

# A Schwarz Waveform Relaxation Method for Advection–Diffusion–Reaction Problems with Discontinuous Coefficients and Non-matching Grids

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# Outline

## 1 Motivations and problem setting

## 2 Domain decomposition

- First iterative algorithm
- Improved transmission conditions

## 3 Some theory

- Subdomain problem
- Convergence of the iterative allgorithm

## 4 Numerical method and results

- Description of the numerical scheme
- Examples

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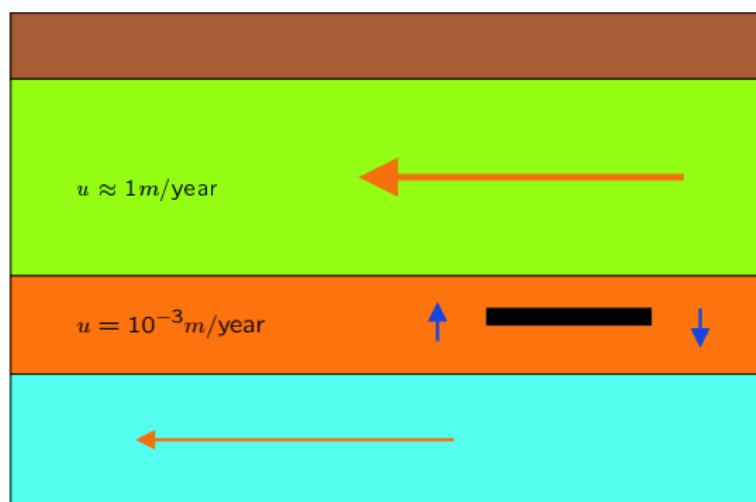
## 4 Numerical method and results

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# Nuclear Waste Deep Storage

Widely **varying** coefficients ( $1 - 10^{-6}$ ), very **long** simulation times ( $10^6$  years).

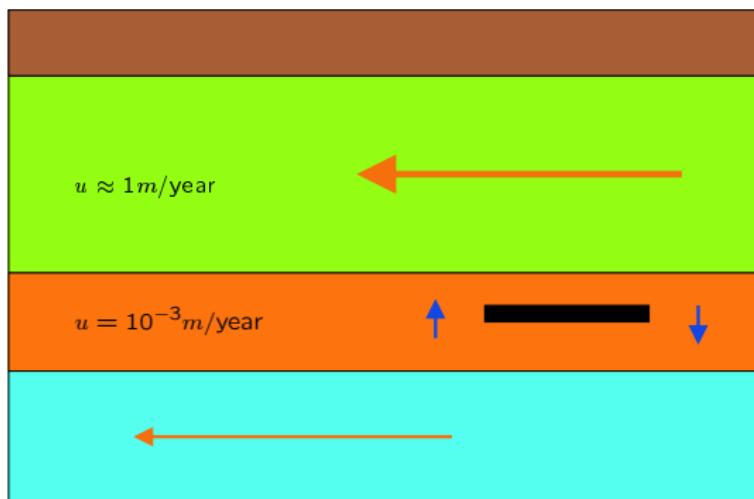
Example : COUPLEX (Comp. Geosc., 2004)



# Nuclear Waste Deep Storage

Widely **varying** coefficients ( $1 - 10^{-6}$ ), very **long** simulation times ( $10^6$  years).

Example : COUPLEX (Comp. Geosc., 2004)



Method with **different time steps** in each layer ?

# Mathematical Model

1D convection–diffusion–reaction equation, discontinuous coefficients

$$\begin{cases} \frac{\partial \textcolor{red}{u}}{\partial t} - \frac{\partial}{\partial x} \left( \textcolor{green}{D} \frac{\partial \textcolor{red}{u}}{\partial x} - \textcolor{green}{a} u \right) + \textcolor{blue}{b} u = \textcolor{blue}{f}, \text{ on } \mathbf{R} \times [0, T] \\ \textcolor{red}{u}(x, 0) = \textcolor{blue}{u}_0(x), \quad x \in \mathbf{R} \end{cases}$$

$\textcolor{green}{D}$  Molecular diffusion

$\textcolor{green}{a}$  Darcy velocity

$\textcolor{blue}{b}$  Radioactive decay

$$(\textcolor{green}{D}, \textcolor{green}{a}) = \begin{cases} (\textcolor{green}{D}^-, \textcolor{green}{a}^-) & x < 0 \\ (\textcolor{green}{D}^+, \textcolor{green}{a}^+) & x > 0 \end{cases}$$

# Mathematical Model

1D convection–diffusion–reaction equation, discontinuous coefficients

$$\begin{cases} \frac{\partial \mathbf{u}}{\partial t} - \frac{\partial}{\partial x} \left( D \frac{\partial \mathbf{u}}{\partial x} - a \mathbf{u} \right) + b \mathbf{u} = \mathbf{f}, \text{ on } \mathbf{R} \times [0, T] \\ \mathbf{u}(x, 0) = \mathbf{u}_0(x), \quad x \in \mathbf{R} \end{cases}$$

$D$  Molecular diffusion

$a$  Darcy velocity

$b$  Radioactive decay

$$(D, a) = \begin{cases} (D^-, a^-) & x < 0 \\ (D^+, a^+) & x > 0 \end{cases}$$

Weak solution  $\mathbf{u} \in L^\infty(0, T; L^2(\mathbf{R})) \cap L^2(0, T; H^1(\mathbf{R}))$  via standard variational theory

Notation:  $\mathbf{u}^- = \mathbf{u}|_{\mathbf{R}^-}$ ,  $\mathbf{u}^+ = \mathbf{u}|_{\mathbf{R}^+}$ .

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# Equivalent Transmission Problem

## Subdomain problems

$$\frac{\partial \mathbf{u}^-}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^- \frac{\partial \mathbf{u}^-}{\partial \mathbf{x}} - \mathbf{a}^- \mathbf{u}^- \right) + \mathbf{b} \mathbf{u}^- = \mathbf{f}, \quad \text{on } \mathbf{R}^- \times [0, T]$$

$$\mathbf{u}^-(\mathbf{x}, 0) = \mathbf{u}_0(\mathbf{x}), \quad \mathbf{x} \in \mathbf{R}^-$$

$$\frac{\partial \mathbf{u}^+}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^+ \frac{\partial \mathbf{u}^+}{\partial \mathbf{x}} + \mathbf{a}^+ \mathbf{u}^+ \right) + \mathbf{b} \mathbf{u}^+ = \mathbf{f}, \quad \text{on } \mathbf{R}^+ \times [0, T]$$

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$$\mathbf{u}^+(\mathbf{x}, 0) = \mathbf{u}_0(\mathbf{x}), \quad \mathbf{x} \in \mathbf{R}^+$$

## Transmission conditions

$$\mathbf{u}^+(0, t) = \mathbf{u}^-(0, t)$$

$$\left( \mathbf{a}^+ - \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} \right) \mathbf{u}^+(0, t) = \left( \mathbf{a}^- - \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} \right) \mathbf{u}^-(0, t)$$

