} and $\tilde{u}_k$, is given by (83) and (84) with x^n_{jk} replaced by $x^{n\star}_{jk}$ in the second line of (83), where $x^{n\star}_{jk} = x^n_{jk}$ when $j > k$ and x^{n+1}_{jk} when $j < i$.

Remark 5. *Contrary to the Jacobi's algorithm, the Gauss–Seidel's algorithm is not parallelizable: the computation of the solution of each local subproblem must be solved accordingly to the ordering of the subdomains. On the other hand, one may expect a better convergence. We shall come back on this point in Section* 5.

3.3. General transmission operators

3.3.1. Introduction

As we shall see later, the convergence of the previous algorithms is not exponential. This is due to the fact that operator A admits a sequence of eigenvalues which converges to 1 in modulus (see Section 5). The purpose of this section is to generalize the iterative algorithm in the hope to get an exponential rate of convergence. The basic idea is to replace the quantities $\partial_{v_k} u_{k/\Sigma_{kj}} \pm i\omega u_k/\Sigma_{kj}$ by quantities

$$\partial_{v_k} u_{k/\Sigma_{kj}} \pm i\omega T_{jk}(u_{k/\Sigma_{kj}}), \tag{91}$$

where $T_{kj} = T_{jk}$ are operators acting on functions defined on $\partial\Omega_k \cap \partial\Omega_j$, having appropriate properties that will be specified later. The more interesting case will be the one where T_{jk} is some L^2-symmetric pseudo-differential operator of order 1. In that case both quantities $\partial_{v_k} u_k$ and $T_{jk} u_k$ belong to the same Sobolev space on $\partial\Omega_k \cap \partial\Omega_j$ so that the linear combinations (91) are in some sense more natural than the quantities $\partial_{v_k} u_k \pm i\omega u_k$. We shall see in Section 4 why such a property may lead to an exponential rate of convergence. Let us nevertheless remark that in practice this property will impose that the operators T_{jk} be non local. We shall come back to this point at the end of Section 4.

In what follows, if Γ denotes a Lipschitz-continuous manifold of dimension $n - 1$, we shall adopt the notation $(\cdot, \cdot)_{L^2(\Gamma)}$ for the scalar product in $L^2(\Gamma)$ and $\langle \cdot, \cdot \rangle_\Gamma$ for the duality product between $H^s(\Gamma)$ and $H^{-s}(\Gamma)$ for $|s| \leqslant 1/2$. For each $\{j, k\}$ such that $\partial\Omega_k \cap \partial\Omega_j$ is not empty, we introduce an operator T_{jk} satisfying:

$$T_{jk} = T_{kj} \in \mathcal{L}\big(H^{1/2}(\partial\Omega_k \cap \partial\Omega_j), H^{-(1/2)}(\partial\Omega_k \cap \partial\Omega_j), \big),$$

$$T_{jk} \quad \text{is injective.} \tag{92}$$

The identity operator is an example of such an operator. Moreover, the injectivity of the operators T_{jk} implies that our original problems (P_Ω^f) are equivalent to the following collection of local problems posed in Ω_k that we write formally:

$$-\Delta u_k - \omega^2 u_k = f \quad \text{in } \Omega_k,$$

$$\partial_{v_k} u_k + i\omega T_{jk}(u_{k/\Sigma_{kj}}) = -\partial_{v_j} u_j + i\omega T_{jk}(u_{j/\Sigma_{jk}}) \quad \text{on } \Sigma_{kj}, \; k \neq j, \tag{93}$$

$$\partial_{v_k} u_k + i\omega u_k = 0, \quad \text{on } \Sigma_{kk}.$$

In order to ensure the well-posedness of each local problem (93), we shall be led to make the following assumption on T_{jk}, (which more or less expresses the fact that it is symmetric and positive)

There exists $s \in [0, (1/2)]$ and Λ_{jk} an isomorphism from

$H^s(\partial\Omega_k \cap \partial\Omega_j)$ into $L^2(\partial\Omega_k \cap \partial\Omega_j)$ such that: $\tag{94}$

$T_{jk} = \Lambda_{jk}^* \Lambda_{jk}.$

In (94), s may depend on the operator T_{jk}, even it is not mentioned explicitly. The operator Λ_{jk}^*, adjoint of Λ_{jk}, is an isomorphism from $L^2(\partial\Omega_k \cap \partial\Omega_j)$ onto $H^{-s}(\partial\Omega_k \cap \partial\Omega_j)$. One deduces that T_{jk} is a symmetric positive one to one operator from $H^s(\partial\Omega_k \cap \partial\Omega_j)$ onto $H^{-s}(\partial\Omega_k \cap \partial\Omega_j)$. We have

$$\forall \varphi \in H^{1/2}(\Sigma_{kj}), \quad \langle T_{jk}\varphi, \varphi \rangle \Sigma_{kj} = \|\Lambda_{jk}\varphi\|_{L^2(\Sigma_{kj})}^2. \tag{95}$$

One can notice that the case of the identity operator corresponds to $s = 0$ and $\Lambda_{jk} = \Lambda_{jk}^* = I$. However, the interesting case, as said previously, will correspond to $s = 1/2$.

Remark 6. *In the sequel, it will be useful to set $T_{kk} = I$. We thus have $s = 0$ for T_{kk}.*

3.3.2. Well-posedness of the local problems

We can now consider the model problems:

$$-\Delta u_k - \omega^2 u_k = f_k \quad \text{in } \Omega_k,$$

$$\partial_{v_k} u_k + i\omega T_{jk}(u_{k/\Sigma_{kj}}) = g_{kj} \quad \text{on } \Sigma_{kj}. \tag{96}$$

Here we must be precise on the meaning of a solution of (96) which implies to raise some subtle difficulties in the sense one gives to the boundary condition. If we look for a u_k in $H^1(\Omega_k)$, then for $f \in L^2(\Omega_k)$, we shall also have $\Delta u_k \in L^2(\Omega_k)$ so that $\partial_{v_k} u_{k/\partial\Omega_k}$ belongs to $H^{-(1/2)}(\partial\Omega_k)$. It is known that the restriction of $\partial_{v_k} u_{k/\partial\Omega_k}$ to Σ_{kj}, namely: $\partial_{v_k} u_{k/\Sigma_{kj}}$ can only be defined as an element of a space which strictly contains $H^{-(1/2)}(\partial\Omega_k)$, namely

$$H_{00}^{1/2}(\Sigma_{kj})' \equiv \text{the dual space of } H_{00}^{1/2}(\Sigma_{kj}),$$

where we remind that $H_{00}^{1/2}(\Sigma_{kj})$ can be defined as the subspace of functions of $H^{1/2}(\Sigma_{kj})$ whose extension by 0 belongs to $H^{1/2}(\partial\Omega_k)$ (see [25] or [23] for equivalent definitions and various properties of these spaces). Therefore, the equality

$$\partial_{v_k} u_{k/\Sigma_{kj}} + i\omega T_{jk}(u_{k/\Sigma_{kj}}) = g_{kj}, \tag{97}$$

a priori holds in $H_{00}^{1/2}(\Sigma_{kj})'$. However, if g_{kj} belongs to $H^{-(1/2)}(\Sigma_{kj})$, we have, since $T_{jk} \in \mathcal{L}(H^s(\Sigma_{kj}), H^{-s}(\Sigma_{kj}))$ with $s \in [0, (1/2)]$

$$\partial_{v_k} u_{k/\Sigma_{kj}} \in H^{-(1/2)}(\Sigma_{kj}). \tag{98}$$

In such a case, one has the following property

$$\forall \varphi \in H^{1/2}(\partial\Omega_k), \quad \langle \partial_{v_k} u_k, \varphi \rangle_{\partial\Omega_k} = \sum_j \langle \partial_{v_k} u_{k/\Sigma_{kj}}, \varphi_{/\Sigma_{kj}} \rangle_{\Sigma_{kj}}, \tag{99}$$

due to the density in $H^{1/2}(\partial\Omega_k)$ of the space of functions in $H^{1/2}(\partial\Omega_k)$ whose restriction to each Σ_{kj} belongs to $H_{00}^{1/2}(\Sigma_{kj})$ [23,25]. Therefore (54) is equivalent to the variational problem:

Find $u_k \in H^1(\Omega_k)$ such that, $\forall v_k \in H^1(\Omega_k)$:

$$\int_{\Omega_k} \left(\nabla u_k \nabla v_k - \omega^2 u_k v_k \right) dx + i\omega \sum_j \langle T_{jk}(u_{k/\Sigma_{kj}}), \bar{v}_{k/\Sigma_{kj}} \rangle_{\Sigma_{kj}} = \int_{\Omega_k} f v_k \, dx + i\omega \sum_j \langle g_{kj}, \bar{v}_{k/\Sigma_{kj}} \rangle_{\Sigma_{kj}}. \tag{100}$$

Theorem 3. (i) *Assume that $f \in L^2(\Omega_k)$ and $g_{kj} \in H^{-(1/2)}(\Sigma_{kj})$ for any j, then (96) admits a unique solution $u_k \in H^1(\Omega_k)$.*
   (ii) *If $g_{kj} \in H^{-s}(\Sigma_{kj})$, the quantities:*

$$\left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}) \pm i\omega \Lambda_{kj}(u_{k/\Sigma_{kj}}),$$

belong to $L^2(\Sigma_{kj})$ and one has the identity for each k:

$$\sum_j \left\| \left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}) + i\omega \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right\|_{L^2(\Sigma_{kj})}^2 = \sum_j \left\| \left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}) - i\omega \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right\|_{L^2(\Sigma_{kj})}^2. \tag{101}$$

Proof. (i) It is easy to see that Fredholm's alternative applies to (100). We thus only have to check the uniqueness of the solution. We then consider $f = 0$ and $g_{kj} = 0$. Let us take $v_k = \bar{u}_k$ in (100). Taking the imaginary part of the resulting equality leads to

$$\omega \sum_j \left\| \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right\|_{L^2(\Sigma_{kj})}^2 = 0. \tag{102}$$

Therefore $\Lambda_{kj}(u_{k/\Sigma_{kj}}) = 0$ and by injectivity, $u_{k/\Sigma_{kj}} = 0$. This is true for all j, therefore $u_{k/\partial\Omega_k} = 0$. Going back to (100) with $f = g_{kj} = 0$, we get

$$\int_{\Omega_k} \left(\nabla u_k \nabla v_k - \omega^2 u_k v_k \right) dx = 0, \quad \forall v_k \in H^1(\Omega_k), \tag{103}$$

which means that $\Delta u_k + \omega^2 u_k = 0$ and $\partial_{v_k} u_k = 0$ on $\partial\Omega_k$. As we also have $u_{k/\partial\Omega_k} = 0$ we conclude that $u_k = 0$ by classical arguments [11, p. 211].
   (ii) Note that

$$\left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}) \pm i\omega \Lambda_{kj}(u_{k/\Sigma_{kj}}) = \left(\Lambda_{kj}^* \right)^{-1} \left(\partial_{v_k} u_{k/\Sigma_{kj}} \pm i\omega T_{kj}(u_{k/\Sigma_{kj}}) \right). \tag{104}$$

As $\partial_{v_k} u_{k/\Sigma_{kj}} + i\omega T_{kj}(u_{k/\Sigma_{kj}}) = g_{kj}$ is $H^{-s}(\Sigma_{kj})$ so is $\partial_{v_k} u_{k/\Sigma_{kj}} - i\omega T_{kj}(u_{k/\Sigma_{kj}}) = g_{kj} - 2i\omega T_{kj}(u_{k/\Sigma_{kj}})$. Furthermore, $(\Lambda_{kj}^*)^{-1} \in \mathscr{L}(H^{-s}(\Sigma_{kj}), L^2(\Sigma_{kj}))$, this implies that the left hand-side of (104) is in $L^2(\Sigma_{kj})$. Moreover, we have

$$\left\| \left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}) \pm i\omega \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right\|_{L^2(\Sigma_{kj})}^2 = \left\| \left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}) \right\|_{L^2(\Sigma_{kj})}^2 + \omega^2 \left\| \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right\|_{L^2(\Sigma_{kj})}^2$$

$$\pm 2\omega \Im \left(\left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}), \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right)_{L^2(\Sigma_{kj})}. \tag{105}$$

To conclude, it suffices to prove that $\sum_j \left(\left(\Lambda_{kj}^* \right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}), \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right)_{L^2(\Sigma_{kj})}$ is real. At first, we remark that

$$\left(\left(\Lambda_{kj}^* \right)^{-1} \partial_{v_k}(u_{k/\Sigma_{kj}}), \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right)_{L^2(\Sigma_{kj})} = \langle \partial_{v_k} u_k, u_k \rangle_{\Sigma_{kj}}. \tag{106}$$

Thanks to (99) and Green's formula, we get

$$\sum_j \left(\left(\Lambda_{kj}^*\right)^{-1}(\partial_{v_k} u_{k/\Sigma_{kj}}), \Lambda_{kj}(u_{k/\Sigma_{kj}}) \right)_{L^2(\Sigma_{kj})} = \langle \partial_{v_k} u_k, u_k \rangle_{\partial\Omega_k} = \int_{\Omega_k} \left(|\nabla u_k|^2 - \omega^2 |u_k|^2 \right) dx$$
$$= \text{some real quantity,} \tag{107}$$

which concludes the proof. $\square$

3.3.3. Reformulation at interfaces

We generalize in this section the content of Section 2.2. The new unknown at interfaces will be

$$x = (x_k), \quad x_{k/\Sigma_{kj}} = -\left(\Lambda_{kj}^*\right)^{-1}(\partial_{v_k} u_{k/\Sigma_{kj}}) + \mathrm{i}\omega\Lambda_{kj}(u_{k/\Sigma_{kj}}), \tag{108}$$

where we have assumed that

$$\partial_{v_k} u_{k/\Sigma_{kj}} \in H^{-s}(\Sigma_{kj}), \quad \forall k, j, \tag{109}$$

so that

$$x \in V. \tag{110}$$

We shall now give a new formulation at interfaces of the transmission problem (93). It will fit the same abstract framework as that of Section 2 via a new definition of operators E, S and A.

