
Absorbing boundary conditions and domain decomposition methods for nonlinear Schrödinger equations

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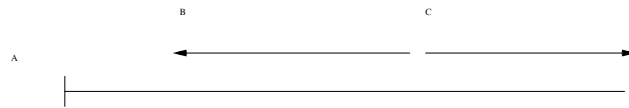
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**Absorbing Boundary Conditions for nonlinear
Schrödinger equations**

ABC for linear variable coefficients equations (Engquist-Majda)

1D wave equation : $\partial_t^2 - \partial_x^2 = -(\partial_x + \partial_t)(\partial_x - \partial_t)$.

We annihilate the reflected wave : $(\partial_x - \partial_t)u = 0$.



$$\begin{aligned} & \partial_t^2 - \partial_x^2 + \alpha(t, x) + \beta(t, x)\partial_t + \gamma(t, x)\partial_x \\ &= -(\partial_x - a(x, t, D_t))(\partial_x - b(x, t, D_t)) + S^{-\infty} \end{aligned}$$

where $a(x, t, D_t)$ and $b(x, t, D_t)$ are pseudodifferential operators

Transparent boundary condition : $(\partial_x - b(x, t, D_t))u = 0$

ABC for linear variable coefficients equations (Engquist-Majda)

$$a(x, t, \tau) = \sum_{j \geq 0} a_{1-j}(x, t, \tau) \text{ and } b(x, t, \tau) = \sum_{j \geq 0} b_{1-j}(x, t, \tau)$$

$$\left\{ \begin{array}{l} b_1 = i\tau \text{ and } a_1 = -i\tau, \\ b_0 = \frac{\gamma - \beta}{2} \text{ and } a_0 = \frac{\beta + \gamma}{2}, \\ b_{-1} = \frac{-\alpha + (\beta^2 - \gamma^2)/4 + (\partial_x - \partial_t)(\gamma - \beta)/2}{2i\tau} = -a_{-1}. \end{array} \right.$$

The k -th order absorbing boundary condition is

$$\left(\partial_x - \sum_{j=0}^k b_{1-j}(0, t, D_t) \right) u|_{x=0} = 0.$$

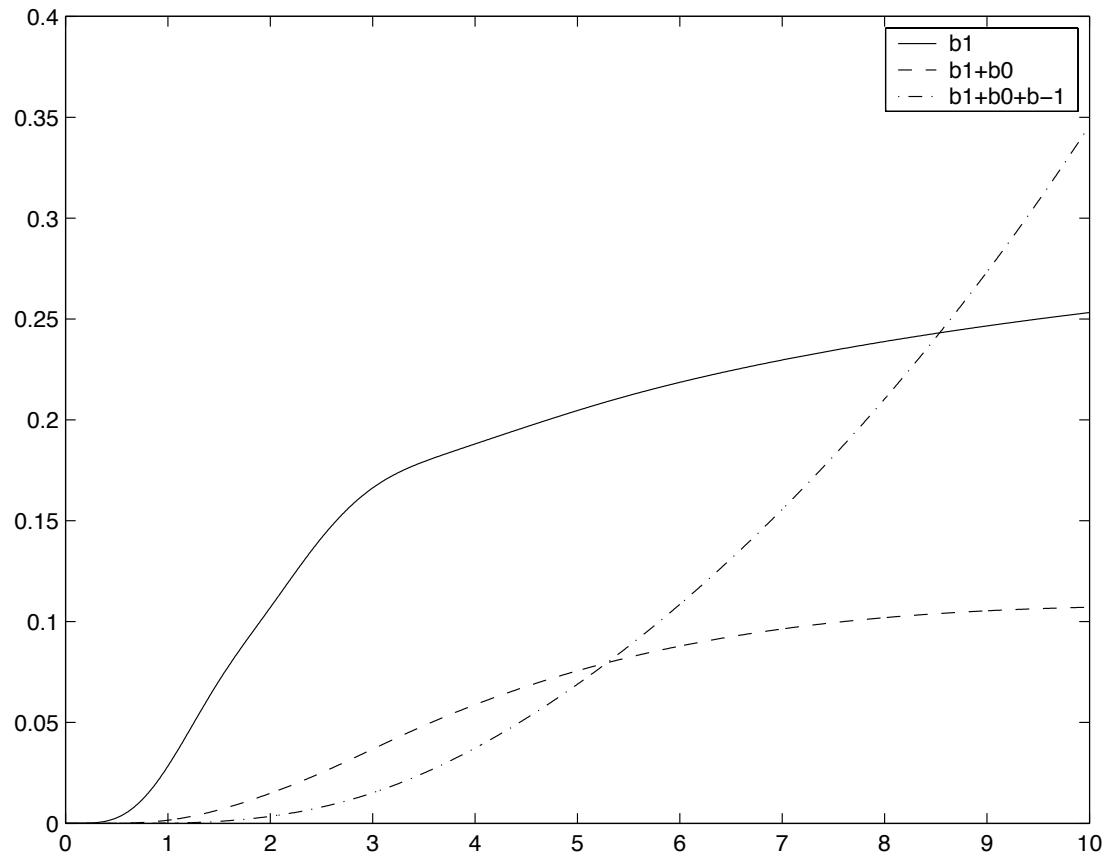


FIG. 1 — $\partial_t^2 u - \partial_x^2 u + \partial_t u = 0$. Relative error in L^2 in function of time. abc
 — order 0, — — order 1 and · — · order 2