# An iterative algorithm

## Algorithm with Dirichlet TC

$$\frac{\partial \textcolor{red}{u}_{k+1}^-}{\partial t} - \frac{\partial}{\partial x} \left( \textcolor{green}{D}^- \frac{\partial \textcolor{red}{u}_{k+1}^-}{\partial x} - \textcolor{green}{a}^- \textcolor{red}{u}_{k+1}^- \right) + \textcolor{blue}{b} \textcolor{red}{u}_{k+1}^- = \textcolor{blue}{f}, \quad \text{on } \mathbf{R}^- \times [0, T]$$

$$\textcolor{red}{u}_{k+1}^-(0, t) = \textcolor{red}{u}_k^+(0, t), \quad t \in [0, T]$$

# An iterative algorithm

## Algorithm with Dirichlet TC

$$\frac{\partial \mathbf{u}_{k+1}^-}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^- \frac{\partial \mathbf{u}_{k+1}^-}{\partial \mathbf{x}} - \mathbf{a}^- \mathbf{u}_{k+1}^- \right) + \mathbf{b} \mathbf{u}_{k+1}^- = \mathbf{f}, \quad \text{on } \mathbf{R}^- \times [0, T]$$

$$\mathbf{u}_{k+1}^-(0, t) = \mathbf{u}_k^+(0, t), \quad t \in [0, T]$$

$$\frac{\partial \mathbf{u}_{k+1}^+}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^+ \frac{\partial \mathbf{u}_{k+1}^+}{\partial \mathbf{x}} + \mathbf{a}^+ \mathbf{u}_{k+1}^+ \right) + \mathbf{b} \mathbf{u}_{k+1}^+ = \mathbf{f}, \quad \text{on } \mathbf{R}^+ \times [0, T]$$

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$$\mathbf{u}_{k+1}^+(0, t) = \mathbf{u}_k^-(0, t), \quad t \in [0, T]$$

Dirichlet TCs: **Slow** convergence

Acceleration possible by using **better** transmission conditions

# New transmission conditions

$$\left( \mathbf{a}^- - \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} - \Lambda^- \right) \mathbf{u}^-(0, t) = \left( \mathbf{a}^+ - \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} - \Lambda^- \right) \mathbf{u}^+(0, t)$$

$$\left( \mathbf{a}^+ - \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} + \Lambda^+ \right) \mathbf{u}^+(0, t) = \left( \mathbf{a}^- - \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} + \Lambda^+ \right) \mathbf{u}^-(0, t).$$

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$\Lambda^\pm$  (pseudo-differential) operators in time,  $\lambda^\pm$  symbol of  $\Lambda^\pm$  ( $\widehat{g}$  Fourier transform of  $g$ )

$$\forall g \in L^2(\mathbf{R}), \quad \widehat{\Lambda^\pm g}(\omega) = \lambda^\pm(\omega) \widehat{g}(\omega)$$

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$\Lambda^\pm$  (pseudo-differential) operators in time,  $\lambda^\pm$  symbol of  $\Lambda^\pm$  ( $\widehat{g}$  Fourier transform of  $g$ )

$$\forall g \in L^2(\mathbf{R}), \quad \widehat{\Lambda^\pm g}(\omega) = \lambda^\pm(\omega) \widehat{g}(\omega)$$

Still equivalent to original problem (if  $\Lambda^+ \neq \Lambda^-$  )

# New iterative algorithm

## Left subdomain

$$\frac{\partial \mathbf{u}_{k+1}^-}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^- \frac{\partial \mathbf{u}_{k+1}^-}{\partial \mathbf{x}} - \mathbf{a}^- \mathbf{u}_{k+1}^- \right) + \mathbf{b} \mathbf{u}_{k+1}^- = \mathbf{f}, \quad \text{on } \mathbf{R}^- \times [0, T]$$

$$\mathbf{u}^-(\mathbf{x}, 0) = \mathbf{u}_0(\mathbf{x}), \quad \mathbf{x} \in \mathbf{R}^-$$

$$\left( \mathbf{a}^- - \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} - \Lambda^- \right) \mathbf{u}_{k+1}^-(0, t) = \left( \mathbf{a}^+ - \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} - \Lambda^+ \right) \mathbf{u}_k^+(0, t)$$

# New iterative algorithm

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## Right subdomain

$$\frac{\partial \mathbf{u}_{k+1}^+}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^+ \frac{\partial \mathbf{u}_{k+1}^+}{\partial \mathbf{x}} + \mathbf{a}^+ \mathbf{u}_{k+1}^+ \right) + \mathbf{b} \mathbf{u}_{k+1}^+ = \mathbf{f}, \quad \text{on } \mathbf{R}^+ \times [0, T]$$

$$\mathbf{u}^+(\mathbf{x}, 0) = \mathbf{u}_0(\mathbf{x}), \quad \mathbf{x} \in \mathbf{R}^+$$

$$\left( \mathbf{a}^+ - \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} + \Lambda^+ \right) \mathbf{u}_{k+1}^+(0, t) = \left( \mathbf{a}^- - \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} + \Lambda^+ \right) \mathbf{u}_k^-(0, t)$$

# Properties of iterative algorithm

- Use **outgoing** BC on each subdomain
- If convergence, limit is solution to original problem

Equation for error  $e_k^\pm = u_k^\pm - u^\pm$

$$\frac{\partial e_{k+1}^\pm}{\partial t} - \frac{\partial}{\partial x} \left( D^\pm \frac{\partial e_{k+1}^\pm}{\partial x} - a^\pm e_{k+1}^\pm \right) + b e_{k+1}^- = 0, \quad \text{on } \mathbf{R}^\pm \times [0, T]$$
$$e^\pm(x, 0) = 0, \quad x \in \mathbf{R}^\pm$$

$$\left( a^- - D^- \frac{\partial}{\partial x} - \Lambda^- \right) e_{k+1}^-(0, t) = \left( a^+ - D^+ \frac{\partial}{\partial x} - \Lambda^- \right) e_k^+(0, t)$$

$$\left( a^+ - D^+ \frac{\partial}{\partial x} + \Lambda^+ \right) e_{k+1}^+(0, t) = \left( a^- - D^- \frac{\partial}{\partial x} + \Lambda^+ \right) e_k^-(0, t)$$

# Optimal transmission conditions

Equation for **error**, Fourier transform in time:

$$i\omega \hat{\mathbf{e}}^{\pm} - D^{\pm} \frac{d^2 \hat{\mathbf{e}}^{\pm}}{dx^2} + a^{\pm} \frac{d \hat{\mathbf{e}}^{\pm}}{dx} + b \hat{\mathbf{e}}^{\pm} = 0, \quad x \in \mathbb{R}^{\pm}$$