Definition of E, S and A. We define

$$E \in \mathscr{L}\left(V, \prod_{k=1}^{K} H^1(\Omega_k)\right), \tag{111}$$

$$e = (e_k) \in \prod_{k=1}^{K} H^1(\Omega_k) = E(y) \Longleftrightarrow \begin{cases} -\Delta e_k - \omega^2 e_k = 0, \\ \left(\Lambda_{kj}^*\right)^{-1}(\partial_{v_k} e_{k/\Sigma_{kj}}) + \mathrm{i}\omega\Lambda_{kj}(e_{k/\Sigma_{kj}}) = y_{kj} \quad \text{on } \Sigma_{kj}. \end{cases} \tag{112}$$

The fact that the operator E is well defined is a direct consequence of Theorem 3. By compactness of the embedding $H^1(\Omega_k) \hookrightarrow H^v(\Omega_k)$, $v < 1$, E is compact from V into $\prod_k H^v(\Omega_k)$. We have a more general result.

Theorem 4. *E is a compact operator from V into $\prod_k H^v(\Omega_k)$ with $v = 1$ in (94) if $s < 1/2$ and $v < 1$ if not.*

Proof. We only need to look at the case $s < 1/2$. Let y^n a bounded sequence of V and $e^n = E(y^n)$. Up to the extraction of a subsequence, there exists some e such that

$$e^n \rightarrow e \qquad \text{in } \prod_k L^2(\Omega_k), \quad \text{strongly,}$$
$$\nabla e^n \rightarrow \nabla e \quad \text{in } \prod_k L^2(\Omega_k), \quad \text{weakly.} \tag{113}$$

By a trace theorem, $e^n_{/\Sigma_{kj}}$ bounded in $H^{1/2}(\partial\Sigma_{kj})$. Since $s < 1/2$, we may assume

$$e^n_{/\Sigma_{kj}} \rightarrow e_{/\Sigma_{kj}} \quad \text{in } H^s(\Sigma_{kj}), \quad \text{strongly.} \tag{114}$$

By Green's theorem, one has in each Ω_k, $(d_k^n = e_k^n - e, d_{kj}^n = d_{k/\Sigma_{kj}}^n)$

$$\int_{\Omega_k} |\nabla d_k^n|^2 \, dx = \omega^2 \int_{\Omega_k} |d_k^n|^2 \, dx + \langle \partial d_k^n, (d_{k/\Omega_k}^n) \rangle_{\partial\Omega_k}. \tag{115}$$

But we have:

$$\langle \partial d_k^n, d_{k/\partial\Omega_k}^n \rangle_{\partial\Omega_k} = \sum_j \langle \partial_{\nu_k} d_{kj}^n, d_{kj}^n \rangle_{\Sigma_{kj}} = \sum_j \left(\left(\Lambda_{kj}^* \right)^{-1} \partial_{\nu_k} d_{kj}^n, \Lambda_{kj} d_{kj}^n \right)_{L^2(\Sigma_{kj})}$$

$$= -i\omega \sum_j \left\| \Lambda_{kj} d_{kj}^n \right\|_{L^2(\Sigma_{kj})}^2 + \sum_j (\Lambda_{kj} d_{kj}^n, y_{kj}^n)_{L^2(\Sigma_{kj})}. \tag{116}$$

Property (114) implies $\Lambda_{kj} d_k^n \to 0$ in $L^2(\Sigma_{kj})$ and therefore

$$(\Lambda_{kj} d_{kj}^n, y_{kj}^n)_{L^2(\Sigma_{kj})} \to 0, \tag{117}$$

we conclude that

$$\int_{\Omega_k} \left| \nabla(e_k^n - e_k) \right|^2 \, dx = \int_{\Omega_k} \left| \nabla d_k^n \right|^2 \, dx \to 0, \tag{118}$$

which ends the proof. □

With the help of operator E we can define $S \in \mathscr{L}(V, V)$ by

$$x = Sy \Longleftrightarrow x_{k/\Sigma_{kj}} = x_{kj} = -\left(\Lambda_{kj}^* \right)^{-1} (\partial e_{k/\Sigma_{kj}}) + i\omega\Lambda_{kj}(e_{k/\Sigma_{kj}}), \quad e = E(y). \tag{119}$$

By Theorem 3-(ii), we know that S is a unitary operator $S^*S = SS^* = I$. Next, we define $A \in \mathscr{L}(V, V)$

$$A = S\Pi. \tag{120}$$

Operator A is clearly contractive and, following the proof of Section 2, one proves that $I - A$ is injective, thanks to the injectivity of Λ_{kj}.

In order to formulate the problem satisfied by x_k, we introduce $v = (v_k) \in \prod_k H^1(\Omega_k)$, where v_k is solution of the local problem

$$-\Delta v_k - \omega^2 v_k = f_k,$$
$$\left(\Lambda_{kj}^* \right)^{-1} (\partial_{\nu_k} v_{k/\Sigma_{kj}}) + i\omega\Lambda_{kj}(v_{k/\Sigma_{kj}}) = 0 \quad \text{on } \Sigma_{kj}, \tag{121}$$

and let us set

$$g_{kj} = -\left(\Lambda_{kj}^* \right)^{-1} (\partial_{\nu_k} v_{k/\Sigma_{kj}}) + i\omega\Lambda_{kj}(v_{k/\Sigma_{kj}}). \tag{122}$$

Following the proof of Theorem 2, it is straightforward to establish the following result.

Theorem 5.
  (i) *If $u = (u_k)$ is solution of (93), then $x \in V$ defined by (108) is solution of*

$$\mathscr{P}_g : (I - A)x = g. \tag{123}$$

  (ii) *Reciprocally, if x is solution of $(\mathscr{P}_g)$, then $u = v + E(\Pi x)$ is solution of (93).*

3.3.4. The domain decomposition algorithms

Once one has identified the spaces V and the operators E, S and A, it is easy to generalize what has been done in Section 3.2.1. We define the relaxed Jacobi and Gauss–Seidel Algorithms by replacing all the quantities $i\omega u_k$ by $i\omega T_{jk} u_k$ in (72) and (85). These algorithms corresponds to the iterative processes (76) and (87).

Finally and if we choose $\tilde{u}^n$ as the approximate value of the solution u, all our algorithms can be described in the general form:

$$\check{x}_{kj}^0 = \Lambda_{kj}^* x_{kj}^0, \quad \text{given in } H^{-s}(\Sigma_{kj}),$$

for $n = 1, \ldots$

    for $k = 1, \ldots, K$

$$\text{solve} \quad \begin{cases} \omega^2 \tilde{u}_k^{n+1} + \Delta \tilde{u}_k^{n+1} = f \quad \text{in } \Omega_k, \\ \partial_{\nu_k} \tilde{u}_k^{n+1} + i\omega T_{jk}\left(\tilde{u}_k^{n+1}{}_{/\Sigma_{kj}} \right) = \check{x}_{jk}^{n^\star} \quad \text{on } \Sigma_{kj},\ j \neq k \\ \partial_{\nu_k} \tilde{u}_k^{n+1} + i\omega \tilde{u}_k^{n+1} = 0 \quad \text{on } \Sigma_{kk}. \end{cases} \tag{124}$$

    update $\quad \check{x}_{kj}^{n+1} = (1-r)\check{x}_{kj}^n + r\left(-\partial_{\nu_k} \tilde{u}_k^{n+1} + i\omega T_{jk}\left(\tilde{u}_k^{n+1}{}_{/\Sigma_{kj}} \right) \right),$

with $\check{x}_{jk}^{n^\star} = \check{x}_{jk}^n$, except for the Gauss–Seidel algorithm for which $\check{x}_{jk}^{n^\star} = \check{x}_{jk}^{n+1}$ when j is less than k. This shows that it is not necessary to use the operator Λ_{kj} but only T_{jk}.

4. Convergence analysis

The aim of what follows is to establish some results of convergence of the two relaxed algorithms (Jacobi and Gauss–Seidel) introduced in Section 3. For the non relaxed case, which is technically more complicated, we refer to [15]. We will work only with the abstract form of these algorithms, so that our proof is valid for both the first versions proposed in Section 3.2 and the second one with the general interface operators considered in Section 3.3.

We begin by introducing some notations. We first define the error at the interface as the quantities

$$\varepsilon^n = x^n - x \in V, \tag{125}$$

and the two errors committed on the unknown u itself, depending whether one considers u^n (formula (78)) or $\tilde{u}^n$ (formula (81)) for the approximation of u

$$e^n = u^n - u \in \prod_{k=1}^K H^1(\Omega_k),$$
$$\tilde{e}^n = \tilde{u}^n - u \in \prod_{k=1}^K H^1(\Omega_k). \tag{126}$$

The sequence ε^n satisfies the equation

$$\varepsilon^{n+1} = (1-r)\varepsilon^n + rA\varepsilon^n, \tag{127}$$

in the case of the Jacobi's algorithm and

$$\varepsilon^{n+1} = (1-r)\varepsilon^n + r(S\Pi^- \varepsilon^{n+1} + S\Pi^+ \varepsilon^n), \tag{128}$$

in the case of the Gauss–Seidel's algorithm. The two sequences e^n and $\tilde{e}^n$ are then deduced from ε^n by

$$e^{n+1} = (1-r)E(S^* \varepsilon^n) + rE(\Pi \varepsilon^n) \quad \text{for Jacobi,}$$
$$\phantom{e^{n+1}} = (1-r)E(S^* \varepsilon^n) + rE(\Pi^+ \varepsilon^n + \Pi^- \varepsilon^{n+1}) \quad \text{for Gauss–Seidel,} \tag{129}$$
$$\tilde{e}^{n+1} = E(\Pi \varepsilon^n).$$

In the sequel, we are going to obtain two types of results:

- A convergence result for the general situation: we only obtain the convergence of ε^n to 0 in the weak topology of V. However, thanks to the compactness properties of E (cf. Theorem 4) we shall obtain what

we really want, namely the strong convergence of u^n (or $\tilde{u}^n$) to u at least in L^2. (We shall get even better – see Section 4.2).
- A convergence result in a particular case for which we shall have strong convergence to 0 for ε^n. Moreover, this convergence will be exponentially fast and will imply the exponential decay of the errors e^n (or $\tilde{e}^n$) in the H^1 topology of each sub-domain. We shall see in Section 4.2 that our proof will require a particular geometrical property of the partition of Ω as well as an adequate choice for the operators T_{jk}.

4.1. A first general result

The main tool of the analysis consists of a priori energy estimates that we deduce of two identities which are the object of the next lemma. It will be useful to introduce the following norm in V ($r \in]0, 1]$)

$$\|x\|_r^2 = (1 - r)\|P_\Sigma^- x\|^2 + \|P_\Gamma x\|^2 + \|P_\Sigma^+ x\|^2. \tag{130}$$

Note that this norm is equivalent to the usual norm

$$(1 - r)^{\frac{1}{2}}\|x\| \leqslant \|x\|_r \leqslant \|x\|. \tag{131}$$

We also define the operator

$$J_r = (I - rS\Pi^-)^{-1} \in \mathcal{L}(V). \tag{132}$$

(Note that $(I - rS\Pi^-)$ is an isomorphism in V: indeed, in the matrix representation of Section 3, it is block lower triangular, the triangular blocks being identity operators).

Lemma 5.
(i) *Jacobi's algorithm*: *The sequence ε^n satisfying* (127) *satisfies the identity*

$$\left\|\varepsilon^{n+1}\right\|^2 = \|\varepsilon^n\|^2 - r(1 - r)\|(I - A)\varepsilon^n\|^2 - r\|P_\Gamma \varepsilon^n\|^2. \tag{133}$$

(ii) *Gauss–Seidel's algorithm*: *The sequence ε^n satisfying* (128) *satisfies the identity*

$$\left\|\varepsilon^{n+1}\right\|_r^2 = \|\varepsilon^n\|_r^2 - r(1 - r)\|J_r(I - A)\varepsilon^n\|^2 - r\|P_\Gamma \varepsilon^n\|^2. \tag{134}$$

Proof. (i) We consider the sequence ε^n satisfying (127). We have, using $A = S\Pi$

$$\left\|\varepsilon^{n+1}\right\|^2 = (1 - r)^2\|\varepsilon^n\|^2 + r^2\|S\Pi\varepsilon^n\|^2 + 2r(1 - r)\Re(S\Pi\varepsilon^n, \varepsilon^n). \tag{135}$$

On the other hand, we use the identity

$$\|(I - S\Pi)\varepsilon^n\|^2 = \|\varepsilon^n\|^2 + \|S\Pi\varepsilon^n\|^2 - 2\Re(S\Pi\varepsilon^n, \varepsilon^n),$$

to eliminate $\Re(S\Pi\varepsilon^n, \varepsilon^n)$ in (135). This leads to

$$\left\|\varepsilon^{n+1}\right\|^2 = (1 - r)\|\varepsilon^n\|^2 + r\|S\Pi\varepsilon^n\|^2 - r(1 - r)\|(I - A)\varepsilon^n\|^2. \tag{136}$$

Using Lemma 2 and the fact that S is an isometry, we get

$$\|S\Pi\varepsilon^n\|^2 = \|\Pi\varepsilon^n\|^2 = \|\varepsilon^n\|^2 - \|P_\Gamma \varepsilon^n\|^2. \tag{137}$$

Plugging (137) in (135) then leads to (133).