Short-time behaviour of ABC

High frequency asymptotic expansions

Numerical frequencies = $[\pi/T, \pi/\Delta t]$

$T \ll 1 \Rightarrow$ only high frequencies

Thus, ABC always good for small times

Goal : We would like to recover the properties of the Engquist-Majda method in the nonlinear case

Potential strategy

Gauge change strategy

Linearization strategy

Numerical results

Potential strategy

$$(i\partial_t + \partial_x^2)u + |u|^2u = 0$$

$$V(t, x) = |u(t, x)|^2$$

$$(i\partial_t + \partial_x^2)u + Vu = 0$$

We can see this as a linear equation with a potential term and apply the Engquist-Majda method

⇒ We obtain BC involving V

$$\text{ABC of order 0 : } \partial_x u + \sqrt{-i\partial_t}u = 0$$

Potential strategy

$$(i\partial_t + \partial_x^2)u + |u|^2u = 0$$

$$\text{ABC order 1 : } \partial_x u + \sqrt{-i\partial_t}u = 0$$

$$\text{ABC order 2 : } \partial_x u + \sqrt{-i\partial_t}u + |u|^2/2\sqrt{-i\partial_t}^{-1}u = 0$$

ABC order 3 :

$$\partial_x u + \sqrt{-i\partial_t}u + |u|^2/2\sqrt{-i\partial_t}^{-1}u + i\partial_x(|u|^2)/4\partial_t^{-1}u = 0$$

$\sqrt{-i\partial_t}$ and $\sqrt{-i\partial_t}^{-1}$ are approximated using quadrature formulas

Other models : reaction-diffusion, semi-linear wave equation

Gauge change strategy

$$(i\partial_t + \partial_x^2)u + |u|^2u = 0$$

$$v(x, t) = e^{-i\mathcal{B}(t,x)}u(x, t), \text{ where } \mathcal{B}(t, x) = \int_0^t |u|^2(\tau, x)d\tau,$$

$$i\partial_t v + \partial_x^2 v + (2i\partial_x \mathcal{B})\partial_x v + (i\partial_x^2 \mathcal{B} - (\partial_x \mathcal{B})^2)v = 0$$

Linear equation for v with coefficients $2i\partial_x \mathcal{B}$ and $i\partial_x^2 \mathcal{B} - (\partial_x \mathcal{B})^2$

We may apply the Engquist-Majda method

\Rightarrow We obtain BC involving \mathcal{B}

$$\text{ABC of order 0 : } \partial_{\mathbf{n}} u + e^{-i\frac{\pi}{4}} e^{i\mathcal{B}} \partial_t^{1/2} (e^{-i\mathcal{B}} u) = 0$$

Gauge change strategy

$$(i\partial_t + \partial_x^2)u + |u|^2u = 0$$

$$\mathcal{B}(t, x) = \int_0^t |u|^2(\tau, x) d\tau$$

$$\text{ABC order 1 : } \partial_{\mathbf{n}}u + e^{-i\frac{\pi}{4}} e^{i\mathcal{B}} \partial_t^{1/2} (e^{-i\mathcal{B}}u) = 0$$

ABC order 3 :

$$\partial_{\mathbf{n}}u + e^{-i\frac{\pi}{4}} e^{i\mathcal{B}} \partial_t^{1/2} (e^{-i\mathcal{B}}u) - i \frac{\partial_{\mathbf{n}}(|u|^2)}{4} e^{i\mathcal{B}} \int_0^t (e^{-i\mathcal{B}}u) d\tau = 0$$

Potential and Gauge change strategy coincide at high frequencies,
but different for low frequencies

Linearization strategy

$$(i\partial_t + \partial_x^2)u + u\partial_x u = 0$$

Linearize around u : $(i\partial_t + \partial_x^2)v + u\partial_x v + \partial_x uv = 0$

This is a linear equation for v with coefficients $u(t, x)$ and $\partial_x u(t, x)$

We may apply the Engquist-Majda method

\Rightarrow We obtain BC for v involving u and $\partial_x u$

We have then to 'unlinearize'