Characteristic equation

$$Dr^2 - ar - (b + i\omega) = 0$$

$r^+(\mathbf{a}, D, \omega)$  (resp.  $r^-(\mathbf{a}, D, \omega)$ ) is root with **positive** (resp. **negative**) real part

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$$\begin{cases} \hat{\mathbf{e}}_k^- = \alpha_k^-(\omega) e^{r^+(\mathbf{a}^-, D^-, \omega)x}, & x < 0 \\ \hat{\mathbf{e}}_k^+ = \alpha_k^+(\omega) e^{r^-(\mathbf{a}^+, D^+, \omega)x}, & x > 0 \end{cases}$$

# Convergence rate

Transmission conditions give

$$\alpha_k^+ (\mathbf{a}^+ - \mathbf{D}^+ r^- (\mathbf{a}^+, \mathbf{D}^+, \omega) + \lambda^+) = \alpha_{k-1}^- (\mathbf{a}^- - \mathbf{D}^- r^+ (\mathbf{a}^-, \mathbf{D}^-, \omega) + \lambda^+)$$

$$\alpha_k^- (\mathbf{a}^- - \mathbf{D}^- r^+ (\mathbf{a}^-, \mathbf{D}^-, \omega) - \lambda^-) = \alpha_{k-1}^+ (\mathbf{a}^+ - \mathbf{D}^+ r^- (\mathbf{a}^+, \mathbf{D}^+, \omega) - \lambda^-).$$

# Convergence rate

Transmission conditions give

$$\begin{aligned}\alpha_k^+ (\mathbf{a}^+ - \mathbf{D}^+ r^- (\mathbf{a}^+, \mathbf{D}^+, \omega) + \lambda^+) &= \alpha_{k-1}^- (\mathbf{a}^- - \mathbf{D}^- r^+ (\mathbf{a}^-, \mathbf{D}^-, \omega) + \lambda^+) \\ \alpha_k^- (\mathbf{a}^- - \mathbf{D}^- r^+ (\mathbf{a}^-, \mathbf{D}^-, \omega) - \lambda^-) &= \alpha_{k-1}^+ (\mathbf{a}^+ - \mathbf{D}^+ r^- (\mathbf{a}^+, \mathbf{D}^+, \omega) - \lambda^-).\end{aligned}$$

## Convergence rate

$$\rho(\omega) = \left( \frac{\mathbf{a}^- - \mathbf{D}^- r^+ (\mathbf{a}^-, \mathbf{D}^-, \omega) + \lambda^+}{\mathbf{a}^+ - \mathbf{D}^+ r^- (\mathbf{a}^+, \mathbf{D}^+, \omega) + \lambda^+} \right) \left( \frac{\mathbf{a}^+ - \mathbf{D}^+ r^- (\mathbf{a}^+, \mathbf{D}^+, \omega) - \lambda^-}{\mathbf{a}^- - \mathbf{D}^- r^+ (\mathbf{a}^-, \mathbf{D}^-, \omega) - \lambda^-} \right)$$

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Choose  $\lambda^+, \lambda^-$  to **minimize** convergence rate.

# Optimal and approximate transmission conditions

## Optimal choice

For

$$\lambda^-(\omega) = \mathbf{a}^+ - \mathbf{D}^+ r^-(\mathbf{a}^+, \mathbf{D}^+, \omega) = \frac{\sqrt{\Delta(\mathbf{a}^+, \mathbf{D}^+)} + \mathbf{a}^+}{2}$$
$$\lambda^+(\omega) = -\mathbf{a}^- + \mathbf{D}^- r^+(\mathbf{a}^-, \mathbf{D}^-, \omega) = \frac{\sqrt{\Delta(\mathbf{a}^-, \mathbf{D}^-)} - \mathbf{a}^-}{2}$$

the algorithm converges in 2 iterations.

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Operators **non-local** in time : need approximations

Approximate  $\sqrt{\Delta(\mathbf{a}, \mathbf{D})} = \sqrt{\mathbf{a}^2 + 4\mathbf{D}(\mathbf{b} + i\omega)}$  by **local** operators.

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Operators **non-local** in time : need approximations

Approximate  $\sqrt{\Delta(\mathbf{a}, \mathbf{D})} = \sqrt{\mathbf{a}^2 + 4\mathbf{D}(\mathbf{b} + i\omega)}$  by **local** operators.

Robin TC Take  $\sqrt{\Delta^\pm} \approx p^\pm$  (constant)

First order TC Take  $\sqrt{\Delta^\pm} \approx p^\pm + iq^\pm\omega$  (cf Absorbing Boundary Conditions)

# Robin transmission conditions

Gander, Halpern, Japhet, Martin, Nataf.

$$\left( \mathcal{D}^- \frac{\partial}{\partial x} - \mathbf{a}^- + \lambda^+ \right) \mathbf{u}_{k+1}^-(0, t) = \left( \mathcal{D}^+ \frac{\partial}{\partial x} - \mathbf{a}^+ + \lambda^+ \right) \mathbf{u}_k^+(0, t),$$

$$\left( \mathcal{D}^+ \frac{\partial}{\partial x} - \mathbf{a}^+ - \lambda^- \right) \mathbf{u}_{k+1}^+(0, t) = \left( \mathcal{D}^- \frac{\partial}{\partial x} - \mathbf{a}^- - \lambda^- \right) \mathbf{u}_k^-(0, t)$$

$$\lambda^- = \frac{\mathbf{p}^+ + \mathbf{a}^+}{2}, \quad \lambda^+ = \frac{\mathbf{p}^- - \mathbf{a}^-}{2}$$

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$$\left( \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} - \mathbf{a}^- + \lambda^+ \right) \mathbf{u}_{k+1}^-(0, t) = \left( \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} - \mathbf{a}^+ + \lambda^+ \right) \mathbf{u}_k^+(0, t),$$

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$$\lambda^- = \frac{\mathbf{p}^+ + \mathbf{a}^+}{2}, \quad \lambda^+ = \frac{\mathbf{p}^- - \mathbf{a}^-}{2}$$

- Low frequency approximation :  $\mathbf{p}^\pm = \sqrt{\mathbf{a}^\mp 2 + 4 \mathbf{b}^\mp \mathbf{D}^\mp}$
- Optimized coefficients : take  $\mathbf{p}^\pm$  to minimize convergence rate

# Iterative algorithm with Robin transmission conditions

Iterative algorithm: given  $g_0^\pm$  on  $[0, T]$

$$\frac{\partial \mathbf{u}_{k+1}^-}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^- \frac{\partial \mathbf{u}_{k+1}^-}{\partial \mathbf{x}} - \mathbf{a}^- \mathbf{u}_{k+1}^- \right) + \mathbf{b} \mathbf{u}_{k+1}^- = \mathbf{f}, \quad \text{on } \mathbf{R}^- \times [0, T]$$

$$\left( \mathbf{a}^- - \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} - \lambda^- \right) \mathbf{u}_{k+1}^-(0, t) = g_k^+(t)$$

$$\frac{\partial \mathbf{u}_{k+1}^+}{\partial t} + \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^+ \frac{\partial \mathbf{u}_{k+1}^+}{\partial \mathbf{x}} + \mathbf{a}^+ \mathbf{u}_{k+1}^+ \right) + \mathbf{b} \mathbf{u}_{k+1}^+ = \mathbf{f}, \quad \text{on } \mathbf{R}^+ \times [0, T]$$