(ii) Let us now consider a sequence ε^n satisfying (127). We have

$$\left\|\varepsilon^{n+1}\right\|^2 = (1 - r)^2\|\varepsilon^n\|^2 + r^2\left\|S(\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1})\right\|^2 + 2r(1 - r)\Re\left(\varepsilon^n, S(\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1})\right). \tag{138}$$

Applying the identity $2\Re(x, Sy) = \|x\|^2 + \|y\|^2 - \|x - Sy\|^2$, to $x = \varepsilon^n$ and $y = \Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1}$, we get, using also the fact that S is an isometry

$$\left\|\varepsilon^{n+1}\right\|^2 = (1-r)^2\|\varepsilon^n\|^2 + r^2\left\|\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1}\right\|^2 + r(1-r)\left(\|\varepsilon^n\|^2 + \|\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1}\|^2\right)$$
$$- r(1-r)\left\|\varepsilon^n - S\left(\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1}\right)\right\|^2, \tag{139}$$

or, equivalently

$$\left\|\varepsilon^{n+1}\right\|^2 = (1-r)\|\varepsilon^n\|^2 + r\left\|\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1}\right\|^2 - r(1-r)\left\|\varepsilon^n - S\left(\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1}\right)\right\|^2. \tag{140}$$

Since, by Lemma 3,

$$\left\|\Pi^+x + \Pi^-y\right\|^2 = \left\|\Pi^+x\right\|^2 + \left\|\Pi^-y\right\|^2 = \left\|P_\Sigma^+x\right\|^2 + \left\|P_\Sigma^-y\right\|^2, \tag{141}$$

we obtain, introducing $z^n = \varepsilon^n - S(\Pi^+\varepsilon^n + \Pi^-\varepsilon^{n+1})$

$$\left\|\varepsilon^{n+1}\right\|^2 = (1-r)\|\varepsilon^n\|^2 + r\left\|P_\Sigma^+\varepsilon^n\right\|^2 + r\left\|P_\Sigma^-\varepsilon^{n+1}\right\|^2 - r(1-r)\|z^n\|^2. \tag{142}$$

We then notice that, by definition of the norm $\|\cdot\|_r$

$$\left\|\varepsilon^{n+1}\right\|^2 - r\left\|P_\Sigma^-\varepsilon^{n+1}\right\|^2 = \left\|\varepsilon^{n+1}\right\|_r^2 (1-r)\|\varepsilon^n\|^2 + r\left\|P_\Sigma^+\varepsilon^n\right\|^2 = \|\varepsilon^n\|_r^2 - r\|P_\Gamma\varepsilon^n\|^2, \tag{143}$$

so that (142) can be rewritten

$$\left\|\varepsilon^{n+1}\right\|_r^2 = \|\varepsilon^n\|_r^2 - r(1-r)\|z^n\|^2 - r\|P_\Gamma\varepsilon^n\|^2, \tag{144}$$

which is nothing but (134) if we show that $z^n = J_r(I-A)\varepsilon^n$. But from (128) we deduce

$$\varepsilon^{n+1} = \varepsilon^n - rz^n. \tag{145}$$

Eq. (128) also shows that ε^{n+1} is related to ε^n by

$$(I - rS\Pi^-)\varepsilon^{n+1} = (1-r)\varepsilon^n + rS\Pi^+\varepsilon^n. \tag{146}$$

Therefore, multiplying (145) by $(I - rS\Pi^-)$ we get

$$(1-r)\varepsilon^n + rS\Pi^+\varepsilon^n = \varepsilon^n - rS\Pi^-\varepsilon^n - r(I - rS\Pi^-)z^n, \tag{147}$$

from which

$$(I - rS\Pi^-)z^n = (I - A)\varepsilon^n, \tag{148}$$

what we wanted to show. $\square$

A direct consequence of this lemma is:

Theorem 6. *Assume that $0 < r < 1$, the sequence ε^n given by* (127) *(or* (128)*) converges weakly to 0 in the space V and we have*

$$u^n \to u \quad \text{strongly in } \prod_{k=1}^K H^v(\Omega_k),$$
$$\tilde{u}^n \to u \quad \text{strongly in } \prod_{k=1}^K H^v(\Omega_k), \tag{149}$$

where $v = 1$ if $s < 1/2$ in (94) *for all T_{jk} and $v < 1$ if not.*

Proof. We only give the proof for the Jacobi's algorithm using (133). The proof for the Gauss–Seidel's algorithm via (134) is almost identical thanks to the fact that $\|\cdot\|_r$ is a norm on V equivalent to $\|\cdot\|$. Identity (133) implies in particular that

ε^n is bounded in V,   ($\|\varepsilon^n\|$ decreases),

$$\sum_{n=0}^{+\infty} \|(I - A)\varepsilon^n\|^2 < +\infty. \tag{150}$$

Therefore, up to the extraction of a subsequence still denoted by ε^n, we know that there exists $\varepsilon \in V$ such that

$$\varepsilon^n \to \varepsilon \quad \text{weakly in } V. \tag{151}$$

Moreover, from (150) we also know that

$$(I - A)\varepsilon^n \to 0 \quad \text{strongly in } V. \tag{152}$$

Therefore, by linearity of A, we get

$$(I - A)\varepsilon = 0, \tag{153}$$

which yields $\varepsilon = 0$ since $I - A$ is injective. Note also that, since 0 is the only possible limit, the whole sequence converges.

The end of the proof is nothing but a straightforward consequence of the compactness properties of the operator E (given in Theorem 4) and of the two equalities (129). □

Remark 7. *From our derivation of the a priori energy estimates, it can be that the equalities* (133) *and* (134) *becomes inequalities if we replace the hypothesis* $\|Sx\| = \|x\|$ *by* $\|Sx\| \leqslant \|x\|$ *(S contractive instead of isometric). Thus, it is possible to extend our result to the damped Helmholtz equations as considered in* [19] *for example.*

4.2. A case of geometric convergence

In this section, we shall assume that the partition $(\Omega_k)_k$ of Ω has the following geometrical property.

For each (j, k), $\partial\Omega_k \cap \partial\Omega_j$ is either empty either a closed manifold of dimension $N - 1$. $\tag{154}$

The second hypothesis lies on the value of s for the operators T_{jk}, $j \neq k$ ($s = 1/2$)

$$T_{jk} = \Lambda_{jk}^* \Lambda_{jk}, \quad \text{with } \Lambda_{jk} \text{ isomorphism from } H^{\frac{1}{2}}(\partial\Omega_k \cap \partial\Omega_j) \text{ into } L^2(\partial\Omega_k \cap \partial\Omega_j). \tag{155}$$

In such a situation, we have the following fundamental result.

Lemma 6. *Under both assumptions* (154) *and* (155) $I - A$ *is one to one in* V.

Proof. As $I - A$ is always injective, we only have to prove that it is onto. Let g be given in V, we look for an x such that

$$(I - A)x = g. \tag{156}$$

If such an x exists, then $u = (u_k) = E(x)$ satisfies in each Ω_k:

$$u_k \in H^1(\Omega_k), \quad -\Delta u_k - \omega^2 u_k = 0 \quad \text{in } \Omega_k$$

$$\left(\Lambda_{kj}^*\right)^{-1}(\partial_{\nu_k} u_{k/\Sigma_{kj}}) + i\omega\Lambda_{kj}(u_{k/\Sigma_{kj}}) = -\left(\Lambda_{kj}^*\right)^{-1}(\partial_{\nu_j} u_{j/\Sigma_{jk}}) + i\omega\Lambda_{kj}(u_{j/\Sigma_{jk}}) + g_{kj} \quad \text{on } \Sigma_{kj}, \; k \neq j, \tag{157}$$

$$\partial_{\nu_k} u_k + i\omega u_k = g_{kk} \quad \text{on } \Sigma_{kk},$$

which is equivalent to

$$u_k \in H^1(\Omega_k), \quad -\Delta u_k - \omega^2 u_k = 0 \quad \text{in } \Omega_k,$$

$$\partial_{v_k} u_k + \partial_{v_j} u_j = \Lambda_{kj}^* \frac{g_{kj} + g_{jk}}{2} \quad \text{on } \Sigma_{kj}, \ k < j,$$

$$(u_k - u_j) = (\Lambda_{kj})^{-1} \frac{g_{kj} - g_{jk}}{2i\omega} \quad \text{on } \Sigma_{kj}, \ k < j, \tag{158}$$

$$\partial_{v_k} u_k + i\omega u_k = g_{kk} \quad \text{on } \Sigma_{kk}.$$

Reciprocally, if there exists an $u \in H^1(\Omega_k)$ solution of (158), one constructs a solution of $(I - A)x = g$ via

$$x = (x_k), \quad x_{k/\Sigma_{kj}} = -\left(\Lambda_{kj}^*\right)^{-1} (\partial_{v_k} u_{k/\Sigma_{kj}}) + i\omega \Lambda_{kj}(u_{k/\Sigma_{kj}}). \tag{159}$$

Let us define

$$\varphi_{kj} = -\varphi_{jk} = (\Lambda_{kj})^{-1} \frac{g_{kj} - g_{jk}}{2i\omega} \in H^{1/2}(\Sigma_{kj}), \tag{160}$$

and let us assume for a while that one can find a $u^* \in \prod_k H^1(\Omega_k)$ with $\Delta u_k^* \in L^2(\Omega_k)$,

$$u_{k/\Sigma_{kj}}^* - u_{j/\Sigma_{jk}}^* = \varphi_{kj}, \qquad \partial_{v_k} u_{k/\Sigma_{kj}}^* + \partial_{v_j} u_{j/\Sigma_{jk}}^* \in H^{-(1/2)}(\Sigma_{kj}), \quad k \neq j, \tag{161}$$

then, setting

$$u = u^* + w, \tag{162}$$

function w must satisfy

$$w \in H^1(\Omega), \quad -\Delta w_k - \omega^2 w_k = \Delta u_k^* + \omega^2 u_k^* \quad \text{in } \Omega_k,$$

$$\partial_{v_k} w_k + \partial_{v_j} w_j, = \tilde{g}_{kj} \quad \text{on } \Sigma_{kj}, \ k \neq j, \tag{163}$$

$$\partial_{v_k} w_k + i\omega w_k = \tilde{g}_{kk} \quad \text{on } \Sigma_{kk},$$

with

$$\tilde{g}_{kk} = g_{kk} - \partial_{v_k} u_{k/\Sigma_{kk}}^* - i\omega u_{k/\Sigma_{kk}}^*,$$

$$\tilde{g}_{kj} = \Lambda_{kj}^* \frac{g_{kj} + g_{jk}}{2} - \partial_{v_k} u_{k/\Sigma_{kj}}^* - \partial_{v_j} u_{j/\Sigma_{jk}}^*. \tag{164}$$

The variational formulation of this problem is

$$\text{Find } w \in H^1(\Omega) \text{ such that, } \forall v \in H^1(\Omega) : \int_\Omega (\nabla w \nabla v - \omega^2 wv)\, dx + i\omega \int_\Gamma wv\, d\sigma$$

$$= \sum \int_{\Omega_k} \left(\Delta u_k^* + \omega^2 u_k^*\right) v_k\, dx + \sum_k \sum_j \langle \tilde{g}_{kj}, \bar{v}_{/\Sigma_{kj}} \rangle_{\Sigma_{kj}}. \tag{165}$$

Problem (165) is a nice variational problem in $H^1(\Omega)$ since the $\tilde{g}_{kj} \in H^{-(1/2)}(\Sigma_{kj})$. The existence of w is easily obtained by Fredholm's alternative. The problem is thus reduced to the construction of u^*. For $k \leqslant j$, we define u_{kj} as the solutions in $H^1(\Omega_k)$ of the following elliptic problem:

$$-\Delta u_{kj} + u_{kj} = 0 \quad \text{in } \Omega_k,$$

$$u_{kj/\Sigma_{kj}} = \tilde{\varphi}_{kj} \quad \text{on } \partial\Omega_k \tag{166}$$

with

$$\tilde{\varphi}_{kj} = \varphi_{kj} \text{ on } \Sigma_{kj}, \qquad \tilde{\varphi}_{kj} = 0 \text{ on } \Sigma_{k\ell}, \quad \ell \neq j. \tag{167}$$