For example, $(\partial_x + \sqrt{-i\partial_t})v + uv/2 = 0$

is unlinearized to give the first order ABC :

$$(\partial_x + \sqrt{-i\partial_t})u + u^2/4 = 0$$

Linearization strategy

$$(i\partial_t + \partial_x^2)u + u\partial_x u = 0$$

$$\text{ABC order 0 : } \partial_x u + \sqrt{-i\partial_t}u = 0$$

$$\text{ABC order 1 : } \partial_x u + \sqrt{-i\partial_t}u + u^2/4 = 0$$

$$\text{ABC order 2 : } \partial_x u + \sqrt{-i\partial_t}u + u^2/8 - \sqrt{-i\partial_t}^{-1}(u\partial_x u/4) = 0$$

Problem of the linearization strategy :

It is not always clear how to 'unlinearize'

Other model : semi-linear wave equation

Numerical results

$$(i\partial_t + \partial_x^2)u + |u|^2u = 0 : \text{relative error}$$

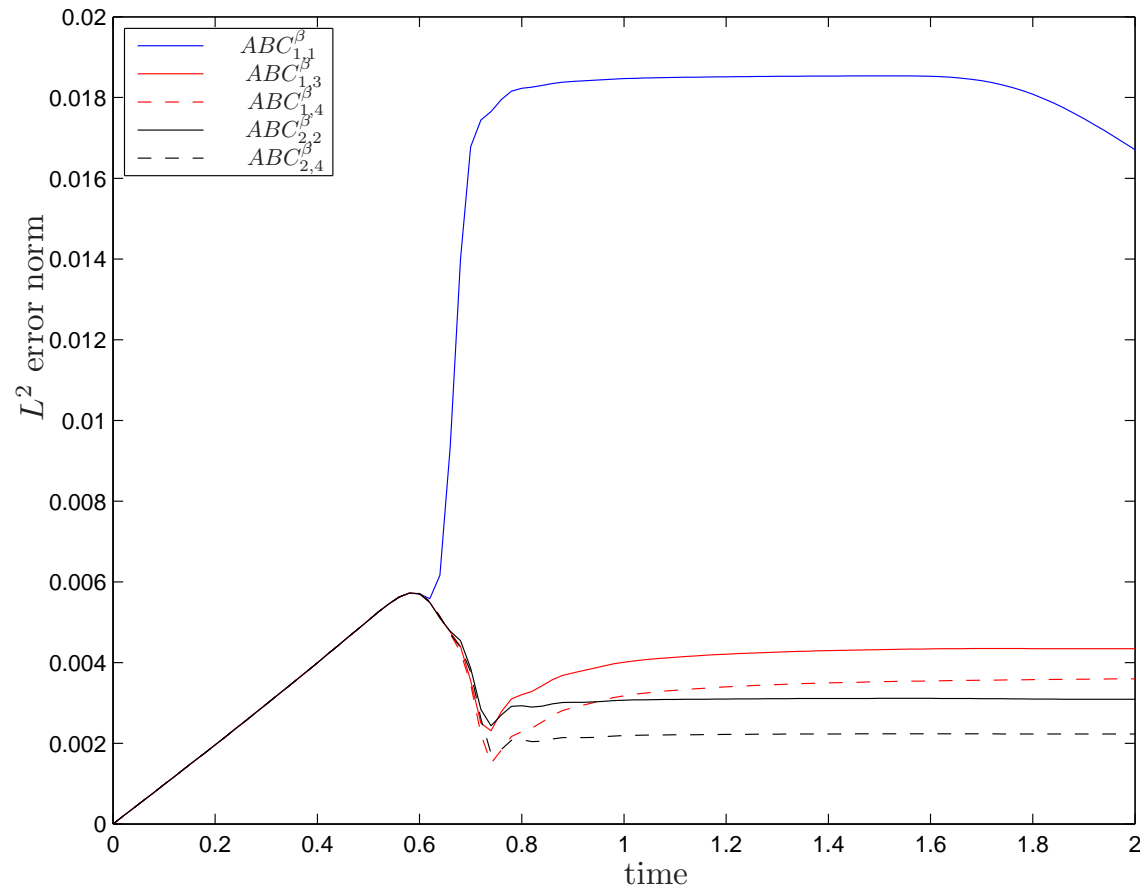


FIG. 2 – Comparison of potential and gauge change strategy

$$(i\partial_t + \partial_x^2)u + u\partial_x u = 0 : \text{relative error}$$

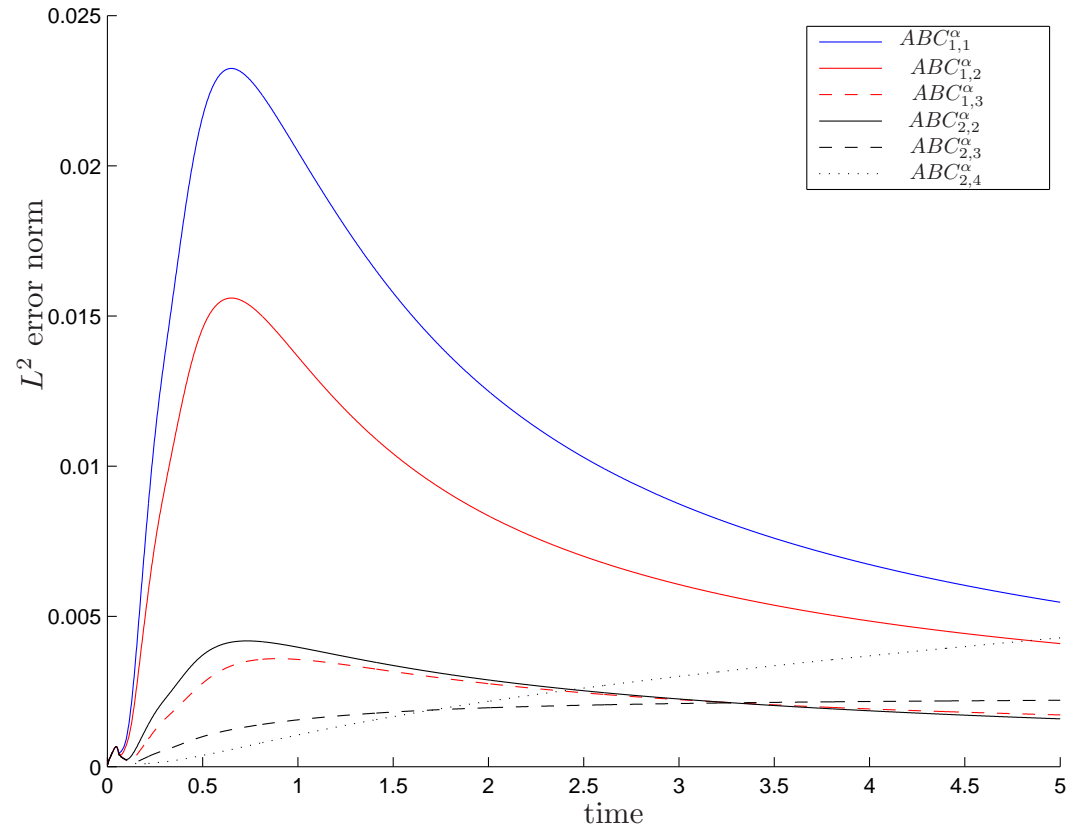


FIG. 3 – Comparison of potential and linearization strategy

Optimality of the ABC

$$\left\{ \begin{array}{l} (i\partial_t + \partial_x^2)u + u\partial_x u = 0 \text{ in }]0, T[\times]0, 2[\\ \partial_x u - \sqrt{-i\partial_t} u - \alpha u^2 = 0 \text{ at } x = 0 \\ \partial_x u + \sqrt{-i\partial_t} u + \alpha u^2 = 0 \text{ at } x = 2 \end{array} \right.$$

order zero : $\alpha = 0$, order 1 with potential strategy : $\alpha = 1/2$, and
 order 1 with linearization strategy : $\alpha = 1/4$

$\alpha = 0$	$\alpha = \frac{1}{8}$	$\alpha = \frac{1}{4}$	$\alpha = \frac{1}{2}$	$\alpha = 1$	$\alpha = 2$	$\alpha = 5$	$\alpha = -\frac{1}{4}$
0.0231	0.0136	0.0042	0.0155	0.0551	0.1390	0.3463	0.0601