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$$\frac{\partial \mathbf{u}_{k+1}^+}{\partial t} + \frac{\partial}{\partial \mathbf{x}} \left( \mathbf{D}^+ \frac{\partial \mathbf{u}_{k+1}^+}{\partial \mathbf{x}} + \mathbf{a}^+ \mathbf{u}_{k+1}^+ \right) + \mathbf{b} \mathbf{u}_{k+1}^+ = \mathbf{f}, \quad \text{on } \mathbf{R}^+ \times [0, T]$$

$$\left( \mathbf{a}^+ - \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} + \lambda^+ \right) \mathbf{u}_{k+1}^+(0, t) = g_k^-(t)$$

$$g_{k+1}^-(t) = \left( \mathbf{a}^- - \mathbf{D}^- \frac{\partial}{\partial \mathbf{x}} + \lambda^+ \right) \mathbf{u}_{k+1}^-(0, t)$$

$$g_{k+1}^+(t) = \left( \mathbf{a}^+ - \mathbf{D}^+ \frac{\partial}{\partial \mathbf{x}} - \lambda^- \right) \mathbf{u}_{k+1}^+(0, t)$$

# Outline

1 Motivations and problem setting

2 Domain decomposition

- First iterative algorithm
- Improved transmission conditions

3 Some theory

- Subdomain problem
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- Description of the numerical scheme
- Examples

# Interlude : Anisotropic Sobolev spaces

Needed for **boundary regularity** (Lions–Magenes, vol. 2)

## Definition

$$H^{r,s}(\Omega \times (0, T)) = L^2(0, T; H^r(\Omega)) \cap H^s(0, T; L^2(\Omega))$$

For  $\textcolor{red}{u} \in H^{2,1}(\Omega \times (0, T))$ , ( $j = 0, 1, 2$ ,  $k = 0, 1$ ):

$$\frac{\partial^j}{\partial \mathbf{x}^j} \frac{\partial^k}{\partial t^k} \textcolor{red}{u} \in H^{2\nu, \nu}(\Omega \times (0, T)), \quad \nu = 1 - (j/2 + k)$$

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## Theorem

**Trace space** For  $\textcolor{red}{u} \in H^{2,1}(\Omega \times (0, T))$

$$\textcolor{red}{u}(x, 0) \in H^1(\Omega), \quad \frac{\partial^j \textcolor{red}{u}}{\partial x^j}(0, t) \in H^{3/4-j/2}(0, T), \quad j = 0, 1$$

(+ compatibility conditions)

# Well-posedness of subdomain problem

## Subdomain problem

$$\frac{\partial \textcolor{red}{v}}{\partial t} - \frac{\partial}{\partial \mathbf{x}} \left( \textcolor{teal}{D}^- \frac{\partial \textcolor{red}{v}}{\partial \mathbf{x}} - \textcolor{green}{a}^- \textcolor{red}{v} \right) + \textcolor{blue}{b} \textcolor{red}{v} = \textcolor{blue}{f}, \quad \text{on } \mathbf{R}^- \times [0, T]$$

$$\textcolor{red}{v}(\mathbf{x}, 0) = \textcolor{blue}{u}_0(\mathbf{x}), \quad \text{on } \mathbf{R}^-$$

$$\left( \textcolor{teal}{D}^- \frac{\partial}{\partial \mathbf{x}} - \textcolor{green}{a}^- + \lambda^- \right) \textcolor{red}{v}(0, t) = \textcolor{blue}{g}^-(t), \quad t \in [0, T]$$

## Energy identity

$$\frac{1}{2} \frac{d}{dt} \|\textcolor{red}{v}\|^2 + \textcolor{teal}{D} \left\| \frac{\partial \textcolor{red}{v}}{\partial \mathbf{x}} \right\|^2 + \textcolor{blue}{b} \|\textcolor{red}{v}\|^2 - \left( \textcolor{teal}{D}^- \frac{\partial \textcolor{red}{v}}{\partial \mathbf{x}} - \frac{\textcolor{green}{a}^-}{2} \textcolor{red}{v} \right)(0) \textcolor{red}{v}(0) = (\textcolor{blue}{f}, \textcolor{red}{v})$$

# Existence and uniqueness

## Theorem

If  $u_0 \in H^1(\mathbf{R}^-)$ ,  $f \in L^2((0, T), L^2(\mathbf{R}^-))$ ,  $g^- \in H^{1/4}(0, T)$ ,  $\lambda^- + a^- > 0$

The subdomain problem has a unique solution  $u \in H^{2,1}((0, T) \times \mathbf{R}^-)$

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## Theorem

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The subdomain problem has a unique solution  $\mathbf{u} \in H^{2,1}((0, T) \times \mathbf{R}^-)$

## Proof (Bennequin, Gander, Halpern (04)).

- ① Lifting, trace theorem : reduce to  $\mathbf{u}_0 = 0$ ,  $\mathbf{g} = 0$ ;
- ② Standard estimates :  $\mathbf{u} \in L^\infty(0, T; L^2(\mathbf{R}^-)) \cap L^2(0, T; H^1(\mathbf{R}^-))$ ;
- ③ Non-standard estimates (multiply by  $\frac{\partial^2 \mathbf{v}}{\partial x^2}$ ) give more smoothness.



# Algorithm with Robin TC is well-defined

Smoothness needed for transmissions conditions

## Theorem

*Under same hypotheses as above, the algorithm is well defined : given  $(\mathbf{g}_0^-, \mathbf{g}_0^-) \in H^{1/4}(0, T)^2$ , the algorithm generates  $(\mathbf{u}_k^+, \mathbf{u}_k^-) \in H^{2,1}(\mathbb{R}^- \times (0, T))$ .*

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## Proof.

By trace theorem,  $t \rightarrow \mathbf{a}\mathbf{u}(0, \cdot) - \mathbf{D}\frac{\partial \mathbf{u}}{\partial \mathbf{x}}(0, \cdot) \in H^{1/4}(0, T)$

If initial guesses  $(\mathbf{g}_0^+, \mathbf{g}_0^-) \in H^{1/4}(\mathbf{R}^-) \times H^{1/4}(\mathbf{R}^+)$ , then **still true** for all iterates :  $(\mathbf{g}_k^+, \mathbf{g}_k^-)$ ,  $k \geq 1$ . □

# Convergence of iterative algorithm

## Theorem

- Same assumptions as above
- $\lambda^+ + \lambda^- \geq 0$ ,  $\lambda^+ - \lambda^- + \mathbf{a}^- > 0$ ,  $-\lambda^+ + \lambda^- + \mathbf{a}^+ > 0$

The sequence  $(\mathbf{u}_k^+, \mathbf{u}_k^-)$  converges to  $(\mathbf{u}^+, \mathbf{u}^-)$  in  $L^\infty(0, T; L^2(\mathbf{R}^-)) \cap L^2(0, T; H^1(\mathbf{R}^-)) \times L^\infty(0, T; L^2(\mathbf{R}^+)) \cap L^2(0, T; H^1(\mathbf{R}^+))$ .