As all the manifolds Σ_{kj} are assumed to be closed, all the $\tilde{\varphi}_{kj}$ are functions of $H^{1/2}(\Omega_k)$ and the existence of u_{kj} is simply obtained using Lax–Milgram theory and the trace theorem. Moreover, since Σ_{kj} is closed

$$\partial_{v_k} u_{kj/\Sigma_{kj}} \in H^{-(1/2)}(\Sigma_{kj}), \quad \partial_{v_k} u_{kj/\Sigma_{kk}} \in H^{-(1/2)}(\Sigma_{kk}). \tag{168}$$

Then it is not difficult to check that a good candidate for u^* is

$$u^* = \sum_{k<j} u_{kj},\tag{169}$$

and the proof of Lemma 6 is complete. $\quad\square$

Remark 8. *The theory of potential*, [10], *gives us a more explicit formulae for* $(I - A)^{-1}g$ *(however restricted to the case of constant coefficients). Indeed, one can verify that*

$$x_{kj} = -(\Lambda_{kj}^*)^{-1}\left(\partial_{v_k}(e_k + e_0)_{/\Sigma_{kj}}\right) + \mathrm{i}\omega\Lambda_{kj}(e_k + e_0)_{/\Sigma_{kj}},\tag{170}$$

where

$$
e(x) = \sum_{k}\sum_{j<k}\left[\int_{\Sigma_{kj}}\Phi(x,y)\frac{1}{2}\Lambda_{jk}^*(g_{kj} + g_{jk})(y)\,\mathrm{d}\sigma(y)\right.
$$
$$
\left. - \int_{\Sigma_{kj}}\partial_{v(y)}\Phi(x,y)\frac{1}{2\omega}\Lambda_{jk}^{-1}(g_{kj} - g_{jk})(y)\,\mathrm{d}\sigma(y)\right],\tag{171}
$$

where $\Phi(x,y)$ *is a Green's function for the Helmholtz equation and* e_0 *is the solution of*

$$-\omega^2 e_0 - \Delta e_0 = 0,\quad \text{in } \Omega$$
$$\partial_v e_0 + \mathrm{i}\omega e_0 = g_{kk} - \partial_v e - \mathrm{i}\omega e\quad \text{on } \Gamma.\tag{172}$$

As a consequence of Lemma 6 the following result:

Theorem 7 (Geometric convergence). *Under both assumptions* (154) *and* (155), *for each value* $r \in]0,1[$, *there exists a real number* $\tau \in]0,1[$ *such that*

$$\|\varepsilon^n\|_V \leqslant C\tau^n.\tag{173}$$

Consequently, we also have

$$\|u - u^n\|_{\Pi_k(H^1(\Omega_k))} \leqslant C\tau^n,\quad \|u - \tilde{u}^n\|_{\Pi_k(H^1(\Omega_k))} \leqslant C\tau^n.\tag{174}$$

Proof. We only give the proof in the case of Jacobi's algorithm. The proof for Gauss–Seidel is almost the same. Since $(I - A)$ is an isomorphism in V, there exists a $\delta > 0$ such that

$$\|(I - A)x\| \geqslant \delta\|x\|,\tag{175}$$

(note that $\delta \leqslant 1$, since $I - A$ is contractive). Therefore from identity (133), we get

$$\|\varepsilon^{n+1}\|^2 \leqslant (1 - r(1 - r)\delta^2)\|\varepsilon^n\|^2,\tag{176}$$

which gives (173) with $C = \|\varepsilon^0\|$ and $\tau = \sqrt{1 - r(1 - r)\delta^2}$. The rest of the proof is a consequence of the continuity of E. $\quad\square$

Remark 9. *In our proof, the best (i.e., the smallest) value of* τ *is given by* $r = 1/2$ *which maximizes* $r(1 - r)$. *We shall come back on this point in* Section 5.

In order to be complete, it remains to give examples of operators T_{jk} satisfying (155). The most natural one consists of fractional powers of the Laplace–Beltrami operator. More precisely, assuming that Σ_{kj} is a

sufficiently regular manifold, we denote by Δ_{jk} the Laplace–Beltrami operator on Σ_{kj} [33, tome 1, pp. 137–139], i.e., the self-adjoint operator in $L^2(\Sigma_{kj})$ associated to the bilinear form with domain $H^1(\Sigma_{kj})$:

$$a_{jk}(\varphi, \psi) = \int_{\Sigma_{kj}} \nabla_\Gamma \varphi \cdot \nabla_\Gamma \psi \, \mathrm{d}\sigma, \tag{177}$$

where ∇_Γ denotes the usual tangential gradient operator on Σ_{kj}. It is then well known that if α and β are two positive real numbers, one can define the positive selfadjoint operator in $L^2(\Sigma_{kj})$ with domain $H^1(\Sigma_{kj})$

$$T_{jk} = \left(\alpha I - \frac{\beta}{\omega^2} \Delta_{jk} \right)^{1/2}. \tag{178}$$

It is well known that T_{jk} can be extended in a unique way as an isomorphism from $H^{1/2}(\Sigma_{kj})$ onto $H^{-(1/2)}(\Sigma_{kj})$ as soon as α and β are strictly positive. One then satisfies (155) with

$$\Lambda_{jk} = \left(\alpha I - \frac{\beta}{\omega^2} \Delta_{jk} \right)^{1/4}. \tag{179}$$

As an example, let us consider the 2D case where Σ_{kj} is the circle of radius a, that we can parameterize by the polar angle θ. In such a case the operator T_{jk} given by (178) is diagonalized in the orthonormal basis of $L^2(\Sigma_{kj})$ defined by

$$e_n(\theta) = (2\pi a)^{1/2} \, \mathrm{e}^{in\theta}, \quad n \in Z, \tag{180}$$

and we have

$$T_{jk} e_n = \left(\alpha + \beta \frac{n^2}{\omega^2 a^2} \right) e_n. \tag{181}$$

One can note that the operator defined by (178) is non local (it is a convolution-type operator) and will produce full matrices after space discretization (cf. Section 6). A possible compromise with purely local would consist in considering an integral operator of the form (δ denoting some strictly positive parameter)

$$\langle T_{jk} \varphi, \psi \rangle = \alpha \int_{\Sigma_{jk}} \varphi(x) \bar{\psi}(x) \, \mathrm{d}\sigma(x)$$

$$+ \beta \int_{\Sigma_{kj}} \int_{\Sigma_{kj}} G(x-y)\chi_\delta(|x-y|) \, (\mathrm{curl}\varphi \wedge v)(y) \, (\mathrm{curl}\bar{\psi} \wedge v)(x) \, \mathrm{d}\sigma(x) \, \mathrm{d}\sigma(y), \tag{182}$$

where G is for instance the fundamental solution of the Laplace operator in R^N

$$G(x) = -\frac{1}{2\pi} \log|x|, \quad N = 2, \qquad G(x) = \frac{1}{4\pi|x|}, \quad N = 3, \tag{183}$$

and $\chi_\delta(r)$, $r > 0$ a smooth positive cut-off function satisfying:

$$\chi_\delta(r) = 1, \quad \text{if } r < \delta,$$

$$\chi_\delta(r) = 0, \quad \text{if } r > 2\delta, \tag{184}$$

whose role is to "localize" the operator T_{jk} (the "local" character of T_{jk} then depending on the value of δ). Such a choice has not yet be tested up to our knowledge. Our conjecture is that such an operator possesses the good properties ensuring exponential convergence: the important fact is that the kernel G has the correct singularity for $x = 0$.

Remark 10. *The choice of the operator T_{jk} defined by* (178) *(Δ_{jk} being associated to* (177) *in $H^1(\Sigma_{kj})$) can also be made even in the case where Σ_{kj} is not a closed manifold (the Laplace–Beltrami operator then corresponds to a "Neumann condition" on $\partial\Sigma_{kj}$). Although our proof of exponential convergence is no longer valid, numerical experiments* [21] *show that this improves the convergence by comparison with the case $T_{jk} = I$.*

5. Determination of the rate of convergence in a simple case

5.1. Presentation of the example, notations

To get a better insight on the rates of convergence of the method, we propose to analyze a very simple 2D example. We consider the square $\Omega =]-L/2, L/2[\times]0, L[$ decomposed into two rectangles similar in size Ω_1 and Ω_2, with only one interface (cf. Fig. 2). The DDM is applied to the solution of the Helmholtz equation in Ω with our usual conditions $(\partial_\nu + i\omega)u = 0$ at left and right sides and a Neumann condition at top and bottom sides.

We shall use in this section the presentation of the algorithm we gave in Section 3.3.4. Hence, we construct two sequences $\tilde{u}_1^n$ and $\tilde{u}_2^n$ which approximate (u_1, u_2), restrictions to Ω_1 and Ω_2 of u. The equations for $\tilde{u}_1^{n+1}$ and $\tilde{u}_2^{n+1}$ are, according to (124)

$$-\omega^2 \tilde{u}_k^{n+1} - \Delta \tilde{u}_k^{n+1} = f_k, \quad k = 1, 2,$$

$$\partial_y \tilde{u}_k^{n+1}(x, 0) = \partial_y \tilde{u}_k^{n+1}(x, L) = 0, \quad k = 1, 2. \tag{185}$$

$$\left(-\partial_x \tilde{u}_1^{n+1} + i\omega \tilde{u}_1^{n+1} \right)\left(-\tfrac{L}{2}, y \right) = 0,$$

$$\left(\partial_x \tilde{u}_2^{n+1} + i\omega \tilde{u}_2^{n+1} \right)\left(+\tfrac{L}{2}, y \right) = 0. \tag{186}$$

The coupling between two successive iterations n and $n + 1$ is made on the interface $x = 0$ (we refer once again to Section 3.3.4)

$$(\Lambda^*)^{-1}\partial_x \tilde{u}_1^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{u}_1^{n+1}{}_{/x=0} = x_{21}^n, \quad -(\Lambda^*)^{-1}\partial_x \tilde{u}_2^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{u}_2^{n+1}{}_{/x=0} = x_{12}^{n^\star}, \tag{187}$$

where $n^\star = n$ (Jacobi) or $n^\star = n + 1$ (Gauss–Seidel). The operator Λ is chosen among the following family $\Lambda_{\alpha,\beta}$ where $\alpha > 0$ and $\beta \geqslant 0$ (Λ is a bijection from $H^{1/2}(0, L)$ onto $L^2(0, L)$ as soon as $\beta > 0$):

$$\Lambda = \Lambda_{\alpha,\beta} = \left(\alpha Id - \frac{\beta}{\omega^2}\partial_{yy}^2 \right)^{1/4}. \tag{188}$$

For a given relaxation parameter r, the x_{12}^n and x_{21}^n are updated by the formula:

$$x_{12}^{n+1} = (1 - r)x_{12}^n + r\left(-(\Lambda^*)^{-1}\partial_x \tilde{u}_1^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{u}_1^{n+1}{}_{/x=0} \right),$$

$$x_{21}^{n+1} = (1 - r)x_{21}^n + r\left((\Lambda^*)^{-1}\partial_x \tilde{u}_2^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{u}_2^{n+1}{}_{/x=0} \right). \tag{189}$$

From the general theory, we already know that the errors $\tilde{e}_k^n = \tilde{u}_k^n - u_k$, $k = 1, 2$ converge to zero. The aim of what follows is to determine the rate of this convergence. At first, the errors satisfy both the Helmholtz
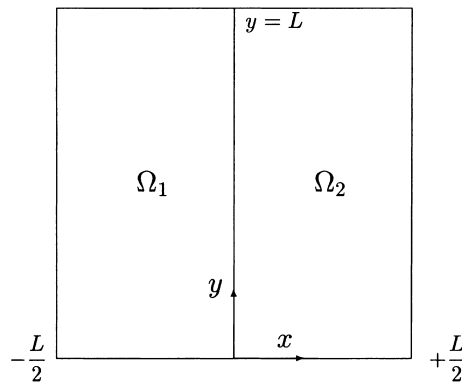

Fig. 2. Geometry of the simple model.

equation with zero right hand side and the absorbing boundary condition at $x = +(L/2)(k = 2)$ and $x = -(L/2)(k = 1)$. Using Theorem 3, we deduce the estimates:

$$
\begin{aligned}
\left\| \tilde{e}_1^{n+1} \right\|_{H^1(\Omega_1)}^2 &\leqslant C_0 \left\| (\Lambda^*)^{-1} \partial_x \tilde{e}_1^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{e}_1^{n+1}{}_{/x=0} \right\|_{L^2(0,L)}^2 \leqslant C_0 \left\| \varepsilon_{21}^n \right\|_{L^2(0,L)}^2, \\
\left\| \tilde{e}_2^{n+1} \right\|_{H^1(\Omega_2)}^2 &\leqslant C_0 \left\| \left(-(\Lambda^*)^{-1} \partial_x \tilde{e}_2^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{e}_2^{n+1} \right)_{/x=0} \right\|_{L^2(0,L)}^2 \leqslant C_0 \left\| \varepsilon_{12}^{n\star} \right\|_{L^2(0,L)}^2.
\end{aligned}
\tag{190}
$$

where ε_{12}^n and ε_{21}^n are the errors at interfaces. They satisfy:

$$
\begin{aligned}
\varepsilon_{12}^{n+1} &= (1-r)\varepsilon_{12}^n + r\left(-(\Lambda^*)^{-1} \partial_x \tilde{e}_1^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{e}_1^{n+1}{}_{/x=0} \right), \\
\varepsilon_{12}^n &= (1-r)\varepsilon_{12}^n - r\left((\Lambda^*)^{-1} \partial_x \tilde{e}_2^{n+1}{}_{/x=0} + i\omega \Lambda \tilde{e}_2^{n+1}{}_{/x=0} \right).
\end{aligned}
\tag{191}
$$

In Section 5.2, we derive explicit expressions for the errors ε_{12}^n and ε_{21}^n which lead us to sharp error estimates.