TAB. 1 – Maximum of the relative error for $0 \leq t \leq 10$ and for various choices of α .

Conclusions and perspectives

- We have extended the method of B. Engquist and A. Majda to nonlinear problems in three ways : the potential strategy, the gauge change strategy and the linearization strategy
- Improve the method for long-time computations
- Extend the results to higher dimensions, curved boundaries and systems

Domain Decomposition Methods for linear Schrödinger equations

The Schrödinger equation

$$\begin{cases} i\partial_t u(t, x) + \partial_x^2 u(t, x) + V(x)u(t, x) = 0 \\ u(0, x) = u_0(x) \end{cases}$$

Quantum mechanics, electromagnetic wave propagation,
optic (Fresnel equation)

Goal : Design efficient Schwarz Waveform Relaxation algorithms
for the Schrödinger equation

Schwarz Waveform Relaxation algorithm
= global in time domain decomposition method

Classical Schwarz Waveform Relaxation

Optimal Schwarz Waveform Relaxation Algorithm

The Quasi-Optimal Algorithm

The Robin Algorithm

Numerical schemes

Numerical results

Classical Schwarz Waveform Relaxation

$$\left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_1^k = 0 \text{ in } \Omega_1 \times (0, T) \\ u_1^k(\cdot, 0) = u_0 \text{ in } \Omega_1 \\ u_1^k(L, \cdot) = u_2^{k-1}(L, \cdot) \text{ in } (0, T) \end{array} \right. \quad \left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_2^k = 0 \text{ in } \Omega_2 \times (0, T) \\ u_2^k(\cdot, 0) = u_0 \text{ in } \Omega_2 \\ u_2^k(0, \cdot) = u_1^{k-1}(0, \cdot) \text{ in } (0, T) \end{array} \right.$$

Convergence factor :

$$\Theta(\tau, L) = e^{-(\tau-V)^{1/2}L}$$

$\Theta(\tau, 0) = 1$: no convergence without overlap!