## Proof.

By energy estimates (Despres (95), Lions (87), Bennequin, Gander, Halpern (04)).

$$\text{Define } \mathcal{E}_k^\pm = \frac{1}{2} \frac{d}{dt} \|\mathbf{e}_k^\pm\|^2 + \mathbf{D}^+ \left\| \frac{\partial \mathbf{e}_k^\pm}{\partial \mathbf{x}} \right\|^2 + b \|\mathbf{e}_k^\pm\|^2$$

$$\text{Also denote } \mathbf{B}^\pm \mathbf{v} = \mathbf{D}^\pm \frac{\partial \mathbf{v}}{\partial \mathbf{x}} - \mathbf{a}^\pm \mathbf{v}$$



# Proof of convergence theorem (ctd).

Energy estimate with transmission conditions

$$\begin{aligned}\mathcal{E}_k^- + \frac{1}{2(\lambda^+ + \lambda^-)} (B^- \mathbf{e}_k^- - \lambda^+ \mathbf{e}_k^-)^2 + (\lambda^+ - \lambda^- + \mathbf{a}^-) |\mathbf{e}_k^-(0)|^2 \\ = \frac{1}{2(\lambda^+ + \lambda^-)} (B^+ \mathbf{e}_{k-1}^+ - \lambda^- \mathbf{e}_{k-1}^+)^2\end{aligned}$$

$$\begin{aligned}\mathcal{E}_k^+ + \frac{1}{2(\lambda^+ + \lambda^-)} (B^+ \mathbf{e}_k^+ - \lambda^- \mathbf{e}_k^+)^2 + (-\lambda^+ + \lambda^- + \mathbf{a}^+) |\mathbf{e}_k^+(0)|^2 \\ = \frac{1}{2(\lambda^+ + \lambda^-)} (B^- \mathbf{e}_{k-1}^- - \lambda^+ \mathbf{e}_{k-1}^-)^2\end{aligned}$$

Add over  $k$  : telescopic sum.

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Add over  $k$  : telescopic sum.

# Optimization of convergence rate

Choose  $\lambda^\pm$  to minimize  $\max_{\omega \in [0, \omega_{\max}]} |\rho(\omega)|$ .

Numerical scheme :  $\omega_{\max} = \pi / \Delta t$ .

# Optimization of convergence rate

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Numerical scheme :  $\omega_{\max} = \pi / \Delta t$ .

Investigate how convergence rate depends on time step

Assume  $a^+, a^- > 0$ ,  $\Delta t$  “small”.

## Theorem (Asymptotic convergence rate)

If  $p^+ = p^- = p$ , then solution of the min-max problem is

$$p \approx \frac{\left(2^3 \pi (D^+ D^-) \left(a^+ - a^- + \sqrt{(a^+)^2 + 4D^+ b} + \sqrt{(a^-)^2 + 4D^- b}\right)^2\right)^{\frac{1}{4}}}{(\sqrt{D^+} + \sqrt{D^-})^{1/2}} \Delta t^{-\frac{1}{4}},$$

## Asymptotic bound on convergence rate

$$|\rho| \leq 1 - \left( \frac{2^5 (\sqrt{D^+} + \sqrt{D^-})^2 \left(a^+ - a^- + \sqrt{(a^+)^2 + 4D^+ b} + \sqrt{(a^-)^2 + 4D^- b}\right)^2}{D^+ D^- \pi} \right)^{\frac{1}{4}} \Delta t^{\frac{1}{4}}.$$

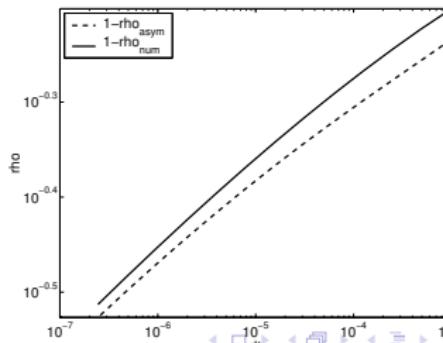
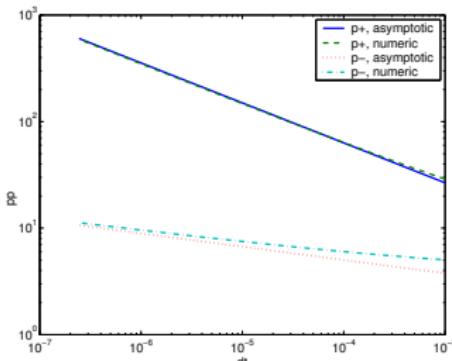
# Theorem

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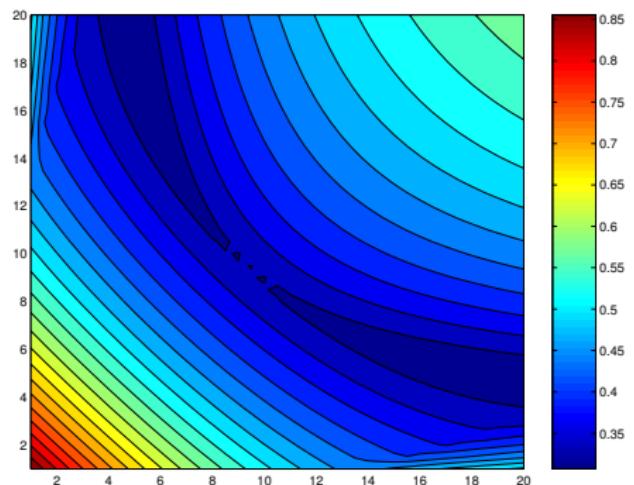
$$\begin{aligned} p^+ &\approx \left( 2^9 \pi^3 D^3 (a^+ - a^- + \sqrt{(a^+)^2 + 4Db} + \sqrt{(a^-)^2 + 4Db})^2 \right)^{\frac{1}{8}} \Delta t^{-\frac{3}{8}}, \\ p^- &\approx \left( 2^{-5} \pi D (a^+ - a^- + \sqrt{(a^+)^2 + 4Db} + \sqrt{(a^-)^2 + 4Db})^6 \right)^{\frac{1}{8}} \Delta t^{-\frac{1}{8}}, \end{aligned}$$