5.2. A general error estimate

The solution of this sequence of problems can be explicitly derived with the help of a technique of separation of variables. We introduce,

$$
\begin{aligned}
\tilde{e}_k^n(x,y) &= \sum_{p=0}^{\infty} \hat{e}_{k,p}^n(x) w_p(y), \quad k = 1, 2, \\
\varepsilon_{kj}^n &= \sum_{p=o}^{\infty} \hat{\varepsilon}_{kj,p}^n w_p(y), \quad kj = 12 \text{ or } 21,
\end{aligned}
\tag{192}
$$

where functions $(w_p(y)$ are the eigenvectors of ∂_{yy} with Neumann boundary conditions and are also eigenfunctions of $\Lambda_{\alpha,\beta}$:

$$
\begin{aligned}
w_p(y) &= \sqrt{\tfrac{2}{L}} \cos\left(\tfrac{\pi}{L}\left(y - \tfrac{L}{2} \right) \right), \\
\Lambda_{\alpha,\beta} w_p &= \sqrt{t_p^{\alpha,\beta}} w_p, \quad t_p^{\alpha,\beta} = \left(\alpha + \tfrac{\beta\pi^2}{\omega^2 L^2} p^2 \right)^{1/2}, \\
(\omega^2 + \Delta) w_p &= k_p^2 w_p + \partial_{xx}^2 w_p, \quad k_p = \sqrt{\omega^2 - \tfrac{\pi^2 p^2}{L^2}},
\end{aligned}
\tag{193}
$$

(the choice of the sign for the complex square root defining k_p does not really matter).

The quantities $\hat{e}_{k,p}^n(x)$ are the Fourier coefficients of the error. We shall say that $\hat{e}_{k,p}^n(x)$ is the mode of order p of the error term $\tilde{e}_k^n$. Each of them satisfies a reduced homogeneous Helmholtz equation

$$
k_p^2 \hat{e}_{k,p}^{n+1} + \partial_{xx}^2 \hat{e}_{k,p}^{n+1} = 0,
\tag{194}
$$

in their respective domains. These equations are solved according to

$$
\begin{aligned}
\hat{e}_{1,p}^{n+1} &= a_{1,p}^{n+1} \exp(ik_p x) + b_{1,p}^{n+1} \exp(-ik_p x), \\
\hat{e}_{2,p}^{n+1} &= a_{2,p}^{n+1} \exp(-ik_p x) + b_{2,p}^{n+1} \exp(ik_p x).
\end{aligned}
\tag{195}
$$

Boundary conditions at $x = 0$ and $x = \pm(L/2)$ allows us to determine $a_{k,p}^{n+1}$ and $b_{k,p}^{n+1}$ as a function of the $\hat{\varepsilon}_{12,p}^n$ and $\hat{\varepsilon}_{21,p}^n$.

$$
A_p \begin{bmatrix} a_{1,p}^{n+1} \\ b_{1,p}^{n+1} \end{bmatrix} = \hat{\varepsilon}_{21,p}^n \begin{bmatrix} \sqrt{t_p^{\alpha,\beta}} \\ 0 \end{bmatrix}, \quad A_p \begin{bmatrix} a_{2,p}^{n+1} \\ b_{2,p}^{n+1} \end{bmatrix} = \hat{\varepsilon}_{12,p}^{n\star} \begin{bmatrix} \sqrt{t_p^{\alpha,\beta}} \\ 0 \end{bmatrix},
\tag{196}
$$

with

$$
A_p = \begin{bmatrix} ik_p + i\omega t_p^{\alpha,\beta} & -ik_p + i\omega t_p^{\alpha,\beta} \\ (-ik_p + i\omega)e^{-ik_p L/2} & (ik_p + i\omega)e^{+ik_p L/2} \end{bmatrix},
\tag{197}
$$

Our goal is to obtain a recursive relation for $(\hat{\varepsilon}_{12,p}^n, \hat{\varepsilon}_{21,p}^n)$. Eq. (191) implies the following equality for the Fourier components of the error at interface:

$$
\hat{\varepsilon}_{12,p}^{n+1} = (1-r)\hat{\varepsilon}_{12,p}^n + r\left(- \left(t_p^{\alpha,\beta}\right)^{-(1/2)} \partial_x \hat{e}_{1,p}^{n+1}{}_{/x=0} + \mathrm{i}\omega\left(t_p^{\alpha,\beta}\right)^{1/2} \hat{e}_{1,p}^{n+1}{}_{/x=0} \right),
$$
$$
\hat{\varepsilon}_{21,p}^{n+1} = (1-r)\hat{\varepsilon}_{21,p}^n + r\left(\left(t_p^{\alpha,\beta}\right)^{-(1/2)} \partial_x \hat{e}_{2,p}^{n+1}{}_{/x=0} + \mathrm{i}\omega\left(t_p^{\alpha,\beta}\right)^{1/2} \hat{e}_{2,p}^{n+1}{}_{/x=0} \right).
$$

(198)

or, using (195):

$$
\hat{\varepsilon}_{12,p}^{n+1} = (1-r)\hat{\varepsilon}_{12,p}^n + \tfrac{r}{\sqrt{t^{\alpha,\beta}}}\left(\left(\mathrm{i}k_p + \mathrm{i}\omega t_p^{\alpha,\beta}\right)a_{1,p}^{n+1} + \left(- \mathrm{i}k_p + \mathrm{i}\omega t_p^{\alpha,\beta}\right)b_{1,p}^{n+1} \right),
$$
$$
\hat{\varepsilon}_{21,p}^{n+1} = (1-r)\hat{\varepsilon}_{21,p}^n + \tfrac{r}{\sqrt{t^{\alpha,\beta}}}\left(\left(k_p + \omega t_p^{\alpha,\beta}\right)a_{2,p}^{n+1} + \left(- k_p + \omega t_p^{\alpha,\beta}\right)b_{2,p}^{n+1} \right).
$$

(199)

Via (196), we can eliminate the $a_{k,p}^{n+1}$ and $b_{k,p}^{n+1}$ and obtain a relationship between two successive iterates. We obtain:

$$
\hat{\varepsilon}_{12,p}^{n+1} = (1-r)\hat{\varepsilon}_{12,p}^n + r\lambda_p \hat{\varepsilon}_{21,p}^n,
$$
$$
\hat{\varepsilon}_{21,p}^{n+1} = (1-r)\hat{\varepsilon}_{21,p}^n + r\lambda_p \hat{\varepsilon}_{12,p}^{n\star},
$$

(200)

with

$$
\lambda_p = \frac{(-k_p + \omega t_p^{\alpha,\beta})(k_p + \omega)\mathrm{e}^{\mathrm{i}(k_p L/2)} - (k_p + \omega t_p^{\alpha,\beta})(-k_p + \omega)\mathrm{e}^{-\mathrm{i}(k_p L/2)}}{(k_p + \omega t_p^{\alpha,\beta})(k_p + \omega)\mathrm{e}^{\mathrm{i}(k_p L/2)} - \left(- k_p + \omega t_p^{\alpha,\beta}\right)(-k_p + \omega)\mathrm{e}^{-\mathrm{i}(k_p L/2)}}.
$$

(201)

Eq. (200) can be written in the matrix form

$$
\begin{bmatrix} \hat{\varepsilon}_{12,p}^{n+1} \\ \hat{\varepsilon}_{21,p}^{n+1} \end{bmatrix} = M_p(r)\begin{bmatrix} \hat{\varepsilon}_{12,p}^n \\ \hat{\varepsilon}_{21,p}^n \end{bmatrix},
$$

(202)

with

Jacobi: $M_p(r) = \begin{bmatrix} 1 - r & r\lambda_p \\ r\lambda_p & (1-r) \end{bmatrix},$

(203)

Gauss–Seidel: $M_p(r) = \begin{bmatrix} 1 - r & r\lambda_p \\ r(1-r)\lambda_p & (1-r) + r^2\lambda_p^2 \end{bmatrix}.$

(204)

From (202), we deduce the sharp estimate (with Jacobi at least, the equality may occur)

$$
\left|\hat{\varepsilon}_{12,p}^{n+1}\right|^2 + \left|\hat{\varepsilon}_{21,p}^{n+1}\right|^2 \leqslant \|M_p(r)\|_2^{2n}\left(\left|\hat{\varepsilon}_{12,p}^0\right|^2 + \left|\hat{\varepsilon}_{21,p}^0\right|^2 \right),
$$

(205)

where $\|M\|_2$ denotes the maximum of the square root of the eigenvalues of the matrix M^*M. Of course, the general theory of convergence indicates us that:

$$
\forall p, \quad \|M_p(r)\|_2 < 1.
$$

(206)

This can also be checked by a direct computation for the Jacobi's algorithm using $|\lambda_p| < 1$ (see Section 5.3) but is less obvious for the Gauss–Seidel algorithm. Now, we define

$$
\tau_r = \sup_p \|M_p(r)\|_2,
$$

(207)

and using Plancherel–Parceval's theorem

$$\left\|\varepsilon_{jk}^n\right\|_{L^2(0,L)}^2 = \sum_p \left|\hat{\varepsilon}_{jk,p}^n\right|^2, \tag{208}$$

inequality (205) leads to the estimate

$$\left\|\tilde{e}_1^{n+1}\right\|_{H^1(\Omega_1)} \leqslant C\tau_r^n, \quad \left\|\tilde{e}_2^{n+1}\right\|_{H^1(\Omega_1)} \leqslant C\tau_r^{n\star}, \tag{209}$$

with

$$C^2 = C_0^2\left\|\varepsilon_{12}^0\right\|_{L^2(0,L)}^2 + C_0^2\left\|\varepsilon_{21}^0\right\|_{L^2(0,L)}^2. \tag{210}$$

As a consequence, a geometrical convergence occurs when τ_r is strictly less than 1. Moreover, τ_r governs the rate of convergence: it is thus the quantity to analyze.

5.3. Analysis of the non zero eigenvalues of A

It is not difficult to realize that the λ_p's are nothing else than the non zero eigenvalues of operator A ($Ax = 0$, if x is localized on the boundaries Σ_{kk}). For A is a contraction, all the $|\lambda_p|$ must be less than 1. We distinguish different cases according to k_p is real, purely imaginary or zero

• If $(p\pi/\omega L) < 1$: The mode is propagative, k_p is real, equals $\omega\sqrt{1 - (p^2\pi^2/\omega^2 L^2)}$ and

$$\lambda_p = \frac{\cos(k_p L/2)k_p\omega(t_p^{\alpha,\beta} - 1) + \mathrm{i}\ \sin(k_p L/2)\left(\omega^2 t_p^{\alpha,\beta} - k_p^2\right)}{\cos(k_p L/2)k_p\omega(t_p^{\alpha,\beta} + 1) + \mathrm{i}\ \sin(k_p L/2)\left(\omega^2 t_p^{\alpha,\beta} + k_p^2\right)}. \tag{211}$$

• If $(p\pi/\omega L) > 1$: The mode is evanescent, k_p is imaginary, equals $\mathrm{i}\xi_p = \mathrm{i}\sqrt{(p^2\pi^2/\omega^2 L^2) - 1}$ and

$$\lambda_p = \frac{\mathrm{i}\ \cosh(\xi_p L/2)\xi_p\omega(t_p^{\alpha,\beta} - 1) - \sinh(\xi_p L/2)\left(\omega^2 t_p^{\alpha,\beta} + \xi_p^2\right)}{\mathrm{i}\ \cosh(\xi_p L/2)\xi_p\omega(t_p^{\alpha,\beta} + 1) - \sinh(\xi_p L/2)\left(\omega^2 t_p^{\alpha,\beta} - \xi_p^2\right)}. \tag{212}$$

• If $(p\pi/\omega L) = 1$: $k_p = 0$ and

$$\lambda_p = \lambda^* \equiv \frac{\sqrt{\alpha + \beta} - 1 + \mathrm{i}(\omega L/2)\sqrt{\alpha + \beta}}{\sqrt{\alpha + \beta} + 1 + \mathrm{i}(\omega L/2)\sqrt{\alpha + \beta}}. \tag{213}$$

In each case, it is not difficult to check that the modulus of λ_p is always strictly less than 1 as we could expect. Analyzing what happens respectively for small and large values of p permits to distinguish the role of the parameters α and β in the convergence process. Indeed:

• When p is 0, we get

$$\lambda_0 = \frac{\sqrt{\alpha} - 1}{\sqrt{\alpha} + 1}, \tag{214}$$

which shows that α will govern essentially the convergence of the low frequencies.