Optimal Schwarz Waveform Relaxation Algorithm

$$\left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_1^k = 0 \text{ in } \Omega_1 \times (0, T) \\ u_1^k(\cdot, 0) = u_0 \text{ in } \Omega_1 \\ (\partial_x + \mathcal{S}_1)u_1^k(L, \cdot) \\ = (\partial_x + \mathcal{S}_1)u_2^{k-1}(L, \cdot) \text{ in } (0, T) \end{array} \right. \left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_2^k = 0 \text{ in } \Omega_2 \times (0, T) \\ u_2^k(\cdot, 0) = u_0 \text{ in } \Omega_2 \\ (\partial_x + \mathcal{S}_2)u_2^k(0, \cdot) \\ = (\partial_x + \mathcal{S}_2)u_1^{k-1}(0, \cdot) \text{ in } (0, T) \end{array} \right.$$

Convergence in **2 iterations** if and only if

$$\sigma_1 = (\tau - V)^{1/2}, \quad \sigma_2 = -(\tau - V)^{1/2}$$

$$(\tau - V)^{1/2} = \begin{cases} \sqrt{\tau - V} & \text{if } \tau \geq V \\ -i\sqrt{-\tau + V} & \text{if } \tau < V \end{cases}$$

For non constant V : optimal operators not at hand

The Quasi-Optimal Algorithm

$$\left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_1^k = 0 \text{ in } \Omega_1 \times (0, T) \\ u_1^k(\cdot, 0) = u_0 \text{ in } \Omega_1 \\ (\partial_x + \sqrt{-i\partial_t - V(L)})u_1^k(L, \cdot) \\ = (\partial_x + \sqrt{-i\partial_t - V(L)})u_2^{k-1}(L, \cdot) \end{array} \right. \quad \left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_2^k = 0 \text{ in } \Omega_2 \times (0, T) \\ u_2^k(\cdot, 0) = u_0 \text{ in } \Omega_2 \\ (\partial_x - \sqrt{-i\partial_t - V(0)})u_2^k(0, \cdot) \\ = (\partial_x - \sqrt{-i\partial_t - V(0)})u_1^{k-1}(0, \cdot) \end{array} \right.$$

$$(\tau - V(x))^{1/2} = \begin{cases} \sqrt{\tau - V(x)} & \text{if } \tau \geq V(x) \\ -i\sqrt{-\tau + V(x)} & \text{if } \tau < V(x) \end{cases}$$

The algorithm converges in

$$\begin{aligned} & (H^{1/4}(0, T, L^2(\Omega_1)) \cap H^{-1/4}(0, T, H^1(\Omega_1))) \\ & \quad \times (H^{1/4}(0, T, L^2(\Omega_2)) \cap H^{-1/4}(0, T, H^1(\Omega_2))) \end{aligned}$$

The Complex Robin Algorithm

Zero order approximation of the Optimal Algorithm :

$$\left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_1^k = 0 \text{ in } \Omega_1 \times (0, T) \\ u_1^k(\cdot, 0) = u_0 \text{ in } \Omega_1 \\ (\partial_x - ip)u_1^k(L, \cdot) \\ = (\partial_x - ip)u_2^{k-1}(L, \cdot) \text{ in } (0, T) \end{array} \right. \left\{ \begin{array}{l} (i\partial_t + \partial_x^2 + V)u_2^k = 0 \text{ in } \Omega_2 \times (0, T) \\ u_2^k(\cdot, 0) = u_0 \text{ in } \Omega_2 \\ (\partial_x + ip)u_2^k(0, \cdot) \\ = (\partial_x + ip)u_1^{k-1}(0, \cdot) \text{ in } (0, T) \end{array} \right.$$

$$\rho(\tau, p, L) = \frac{ip + (\tau - V)^{1/2}}{ip - (\tau - V)^{1/2}} e^{-(\tau - V)^{1/2} L}$$

Convergence for any $p > 0$ even without overlap

Optimization with respect to $p > 0$ to improve the convergence

Numerical schemes

Finite volumes discretization

Interior = Crank Nicolson scheme

Quasi-Optimal algorithm : discretize $\sqrt{-i\partial_t + V}$

Approximation of Arnold and Ehrhardt

$$\sqrt{-i\partial_t + V}U(0, n) \simeq \sum_{m=0}^n S(n - m)U(0, m)$$

where $S(m)$ is given by a recurrence formula

Other possible approximations : Antoine and Besse, ...