Asymptotic bound on the convergence rate

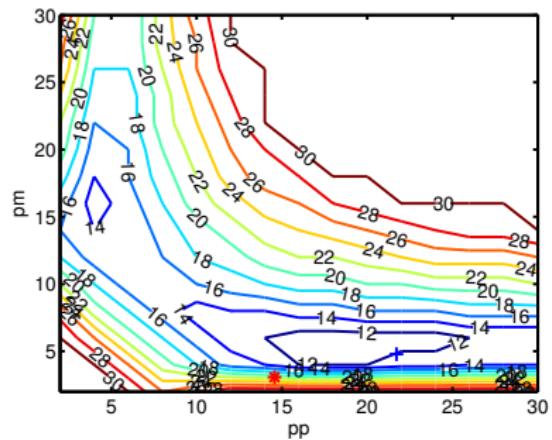
$$|\rho| \leq 1 - \left( \frac{2^{13} (a^+ - a^- + \sqrt{(a^+)^2 + 4Db} + \sqrt{(a^-)^2 + 4Db})^2}{D\pi} \right)^{\frac{1}{8}} \Delta t^{\frac{1}{8}}.$$



# Theoretical and numerical convergence rate



Theoretical convergence rate



Experimental convergence rate  
(blue cross: “optimal parameters”,  
red cross: asymptotic parameters)

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- First iterative algorithm
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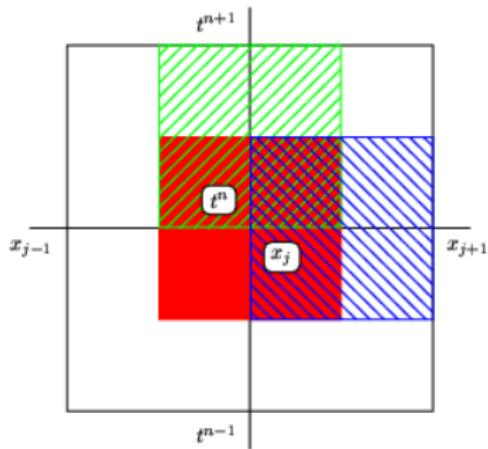
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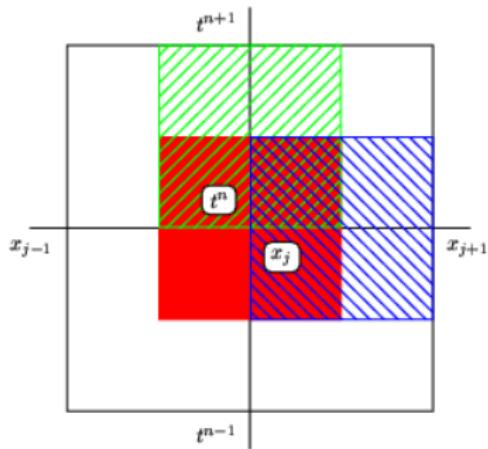
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# A Space–Time Finite Volume scheme



- Function constant on **square**;
- space and time derivatives defined by difference quotient on **staggered grids**, ;
- Implicit upwind scheme, finite difference in interior

# A Space–Time Finite Volume scheme



- Function constant on **square**;
- space and time derivatives defined by difference quotient on **staggered grids** ;
- Implicit upwind scheme, finite difference in interior

Green's formula :  $I_L + I_R + I_T + I_B = \int_{\text{square}} f$  with

$$I_{\text{side}} = \int_{\text{side}} \left( -(\mathbf{D} \frac{\partial \mathbf{u}}{\partial \mathbf{x}} - \mathbf{a} \mathbf{u}) \right) \cdot \begin{pmatrix} n_t \\ n_x \end{pmatrix} ds$$

# Interior scheme

3 points difference formula ( $u_j^{n+1/2} = \frac{u_j^n + u_j^{n-1}}{2}$ )

$$\begin{aligned}\frac{u_j^{n+1} - u_j^n}{\Delta t} - D \frac{u_{j+1}^{n+1/2} - 2u_j^{n+1/2} + u_{j-1}^{n+1/2}}{\Delta x^2} + a \frac{u_{j+1}^{n+1/2} - u_{j-1}^{n+1/2}}{2\Delta x} \\ - \frac{\gamma \Delta x}{2} |a| \frac{u_{j+1}^{n+1/2} - 2u_j^{n+1/2} + u_{j-1}^{n+1/2}}{\Delta x^2} + bu_j^{n+1/2} = f_j^{n+1/2}\end{aligned}$$

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$\gamma$  controls **upwinding** ( $\gamma = 0$ : centered,  $\gamma = 1$ : upwind)

Implicit scheme, unconditionally **stable**, order 1 for  $\gamma \neq 0$ , order 2 for  $\gamma = 0$

# Fourier analysis

Look for solution  $u_j^n = g(k)^n e^{ijk\Delta x}$

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Look for solution  $u_j^n = g(k)^n e^{ijk\Delta x}$

$$g(k) = \frac{1 - \frac{b\Delta t}{2} - \Delta t \left( \frac{2D}{\Delta x^2} + \frac{\gamma |a|}{\Delta x} \right) \sin^2 \frac{k\Delta x}{2} - i a \frac{\Delta t}{2\Delta x} \sin k\Delta x}{1 + \frac{b\Delta t}{2} - \Delta t \left( \frac{2D}{\Delta x^2} + \frac{\gamma |a|}{\Delta x} \right) \sin^2 \frac{k\Delta x}{2} + i a \frac{\Delta t}{2\Delta x} \sin k\Delta x}$$

# Fourier analysis

Look for solution  $u_j^n = \textcolor{red}{g}(k)^n e^{ijk\Delta x}$

$$\textcolor{red}{g}(k) = \frac{1 - \frac{b\Delta t}{2} - \Delta t \left( \frac{2D}{\Delta x^2} + \frac{\gamma |\textcolor{green}{a}|}{\Delta x} \right) \sin^2 \frac{k\Delta x}{2} - i\textcolor{green}{a} \frac{\Delta t}{2\Delta x} \sin k\Delta x}{1 + \frac{b\Delta t}{2} - \Delta t \left( \frac{2D}{\Delta x^2} + \frac{\gamma |\textcolor{green}{a}|}{\Delta x} \right) \sin^2 \frac{k\Delta x}{2} + i\textcolor{green}{a} \frac{\Delta t}{2\Delta x} \sin k\Delta x}$$

Can show that

$$|\textcolor{red}{g}(k)|^2 = 1 - \frac{4\alpha}{(1 + \alpha)^2 + \beta^2}.$$

with

$$\alpha = \frac{b\Delta t}{2} + \Delta t \left( \frac{2D}{\Delta x^2} + \frac{\gamma |\textcolor{green}{a}|}{\Delta x} \right) \sin^2 \frac{k\Delta x}{2} \geq 0, \quad \beta = \textcolor{green}{a} \frac{\Delta t}{2\Delta x} \sin k\Delta x.$$