• When p grows to infinity, we get

$$\lim_{p\to\infty} \lambda_p \equiv \lambda_\infty = \frac{\mathrm{i}\sqrt{\beta} - 1}{\mathrm{i}\sqrt{\beta} + 1}, \tag{215}$$

which shows that β will govern essentially the convergence of the high frequencies.

In particular (215) shows that

$$\lim_{p\to\infty} |\lambda_p| = 1, \tag{216}$$

what implies that $\|A\|$ equals to 1: this demonstrates, in this particular case, that the formal series $\sum A^n$ does not converge absolutely.

5.4. The convergence rate for the relaxed Jacobi's algorithm

Let us first check that for the Jacobi's algorithm, one gets exponential convergence as soon as $\beta \neq 0$ and $r \in\ ]0, 1[$. One computes easily that

$$\|M_p(r)\|_2 = \max_{\pm} |1 - r \pm r\lambda_p| < 1. \tag{217}$$

The only delicate point is to ensure that the λ_p do not accumulate near ± 1 in which case $\|M_p(r)\|_2$ would tend to 1 for all values of r and τ_r would be 1. According to (215), we have

$$\tau_r^\infty = \lim_{p \to +\infty} \|M_p(r)\|_2 = \max_{\pm} |1 - r \pm r\lambda_\infty|, \tag{218}$$

which is strictly smaller than 1 for $\beta \neq 0$ and $r \in\ ]0, 1[$ (see Fig. 3).

Remark 11. *Here is the point where we can recover the precise role played by operator $T = \Lambda^*\Lambda$. The fact that T is a pseudo-differential operator of first order reads in the behavior of its symbol $t_p^{\alpha,\beta}$ when p becomes large: $t_p^{\alpha,\beta} \sim C_p$. Note that this choice is the only one to get the limit in (215) different from ± 1.*

Remark 12. *When $\beta = 0$ (local operator), it is easy to see that $\tau_r^\infty = 1$ which implies $\tau_r = 1$. In this case, the convergence is not exponential and the estimate (210) is not sufficient to prove the convergence. One then has to take into account how the quantity $(|\hat{\varepsilon}_{12,p}^0|^2 + |\hat{\varepsilon}_{21,p}^0|^2)$ occurring in (205) goes to zero when p goes to infinity, which involves the regularity on the interface of the exact solution. In [15], it is shown that, if the solution is smooth one gets, for any $\varepsilon > 0$:*

$$\|\tilde{e}_1^n\|_{H^1(\Omega_1)} + \|\tilde{e}_2^n\|_{H^1(\Omega_2)} \leqslant \frac{C_\varepsilon}{n^{1-\varepsilon}}, \qquad \text{for } r \neq 1/2 \text{ and}$$

$$\|\tilde{e}_1^n\|_{H^1(\Omega_1)} + \|\tilde{e}_2^n\|_{H^1(\Omega_2)} \leqslant \frac{C_\varepsilon}{n^{(3/2)-\varepsilon}}, \qquad \text{for } r = 1/2.$$

Of course the next step of the analysis consists in answering the question: what is the choice for the parameters, r, α and β which will minimize $|\tau_r|$? As we are going to see, a first naive analysis based on the examination of low and high frequencies would lead to choose $\alpha = 1$, $\beta = 1$ and $r = 1/2$. However, if one looks at a typical curve $p \to \|M_p(r)\|_2$ one sees that one has to distinguish three regions for p.
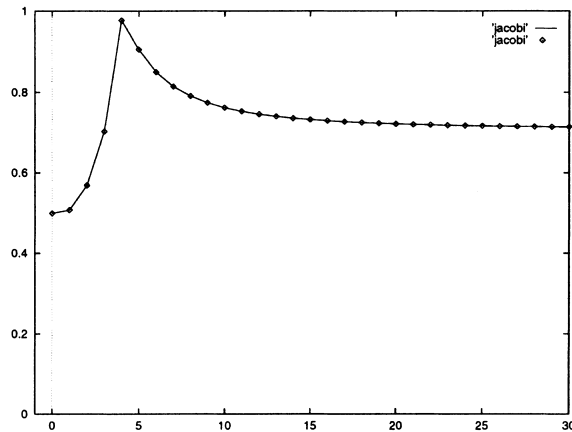

Fig. 3. Domain of two wavelengths: τ_p in function of p for $r = 1/2$, $\alpha = \beta = 1$.

- High frequencies: $(p\pi/\omega L) \gg 1$

  For these values, the convergence is essentially governed by:

  $$\tau_r^\infty = \max_{\pm} |1 - r \pm r\lambda_\infty|.$$

  It is then straightforward to realize that the best way to damp these evanescent modes is to pick $\beta = 1$ and $r = 1/2$: the λ_p's accumulates near i and $\|M_p(r = 1/2)\|_2$ tends to $1/\sqrt{2}$.

- Low frequencies: $(p\pi/\omega L) \ll 1$

  For these values, the convergence is essentially governed by:

  $$\tau_r^0 = \max_{\pm} |1 - r \pm r\lambda_0|.$$

  Using (214), one sees that the best value for α is 1.

Note that if one stops the discussion here, the size L of the sub-domains has no influence on our choice.

- Resonant frequencies: $(p\pi/\omega L) \simeq 1$

  These values of p correspond to a local maximum of the curve $p \to \|M_p(r)\|_2$, which is also the global maximum in a lot of cases. The convergence of the related modes is now governed by

  $$\tau_r^\star = \max_{\pm} |1 - r \pm r\lambda^*|,$$

  where λ^* is given by (213), which shows that trying to minimize τ_r^* involves the value of ωL. In particular, λ^* tends to 1 when ωL tends to $+\infty$, whatever the value of (α, β) or r is.

It appears then that a good strategy for choosing α, β and r consists of minimizing, ωL being given, the quantity:
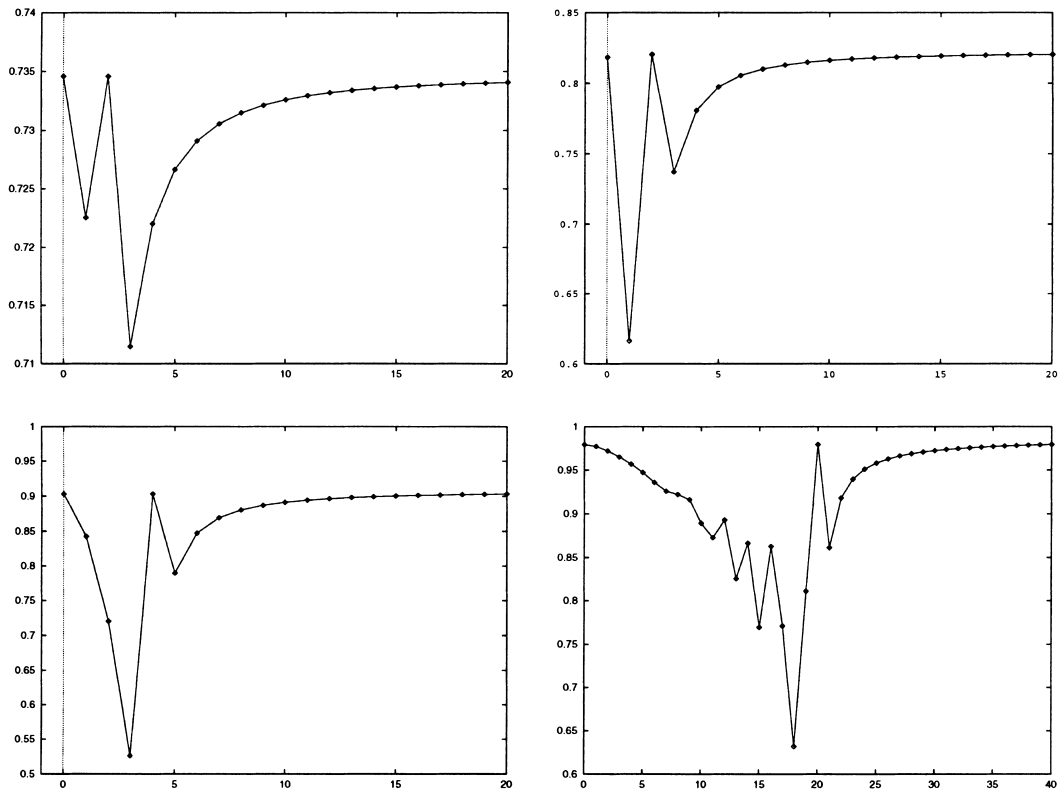

Fig. 4. Optimal curves $p \to \|M_p(r)\|^2$ for the Jacobi's algorithm ($\omega L = \pi$: top left, $\omega L = 2\pi$: top right, $\omega L = 4\pi$: bottom left, $\omega L = 10\pi$: bottom right).

$$\max\{\tau_r^0, \tau_r^*, \tau_r^\infty\}, \tag{219}$$

which can easily be done, at least numerically. We present the results of this optimization procedure in Fig. 4 where, for different values of the frequency (from $\omega L = \pi$ which corresponds to $L = 0.5$ wavelengths

Table 1
Optimal parameters for four different lengths of domain (Jacobi)

	$\omega L = \pi$	$\omega L = 2\pi$	$\omega L = 4\pi$	$\omega L = 10\pi$
α	0.11982	0.0269	0.007435	0.000384
β	0.85500	0.55	0.23076	0.0032
r	0.5160	0.645	0.5160	0.54126
τ_r	0.734588	0.821	0.903	0.9792

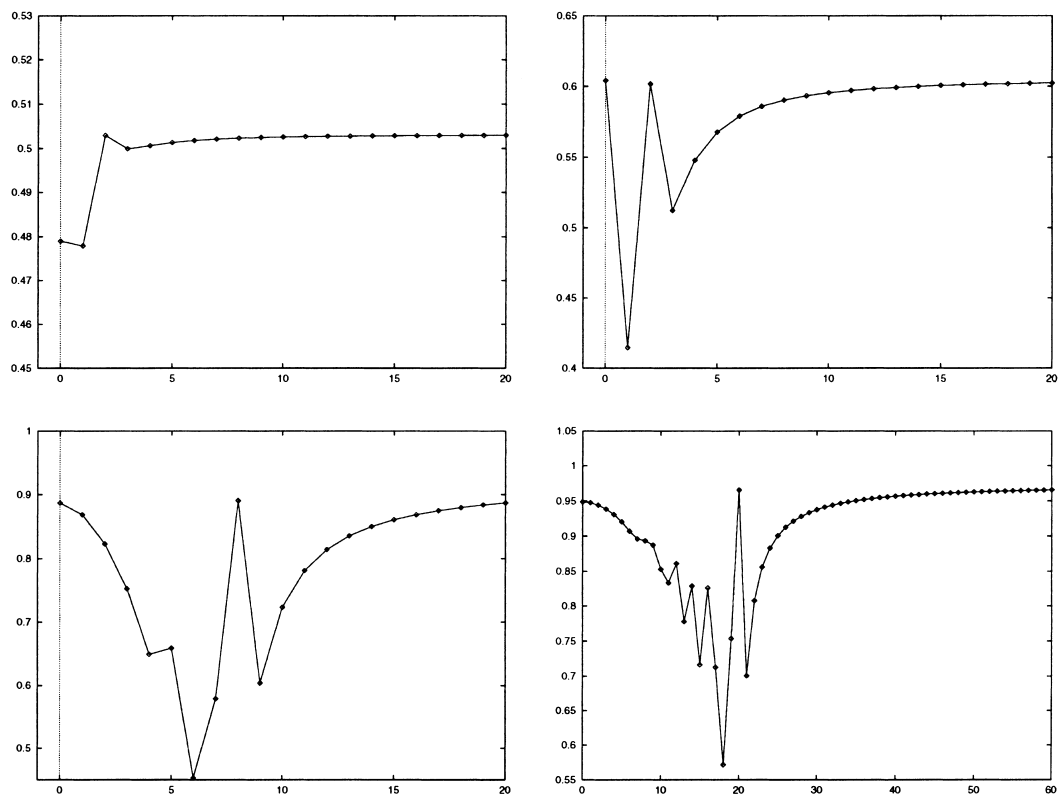
Fig. 5. Optimal curves $p \to \|M_p(r)\|^2$ for the Gauss–Seidel's algorithm ($\omega L = \pi$: top left, $\omega L = 2\pi$: top right, $\omega L = 4\pi$: bottom left, $\omega L = 10\pi$: bottom right).

Table 2
Optimal parameters for four different lengths of domain (Gauss–Seidel)

	$\omega L = \pi$	$\omega L = 2\pi$	$\omega L = 4\pi$	$\omega L = 10\pi$
α	0.1538	0.0512	0.00641	0.001948
β	0.87179	0.43589	0.0871	0.0317
r	0.5	0.55128	0.4151	0.3235
τ_r	0.5029	0.604	0.8921	0.9654

to $\omega L = 10\pi$ which corresponds to $L = 5$ wavelengths), we plot the curves $p \rightarrow \|M_p(r)\|_2$ for the optimal values of (α, β, r) which are given in Table 1, as well as the optimal value of the convergence rate. We observe experimentally that the optimal value of (α, β, r) is the one which ensures the equality

$$\tau_r^0 = \tau_r^* = \tau_r^\infty. \tag{220}$$

Also, we notice that the optimal rate decreases when the frequency increases which is not surprising and at least coherent with the fact τ^* tends to 1 when ωL tends to $+\infty$. Note also that for large frequencies α_{opt} and β_{opt} are really different from 1 while r_{opt} remains closed to $1/2$.