Numerical results

Optimal p for the Robin algorithm

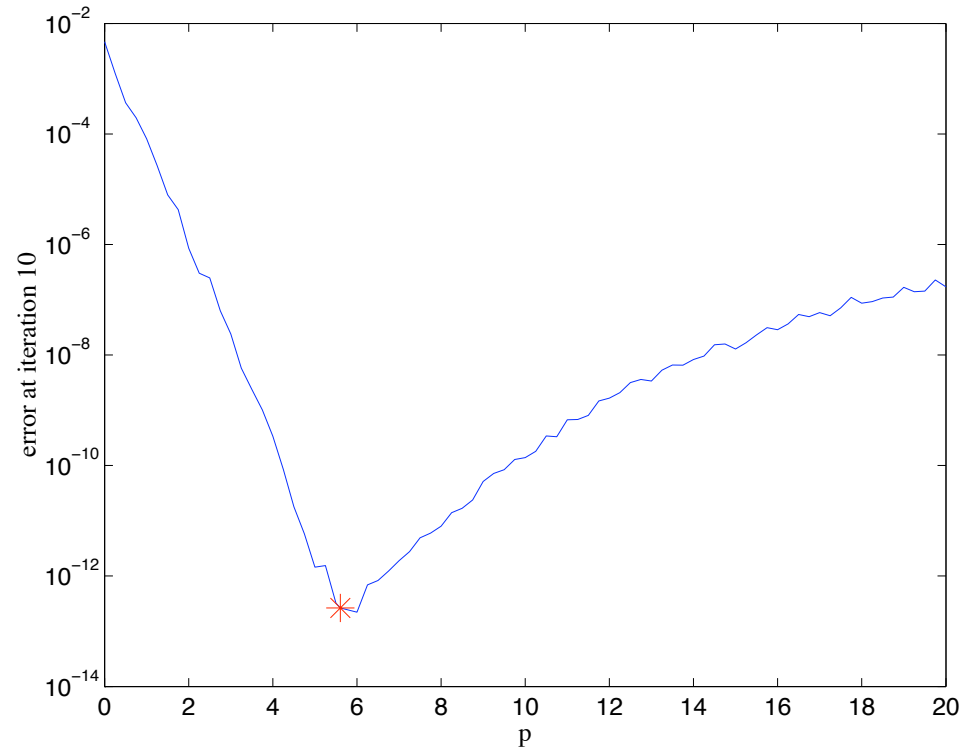


FIG. 4 – Variation of the quadratic error in time and space in Ω_1 as a function of p . The overlap is equal to 1%. The star corresponds to the theoretical optimal value p_T . Free Schrödinger equation