# Interior scheme : examples

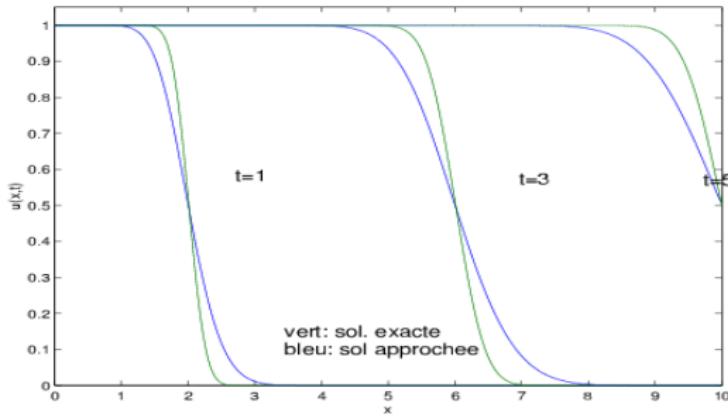
Solution on half-line, with Dirichlet left BC ( $\xi = \sqrt{a^2 + 4bD}$ )

$$u(x, t) = \frac{u_L}{2} \exp\left(\frac{ax}{2D}\right) \left\{ \exp\left(-\frac{x}{2D}\xi\right) \operatorname{erfc}\left(\frac{x - \sqrt{a^2 + 4bD}t}{2\sqrt{Dt}}\right) + \exp\left(\frac{x}{2D}\xi\right) \operatorname{erfc}\left(\frac{x + \sqrt{a^2 + 4bD}t}{2\sqrt{Dt}}\right) \right\},$$

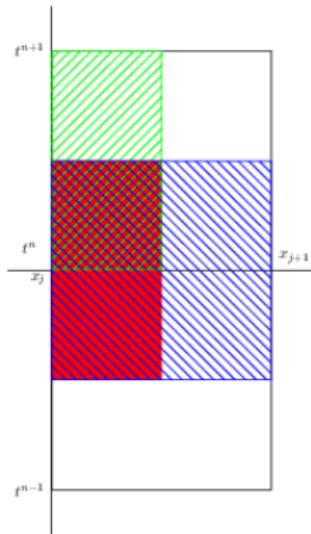
$$D = 0.02$$

$$a = 2$$

$$b = 0$$



# Numerical transmission conditions

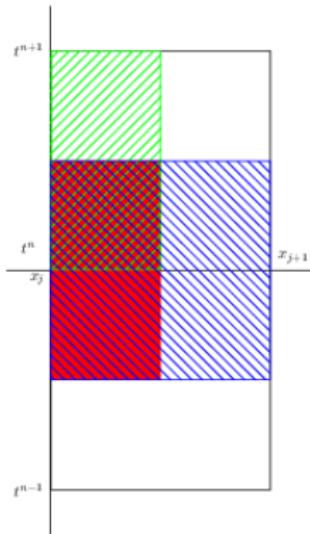


Integrate on  $]0, x_{1/2}[\times]t^n, t^{n+1}[$ , use TC to **close** system

On right subdomain ( $\gamma = 1$ : upwind scheme),

$$g^{+,n+1/2} = \frac{1}{\Delta t} \int_{t^n}^{t^{n+1}} g^+(t) dt$$

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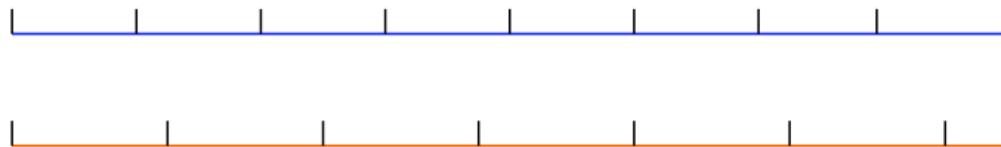
$$\boxed{\frac{\Delta x}{2} \frac{u_0^{+,n+1} - u_0^{+,n}}{\Delta t} - D^+ \frac{u_1^{+,n+1/2} - u_0^{+,n+1/2}}{\Delta x} + a^+ u_0^{+,n+1/2} + \left[ \frac{\Delta x}{2} b u_0^{+,n+1/2} \right] + \lambda^+ u_0^{+,n+1/2} = g^{+,n+1/2}}$$

# Numerical transmission conditions (contd.)

$$g^{+,n+1/2} = \boxed{-\frac{\Delta x}{2} \frac{u_0^{-,n+1} - u_0^{-,n}}{\Delta t} - D^- \frac{u_0^{-,n+1/2} - u_{-1}^{-,n+1/2}}{\Delta x}}$$
$$+ a^- u_{-1}^{-,n+1/2} + \boxed{\frac{\Delta x}{2} b u_0^{-,n+1/2}} + \lambda^- u_0^{-,n+1/2}$$

Consistent with interior scheme.

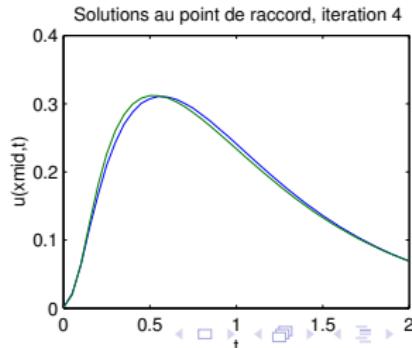
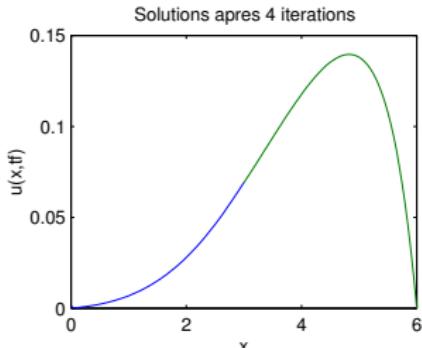
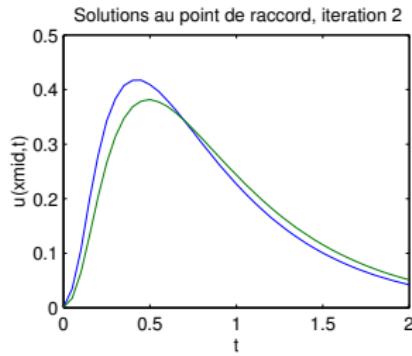
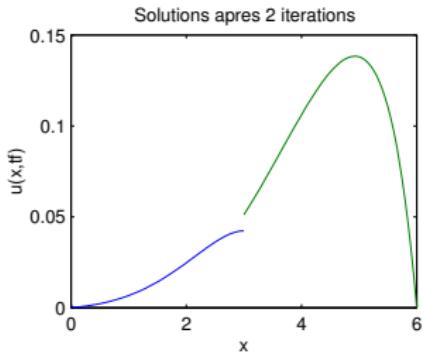
If different time steps, project  $g^+$  on left grid (recompute integral on other grid)



Matlab code (M. Gander)

# Homogeneous example

Homogeneous medium, with  $a = 2$ ,  $D = 1$ ,  $b = 0.1$ ,  
 $u_0(x) = e^{(-3(3/2-x)^2)}$ ,  $0 < x < 6$ . Interface at  $x = 3$ .