5.5. The convergence rate for the Gauss–Seidel's algorithm

We do not reproduce the discussion which is essentially the same than for the Jacobi's algorithm except the fact that the computations are heavier. We only present in Fig. 5 and in Table 2 the results of the research optimal parameters. Let us emphasize the fact that, for a given value of ωL, we get a better convergence with the Gauss–Seidel algorithm than with the Jacobi's algorithm.

6. Other applications

Thanks to our general framework for domain decomposition methods is, the presentation of their extension to other physical models is simplified. Moreover, the abstract convergence proof remains true once one has verified that the operators E, S and A have the good properties. In Section 6.2, we shall consider the two important examples of the elastodynamics equations and of the 3D Maxwell's system.

Another advantage of our abstract framework is to include the discrete problems obtained after approximation in space, provided a good choice of the approximation process. We are going to show this in Section 6.1.

6.1. Space discretization and domain decomposition

In this section, we go back to our model problem (P_Ω^f) in dimension $N = 2$ or 3, and consider the case where the operator Λ_{jk} is defined by (cf. Section 4):

$$\Lambda_{jk} = \left(\alpha I - \frac{\beta}{\omega^2} \Delta_{jk} \right)^{1/4},$$

and assume for simplicity that the Ω_k's are polyhedral domains. The importance of this section lies in the fact the convergence of the algorithm for the continuous problem does not imply the convergence for any discrete approximation. Initial numerical experiments performed by B. Despres (personal communication) show that a naïve finite-difference approximation of the interface conditions might lead to unstable iterative algorithms.

In order to avoid this kind of accident, the first major idea is that the discretization should have the same variational structure as the continuous one. This leads naturally to consider finite element methods. In this class of methods, the choice of mixed-hybrid finite element methods was guided by the fact that the quantities transmitted at interfaces, namely the trace of the solution and of its normal derivative, were easily and consistently computable from the degrees of freedom, which is not the case for instance with Lagrange elements. Finally, we realized that this choice permitted to set the resulting discrete problem within our general framework of Sections 2 and 3, ensuring the convergence of the iterative method. In the rest of this section, we assume for simplicity that the Ω_k's are polyhedral domains. We restrict ourselves to the lower order element for simplicity.

The meshes. For each sub-domain Ω_k, we introduce a finite element triangulation $\mathcal{T}_h(\Omega_k)$ (of triangles in 2D, tetrahedra in 3D) of step-size $h > 0$. We shall denote by K the current element of $\mathcal{T}_h(\Omega_k)$. Moreover, for each interface Σ_{jk}, we shall assume that the traces of the two triangulations $\mathcal{T}_h(\Omega_k)$ and $\mathcal{T}_h(\Omega_j)$ coincide with one single surface triangulation of Σ_{jk}, denoted $\mathcal{T}_h(\Sigma_{jk})$: in other words, we work with matching grids.

We shall denote by T the current element of such a discretization. We also denote by $\mathscr{T}_h(\Sigma_{kk})$ the trace of $\mathscr{T}_h(\Omega_k)$ on Σ_{kk}.

The space V^h and the unknown x^h. Here, we identify the space V^h which will play the role of the space V of Section 2. We have

$$V^h = \prod_{j,k=1}^{K} V_{kj}^h \subset V, \tag{221}$$

where

$$V_{kj}^h = \{x_{kj}^h \in L^2(\Sigma_{jk})/\forall T \in \mathscr{T}_h(\Sigma_{jk}), x_{kj/T}^h \in P_0(T)\}. \tag{222}$$

Of course we shall look for $x^h \in V^h$ as an approximation of $x \in V$. Let us notice that:

$$\Pi V^h \subset V^h, \tag{223}$$

where Π is defined in (26).

The operators E^h, S^h and A^h. Let us admit for a while the existence of selfadjoint operators

$$\Lambda_{kj}^h \in \mathscr{L}(V_{kj}^h), \tag{224}$$

which are supposed to be approximations of Λ_{kj}. In order to define E^h, we first introduce a space M^h for the approximation of $\prod_{k=1}^{K} L^2(\Omega_k)$:

$$M^h = \prod_{k=1}^{K} M_k^h, \tag{225}$$

where

$$M_k^h = \{u_k^h \in L^2(\Omega_k)/\forall K \in \mathscr{T}_h(\Omega_k), u_{k/K}^h \in P_0(K)\}, \tag{226}$$

and a space X^h for the approximation of $\prod_{k=1}^{K} H(\mathrm{div}; \Omega_K)$

$$X^h = \prod_{k=1}^{K} X_k^h, \tag{227}$$

where

$$X_k^h = \{p_k^h \in H(\mathrm{div}; \Omega_k)/\forall K \in \mathscr{T}_h(\Omega_k), p_{k/K}^h \in RT_0(K)\}. \tag{228}$$

Here, $RT_0(K)$ denotes the usual Raviart–Thomas–Nédelec space of lowest order [30]. Then, we can define:

$$E^h \in \mathscr{L}(V^h, M^h), \tag{229}$$

such that:

$$e^h = (e_k^h) = E^h(y^h), \tag{230}$$

if and only if $(e^h, p^h, t^h) \in M^h \times X^h \times V^h$ is the solution of the following collection of local variational problems

$$\int_{\Omega_k} p_k^h \cdot q_k^h \, \mathrm{d}x + \int_{\Omega_k} e_k^h \, \mathrm{div} \, q_k^h \, \mathrm{d}x = \sum_j \int_{\Sigma_{kj}} t_{kj}^h q_k^h \cdot v_k \, \mathrm{d}\sigma, \quad \forall q_k^h \in X_k^h,$$

$$-\int_{\Omega_k} \mathrm{div} \, p_k^h \cdot v_k^h \, \mathrm{d}x - \omega^2 \int_{\Omega_k} e_k^h v_k^h \, \mathrm{d}x = \int_{\Omega_k} f v_k^h \, \mathrm{d}x, \quad \forall v_k^h \in M_k^h, \tag{231}$$

$$\left(\Lambda_{kj}^h\right)^{-1} \left(p_k^h \cdot v_{k/\Sigma_{kj}}\right) + \mathrm{i}\omega \Lambda_{kj}^h(t_k^h) = y_{kj}^h, \quad \text{in } V_{kj}^h,$$

which is nothing but a standard mixed-hybrid approximation of the local problem defining e_k in (31). We then define:

$$S^h \in \mathcal{L}(V^h), \tag{232}$$

by expressing that:

$$x^h = S^h y^h, \tag{233}$$

if and only if:

$$\forall (k,j), \qquad x_{kj}^h = -\left(\Lambda_{kj}^h\right)^{-1}\left(p_k^h \cdot v_{k/\Sigma_{kj}}\right) + \mathrm{i}\omega\, \Lambda_{kj}^h e_{k/\Sigma_{kj}}^h, \quad \text{with } (e_k^h, p_k^h, t_k^h) \text{ solution of } (231). \tag{234}$$

The reader will easily check that S^h is an isometry, that A^h is a contraction and that $I - A^h$ is invertible. For completeness we need to explain how one can construct Λ_{jk}^h. The idea is to define

$$\Lambda_{jk}^h = \left(\alpha I - \frac{\beta}{\omega^2}\Delta_{jk}^h\right)^{1/4}, \tag{235}$$

where $\Delta_{jk}^h \in \mathcal{L}(V_{jk}^h)$ is an approximation of the Laplace–Beltrami operator Δ_{jk}, that one can construct using mixed finite elements as follows. We first define the space

$$\Phi_{jk}^h = \{\varphi_{jk}^h \in H(\mathrm{div}_\Gamma; \Sigma_{jk})/\forall T \in \mathcal{T}^h(\Sigma_{jk}), \varphi_{jk/T}^h \in RT_0(T)\}, \tag{236}$$

and we define:

$$y_{jk}^h = -\Delta_{jk}^h x_{jk}^h, \tag{237}$$

if and only if:

$$\begin{aligned}
\int_{\Sigma_{jk}} y_{jk}^h z_{jk}^h \,\mathrm{d}\sigma &= \int_{\Sigma_{jk}} \mathrm{div}\,\varphi_{jk}^h z_{jk}^h \,\mathrm{d}\sigma, \quad \forall z_{jk}^h \in V_{jk}^h, \\
\int_{\Sigma_{jk}} \varphi_{jk}^h \cdot \psi_{jk}^h \,\mathrm{d}\sigma &= \int_{\Sigma_{jk}} x_{jk}^h \mathrm{div}\,\psi_{jk}^h \,\mathrm{d}\sigma, \quad \forall \psi_{jk}^h \in \Phi_{jk}^h.
\end{aligned} \tag{238}$$

Note that in practice, one has to construct the matrix representing $-\Delta_{jk}^h$ and to diagonalize this matrix in order to compute Λ_{jk}^h.

6.2. Other physical models

We present here how to adapt our domain decomposition method to problems whose unknown are vector fields. We try essentially to keep the notation of the previous section and only mention the new ones when we feel this is useful. For simplicity, we only present the case where the operators Λ_{jk} are identity operators. For brevity, we simply make precise what is the interface unknown x, the space V and the operators E, S and A which permits us to apply our abstract iterative methods. We let the reader check that the required properties are satisfied. This is straightforward in the case of elastodynamics equations, but less obvious for Maxwell's equations (we give some indications in this case).

We do not speak at all here of the discretization of these problems. The extension of the content of Section 6.1 is easy provided that one uses appropriate mixed elements for elasticity [4,30]. The case of Maxwell's equations is less standard and requires the hybrid version of the mixed elements of Nédélec [29]. We refer the reader to [18] or [9].

6.2.1. Elastodynamics equations
Notation. In what follows:
- $\Omega \subset R^N$ is the (bounded) propagation medium,
- ρ is the density of this medium,
- $u = (u_i)$ denotes the displacement field,
- $\varepsilon(u) = ((\varepsilon_{ij}(u)))$ is the strain tensor,
- $\sigma(u) = ((\sigma_{ij}(u)))$ is the stress tensor.

Let us recall that, for small deformations:

$$\varepsilon_{ij}(u) = \frac{1}{2}\left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i}\right), \tag{239}$$

and that in a linear medium, $\sigma(u)$ and $\varepsilon(u)$ are related by the Hooke's law

$$\sigma(u) = C \cdot \varepsilon(u), \tag{240}$$

where C is a fourth-order tensor characteristic of the medium.

The model problem. For a given source term $f \in L^2(\Omega)^N$, we consider the problem:

Find $u \in H^1(\Omega)^N$ such that

$$-\mathrm{div}\,\sigma(u) - \rho\omega^2 u = f \quad \text{in } \Omega, \tag{241}$$

$$\sigma(u)v + \mathrm{i}\omega\rho u = 0 \quad \text{on } \partial\Omega.$$

The space V and the unknown x. In this case we have

$$V = \prod_{k=1}^{K} V_k, \quad V_k = L^2(\partial\Omega_k)^N,$$
$$x = (x_k), \quad x_k = -\sigma(u_k)v_k + \mathrm{i}\omega\rho u_k. \tag{242}$$

The operators E, S and A. We define

$$E \in \mathcal{L}\left(V, \prod_{k=1}^{K} H^1(\partial\Omega_k)^N\right), \tag{243}$$

such that

$$e = (e_k) = E(y), \tag{244}$$

if and only if e_k is the solution of

$$-\mathrm{div}\,\sigma(e_k) - \rho\omega^2 e_k = 0, \quad \text{in } \Omega_k,$$
$$\sigma(e_k)v + \mathrm{i}\omega\rho e_k = y_k \quad \text{on } \partial\Omega_k, \tag{245}$$

then

$$S \in \mathcal{L}(V), \tag{246}$$

by

$$x = Sy, \tag{247}$$

if and only if

$$\forall k \leqslant K, \qquad x_k = -\sigma(e_k)v + \mathrm{i}\omega\rho e_k, \quad (e_k) \text{ solution of (245)}, \tag{248}$$

and finally

$$A = S\Pi \in \mathcal{L}(V), \tag{249}$$

where Π is defined as in Section 2, except the fact it is now applied to vector fields.

6.2.2. 3D Maxwell's system

Notation. In what follows:
- $\Omega \subset R^3$ is the (bounded) propagation medium,
- ε is the electric permittivity of this medium,
- μ is its magnetic permeability,
- $\mathbf{E} = (\mathbf{E}_i)$ denotes the electric field (the unknown).