Comparison Classical/Robin algorithms

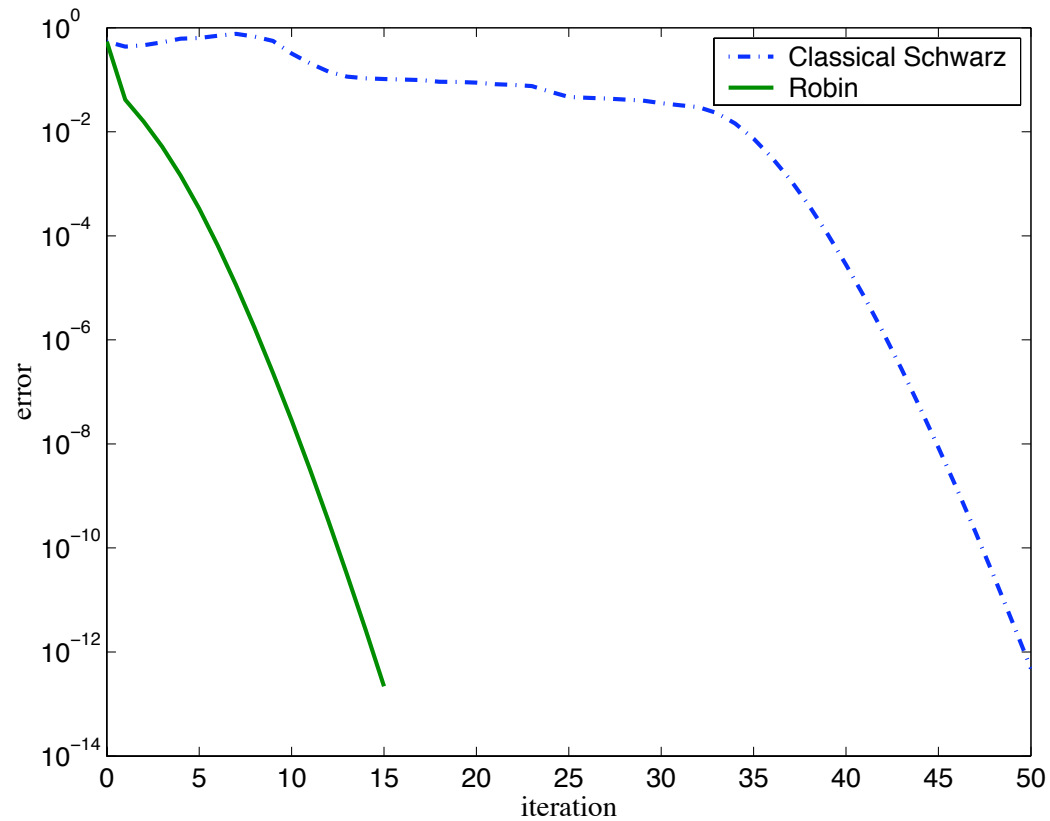


FIG. 5 – Convergence history : comparison of the Classical and Optimized Robin Schwarz algorithm. Free Schrödinger equation. $\delta = 2\%$

Robin algorithm without overlap

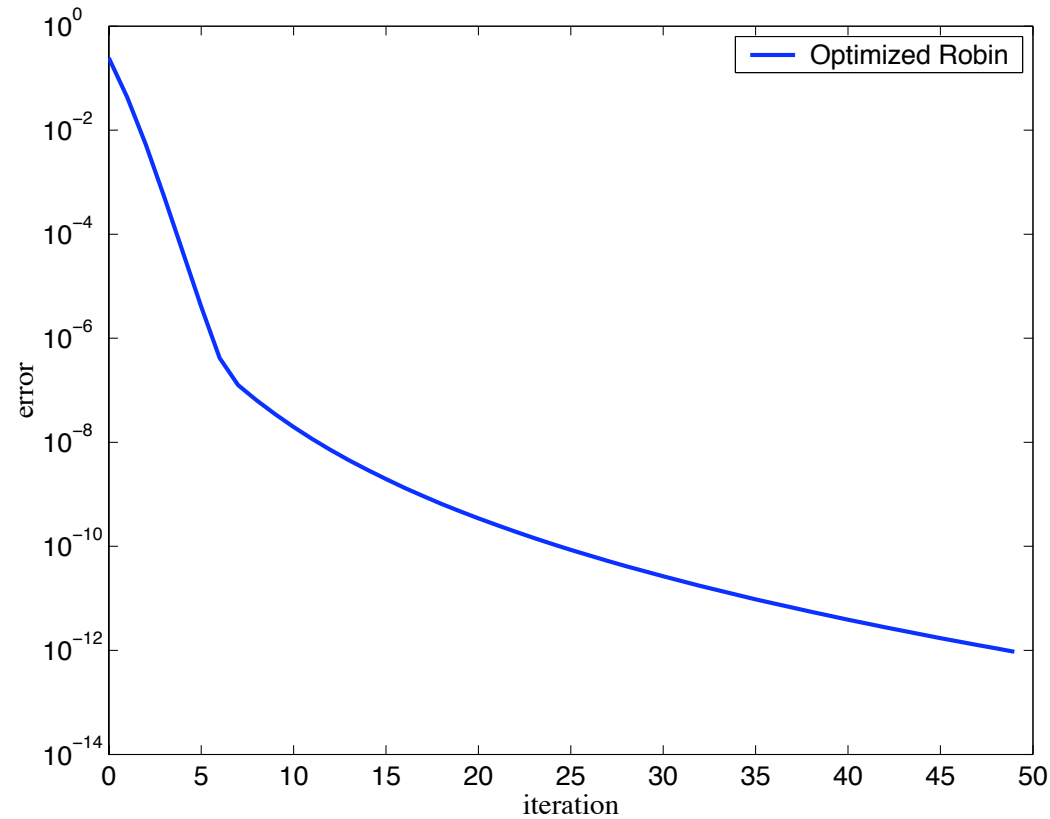


FIG. 6 – Convergence history for the Optimized Robin Schwarz algorithm in the non overlapping case. Free Schrödinger equation