# Heterogeneous example

## Left subdomain [0, 1]

$$D^- = 4 \cdot 10^{-2}, \quad a^- = 4,$$
$$\Delta x^- = 10^{-2}, \quad \Delta t^- = 4 \cdot 10^{-3}$$

## Right subdomain [1, 1.8]

$$D^- = 12 \cdot 10^{-2}, \quad a^- = 2,$$
$$\Delta x^- = 8 \cdot 10^{-2}, \quad \Delta t^- = 10^{-2}$$

# Heterogeneous example

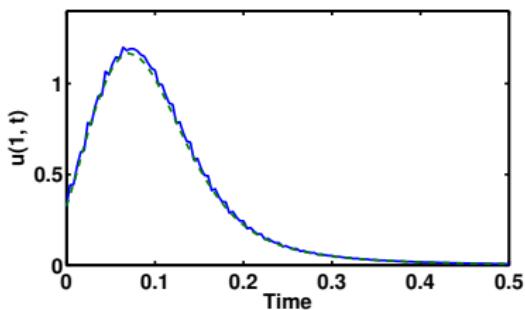
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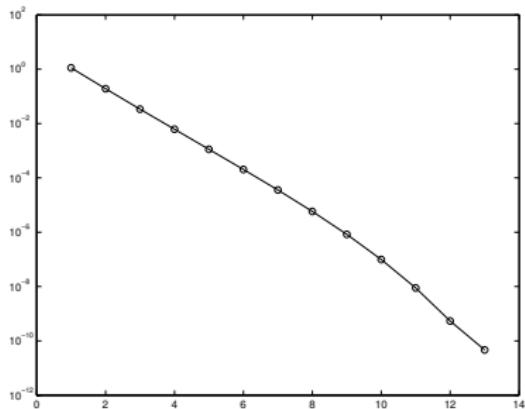
## Right subdomain $[1, 1.8]$

$$D^- = 12 \cdot 10^{-2}, \quad a^- = 2, \\ \Delta x^- = 8 \cdot 10^{-2}, \quad \Delta t^- = 10^{-2}$$

Solution on the interface, iteration 3.  
Solid line: left subdomain,  
dashed line: right subdomain



Solutions on the interface



Convergence history

# Heterogeneous example (ctd.)

Left subdomain  $[0, 1]$

$$D^- = 4 \cdot 10^{-2}, \quad a^- = 4, \\ \Delta x^- = 10^{-2}, \quad \Delta t^- = 10^{-3}$$

Right subdomain  $[1, 1.8]$

$$D^- = 12 \cdot 10^{-2}, \quad a^- = 2, \\ \Delta x^- = 2 \cdot 10^{-2}, \quad \Delta t^- = 2 \cdot 10^{-3}$$

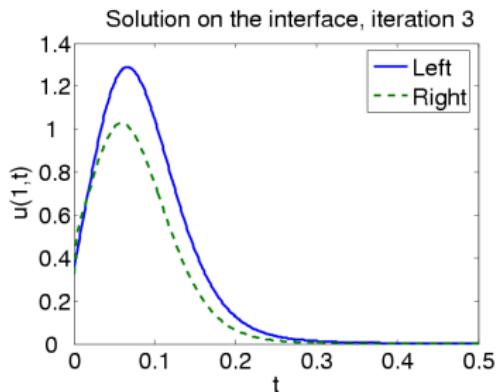
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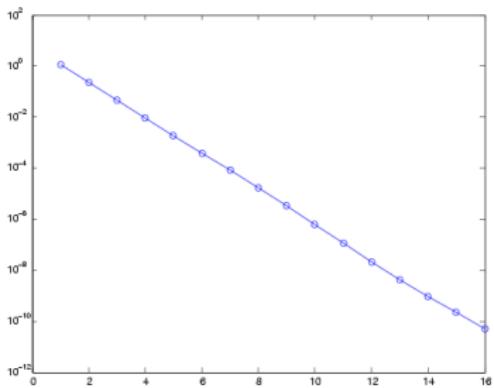
$$D^- = 4 \cdot 10^{-2}, \quad a^- = 4, \\ \Delta x^- = 10^{-2}, \quad \Delta t^- = 10^{-3}$$

Right subdomain  $[1, 1.8]$

$$D^- = 12 \cdot 10^{-2}, \quad a^- = 2, \\ \Delta x^- = 2 \cdot 10^{-2}, \quad \Delta t^- = 2 \cdot 10^{-3}$$



Solutions on the interface  
Sol. after 2 iterations  
Sol. at convergence



Convergence history

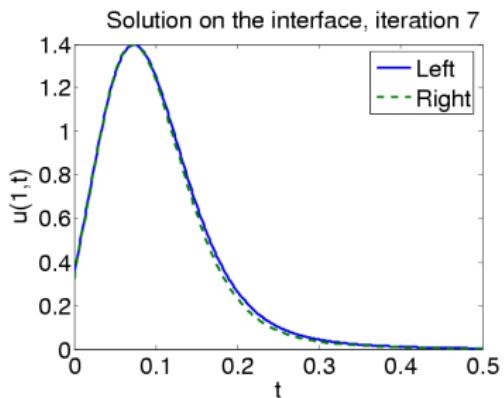
# Heterogeneous example (ctd.)

Left subdomain  $[0, 1]$

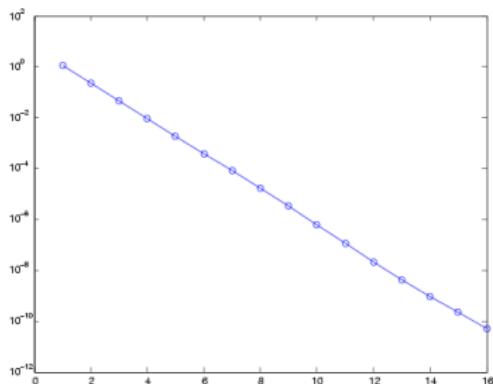
$$D^- = 4 \cdot 10^{-2}, \quad a^- = 4, \\ \Delta x^- = 10^{-2}, \quad \Delta t^- = 10^{-3}$$

Right subdomain  $[1, 1.8]$

$$D^- = 12 \cdot 10^{-2}, \quad a^- = 2, \\ \Delta x^- = 2 \cdot 10^{-2}, \quad \Delta t^- = 2 \cdot 10^{-3}$$



Solutions on the interface  
Sol. after 2 iterations  
Sol. at convergence



Convergence history



# Conclusions – perspectives

## Conclusions

- Method for CDR problems, discontinuous coefficients, different grids
- Optimized transmission conditions
- Satisfactory behavior on simple examples

## Further work

- More challenging test cases
- More subdomains, 2D
- Substructuring