The model problem. For a given current density $j \in L^2(\Omega)^3$, we consider the problem:

Find $\mathbf{E} \in H(\text{curl}; \Omega)$ such that

$$\text{curl}\,(\mu^{-1}\text{curl}\,\mathbf{E}) - \varepsilon\omega^2\mathbf{E} = i\omega\mathbf{j} \quad \text{in } \Omega, \tag{250}$$

$$\text{curl}\,\mathbf{E} \times v + i\omega v \times (\mathbf{E} \times v) = 0 \quad \text{on } \partial\Omega.$$

Remark 13. *The variational problem equivalent to* (250) *is not so standard. It must be set in the space.*

$$\mathscr{V}(\Omega) = \{\mathbf{E} \in H(\text{curl}; \Omega)/v \times (\mathbf{E} \times v) \in L^2_t(\partial\Omega)\},$$

where $L^2_t(\partial\Omega)$ is the space of tangential square integrable vector fields on $\partial\Omega$

$$L^2_t(\partial\Omega) = \{\varphi \in L^2(\partial\Omega)^3/\varphi \cdot v = 0\}.$$

As a consequence, the solution $\mathbf{E}$ of (250) *satisfies*:

$$(\text{curl}\,\mathbf{E} \times v, v \times (\mathbf{E} \times v)) \in L^2_t(\partial\Omega)^2.$$

The space V and the unknown x. In this case, we have

$$V = \prod_{k=1}^{K} V_k, \quad V_k = L^2_t(\partial\Omega_k), \tag{251}$$

$$x = (x_k), \quad x_k = -\text{curl}\,\mathbf{E}_k \times v + i\omega v \times (\mathbf{E}_k \times v).$$

The operators E, S and A. We define

$$E \in \mathscr{L}\left(V, \prod_{k=1}^{K} H(\text{curl}; \Omega_k)\right), \tag{252}$$

such that

$$\mathbf{F} = (\mathbf{F}_k) = E(y), \tag{253}$$

if and only if $\mathbf{F}_k$ is the solution of:

$$\text{curl}\,(\mu^{-1}\text{curl}\,\mathbf{F}_k) - \varepsilon\omega^2\mathbf{F}_k = i\omega\mathbf{j} \quad \text{in } \Omega,$$

$$\text{curl}\,\mathbf{F}_k \times v + i\omega v \times (\mathbf{F}_k \times v) = 0 \quad \text{on } \partial\Omega. \tag{254}$$

The remark we made for problem (250) also applies here. In fact

$$E \in \mathscr{L}\left(V, \prod_{k=1}^{K} \mathscr{V}(\Omega_k)\right),$$

and each $\mathbf{F}_k$ satisfies:

$$\text{curl}\,\mathbf{F}_k \times v \in L^2_t(\partial\Omega).$$

This permits us to define

$$S \in \mathscr{L}(V), \tag{255}$$

by

$$x = Sy, \tag{256}$$

if and only if

$$\forall k \leqslant K, \qquad x_k = -\text{curl}\,\mathbf{F}_k \times v + i\omega v \times (\mathbf{F}_k \times v), \quad \mathbf{F}_k \text{ solution of (254)}, \tag{257}$$

and finally

$$A = S\Pi \in \mathscr{L}(V). \tag{258}$$

7. Numerical results

In this section, we present the results of 2D numerical simulations on simple test cases. The corresponding results aim at illustrating our theoretical results. In all cases, the domain Ω is a rectangle, as well as the subdomains Ω_k. The space discretization is obtained by the mixed finite element method described in Section 6.1, except the fact we work with a uniform mesh made of small squares (cf. [22] for a precise description of the corresponding discrete system).

7.1. An illustration of the interest of non local operators

In this section, Ω is the square $] - L/2, L/2[\times]0, L[$ and we consider here the two domains decomposition described in Section 5. We choose the frequency such that $\omega L = 6\pi$, i.e., the total size of the domain is 3 wavelengths times 3 wavelengths. The boundary conditions are the same as in Section 5. Our experiment corresponds to a point source located at the center of the left subdomain.

In a first simulation, we fix the stepsize $h = 0.1$ (corresponding to 10 points per wavelength), and we use the operator $T_{\alpha\beta}$ as transmission operator. We compare the convergence of the iterative algorithms when we take successively:

- $\alpha = 1$, $\beta = 0$ (identity operator)
- $\alpha = 1$, $\beta = 0.5$ (non local operator)

In Fig. 6, we plot on the same picture the variations of the normalized numerical residue as a function of the number of iterations for the two simulations. The result illustrates clearly the interest of the non local operator. One observes that the convergence is approximately the same with the two methods during the first iterations. This is due to the fact that we see essentially at the beginning the decay of the low frequency content of the error, which is here dominant because the point source is located sufficiently far from the interface (It would not be the case if we had put it closer, in which case the interest of the non local operator would appear even more clearly). After a while, identity operator has difficulties with the high frequencies, which is not the case with the non local operator.
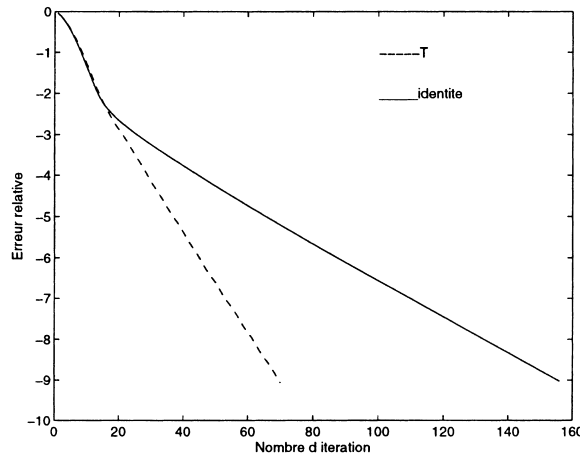

Fig. 6. Logarithm of the relative error versus the number of iterations. Comparison between operator $T_{\alpha,\beta}$ and identity operator (2 domains, $\omega L = 6\pi$ and $h = 0.1$).

In our second experiment, we would like to illustrate the influence of the space discretization on the convergence of the method. Let us first remark that after discretization the problem is finite dimensional and the convergence of the method is always exponential: in other words, the operator A^h of Section 5.1 is strictly contractant. However, taking into account the theoretical results relative to the continuous problem, one may expect that the rate of convergence will depend on h with the identity operator but not with the non local operator.

This is confirmed by the numerical simulations. In Figs. 7 and 8, we represent the variations of the normalized numerical residue as a function of the number of iterations for $h = 0.2, 0.1, 0.0625, 0.05$ respectively for the identity operator (Fig. 7) and the non local operator (Fig. 8). We observe that contrary to the case of the non local operator, with the identity, diminishing h deteriorates the convergence.
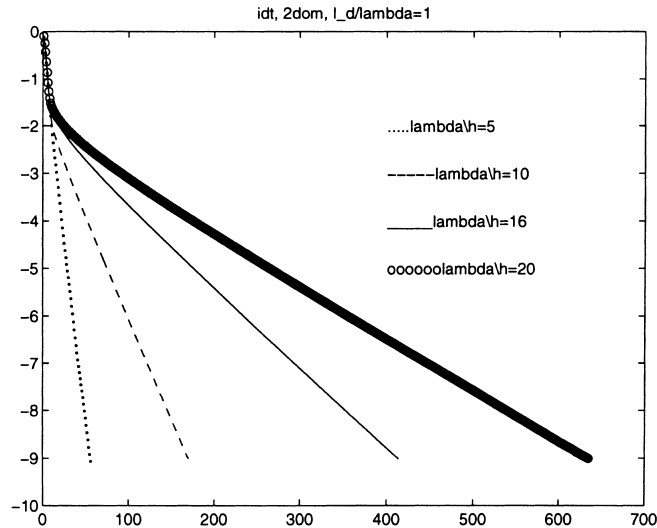

Fig. 7. Logarithm of the relative error versus the number of iterations. Case of the identity operator. The different curves correspond to different discretization step.
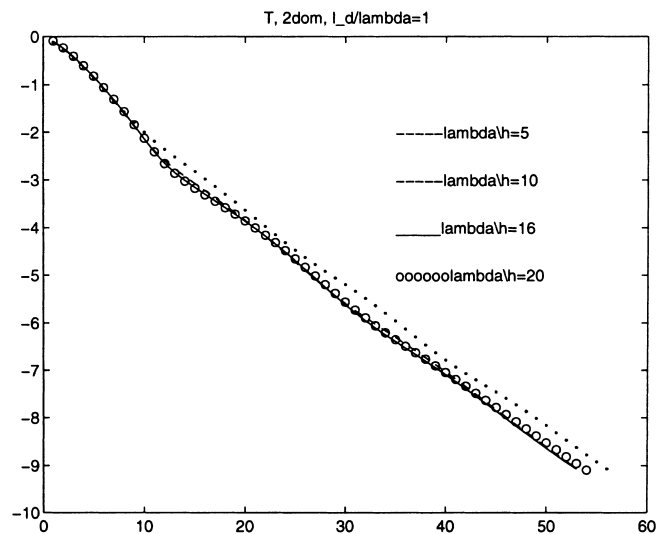

Fig. 8. Logarithm of the relative error versus the number of iterations. The different curves correspond to different discretization step. Case of operator $T_{\alpha,\beta}$.

7.2. A new choice: transparent like operators

In [21], we introduced the notion of transparent transmission operator which corresponds to define T_{jk} by (178) using

$$\alpha = 1, \quad \beta = -1. \tag{259}$$

In this case, as the operator $I + (1/\omega^2)\Delta_{jk}$ is no longer symmetric positive definite, the square root appearing in (178) has to be carefully defined. It appears, see [21], that the good choice is the one which is consistent (think to the diagonalisation of Δ_{jk}) with the complex square root with positive imaginary part. Of course, the operator T_{jk} is no longer symmetric positive definite, and our theory cannot be applied any longer: we have no guaranty of convergence for the corresponding iterative algorithms.

However, such a choice appears naturally if one looks at a "1D like" problem: consider a domain $\Omega \subset R^2$ as the strip $]-a, a[\times]0, L[$ and let us take the same boundary conditions as in Section 5. We want to decompose in vertical strips with interfaces $x = x_k$ as in our 1D model problem (cf. Section 1). Introducing, with the appropriate definition of the square root, the operator

$$T = \left(I + \frac{\partial_{yy}^2}{\omega^2} \right)^{1/2}. \tag{260}$$

The Helmholtz equation writes at least formally

$$\partial_{xx}^2 u + (\omega T)^2 u = f, \tag{261}$$

which as the same form as the 1D Helmholtz equation of Section 1, except that ω is replaced by ωT. This justifies to write transmission in the form

$$[\partial_x u + i\omega T u] = 0,$$

and thus to use the transmission operators T_{jk} introduced above.

In spite of the absence of theory in this case, we have chosen to implement these "transparent" transmission operators and to compare them with the other non local ones.

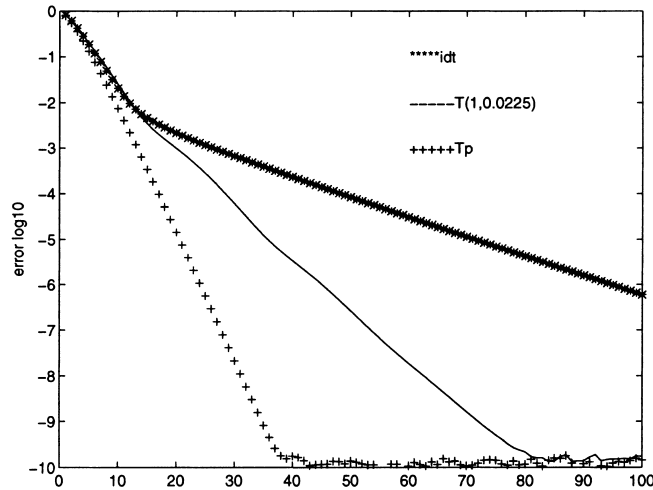Fig. 9 demonstrates clearly the improvement provided by the transparent operator.

Fig. 9. Logarithm of the relative error versus the number of iterations. Comparison between the transparent operator, $T_{\alpha,\beta}$ and identity.

Table 3
Computational time and optimal number of sub-domains for two different sizes of problem and two different strategy of splitting

Error	$\omega L = 8\pi$, layers	$\omega L = 8\pi$, squares	$\omega L = 12\pi$, layers	$\omega L = 12\pi$, squares
$\epsilon = 10^{-3}$	$T = 54s$, $Nd = 8$	$T = 20s$, $Nd = 25$	$T = 106s$, $Nd = 12$	$T = 32s$, $Nd = 36$
$\epsilon = 10^{-6}$	$T = 96s$, $Nd = 8$	$T = 34s$, $Nd = 25$	$T = 220s$, $Nd = 12$	$T = 70s$, $Nd = 30$

7.3. On the influence of the geometry of the partition of Ω

When using the domain decomposition method, one may wonder what is the best way to build the partitioning of the domain. As a matter of fact, it is a delicate question at its answer will necessarily depend on the used computer.

We have chosen to compare a splitting in vertical layers and a second one in small squares of equal surface. We make the numerical experiments for two sizes of domains $\omega L = 8\pi$ and $\omega L = 12\pi$ (i.e., the initial large domain is a square of respectively 4 then 6 wavelengths). For the two splittings, we have experimented the Jacobi algorithm for different size of subdomains (cutting in $Nd = 2, 4, \ldots, 20$ sub-layers or in $Nd = 4, 16, \ldots, 36$ squares) and compared the computation time required for achieving a given accuracy of 10^{-3} or 10^{-6}. As a result, the best splitting appears to be a layer of one half wavelength or a square of approximatively 1 wavelength. Furthermore, the partitioning in square gives better performance than the partitioning in layers (see Table 3).

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