Comparison Classical/Robin algorithms

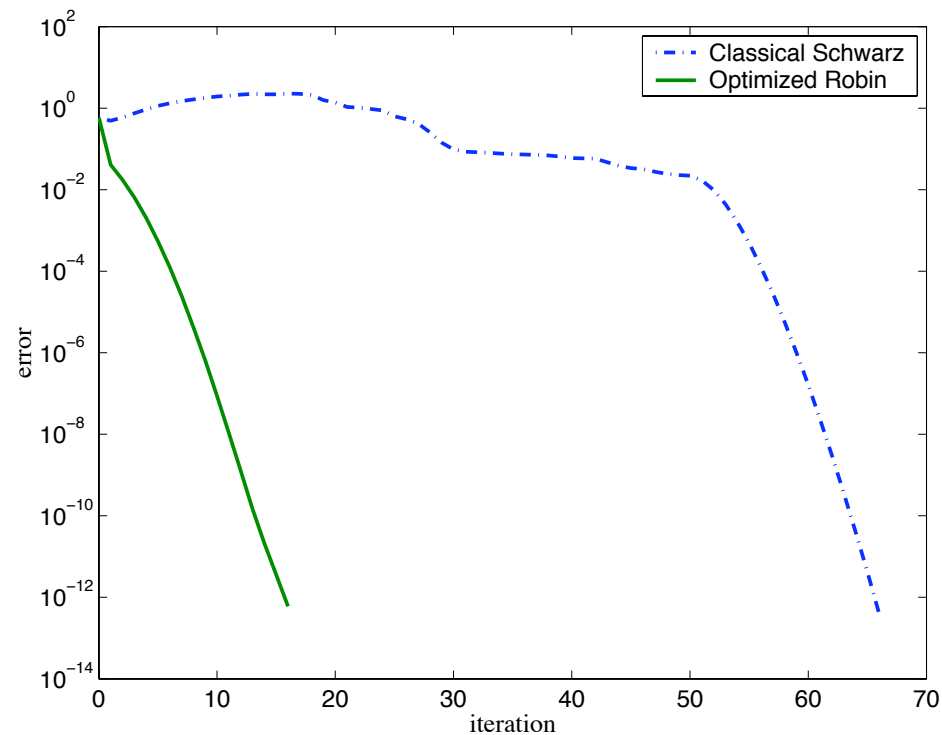


FIG. 7 – Convergence history : comparison of the Classical and Optimized Robin Schwarz algorithm for a [potential barrier](#). The overlap is equal to 4%

The Quasi-Optimal algorithm

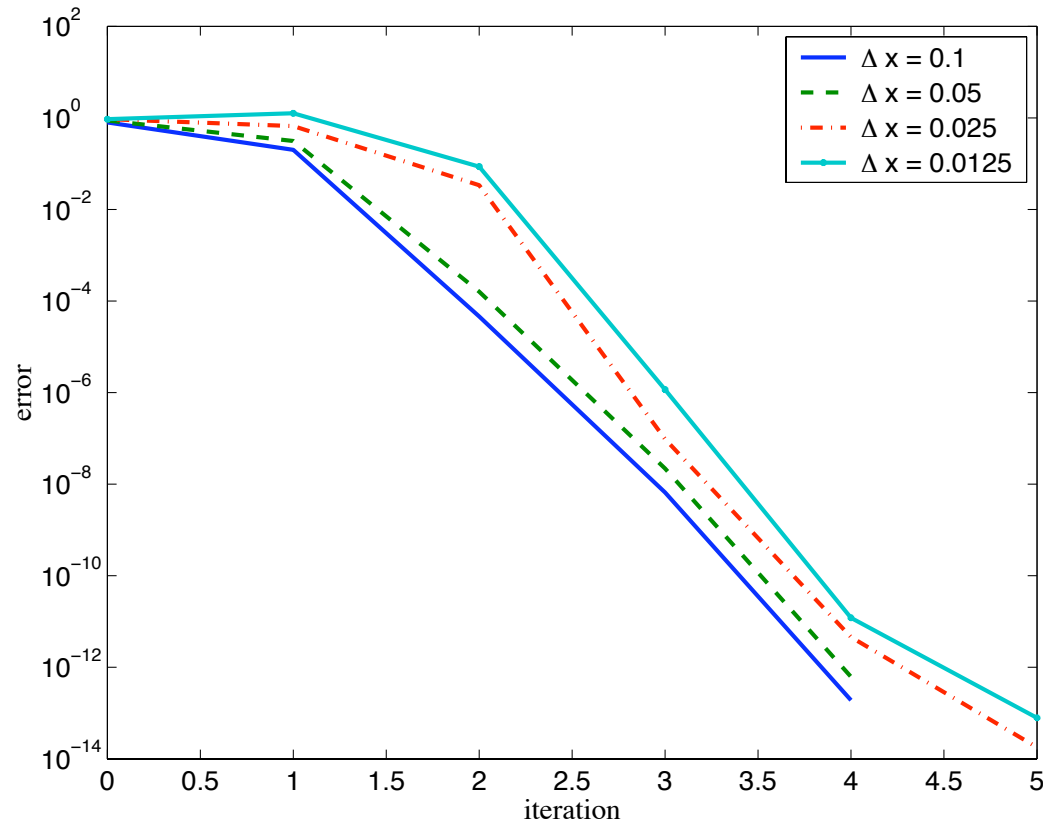


FIG. 8 – Convergence history for the Quasi-Optimal Schwarz algorithm in presence of a [parabolic potential](#)

Conclusions and perspectives

- The Classical algorithm converges extremely slowly for the Schrödinger equation with or without a potential
- We have designed two alternative algorithms : a Complex Optimized Robin algorithm and a Quasi-Optimal algorithm
- These algorithms greatly improve the performances of the classical Schwarz relaxation algorithm
- Extend the results to higher dimensions, nonlinear problems
- Use ABC to improve the transmission conditions