

Time domain decomposition for PDE-constrained optimization

Liu-Di LU

Matematikcentrums
Lunds Universitet

International Workshop on PDE-constrained Optimization: Theory, Numerics and Applications

Tianyuan Mathematics Research Center

June 18, 2026



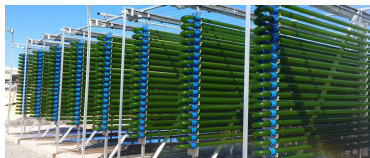
LUNDS
UNIVERSITET

Application: biomass production



Microalgae: photosynthetic micro-organisms 2-50 μm in aquatic environment

Application: biomass production

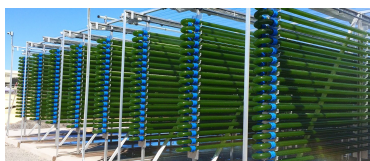


Microalgae: photosynthetic micro-organisms 2-50 μm in aquatic environment

Photobioreactors:

- technology: chemostats, raceway, tubular, etc
- position: indoor/outdoor
- scale: lab/industrial
- purpose: wastewater treatment, biofuel production, etc

Application: biomass production



Microalgae: photosynthetic micro-organisms 2-50 μm in aquatic environment

Photobioreactors:

- technology: chemostats, raceway, tubular, etc
- position: indoor/outdoor
- scale: lab/industrial
- purpose: wastewater treatment, biofuel production, etc

Challenges:

- both physical and biological behavior
- coupled models
- different timescales
- macro/micro levels
- uncertainties: environment, measurement, etc

Example: chemostats

Maximize biomass productivity $P(X)$

Governing equation: reaction-diffusion

$$\partial_t X - \kappa \Delta X + R(X) = f$$



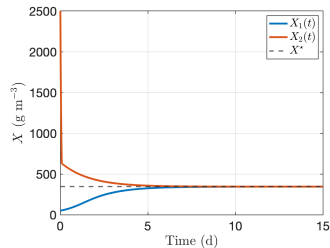
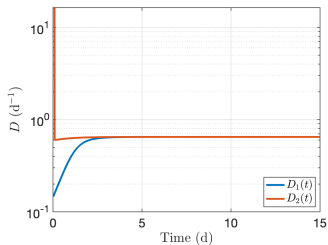
Example: chemostats

Maximize biomass productivity $P(X)$

Governing equation: reaction-diffusion

$$\partial_t X - \kappa \Delta X + R(X) = f$$

e.g., control with the dilution rate $R(X) = D(X)$



Example: raceway ponds



Paddle wheel

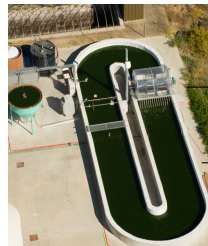
Water tank

Example: raceway ponds

Maximize average growth rate $\bar{\mu}(X)$

Governing physics: hydrodynamics

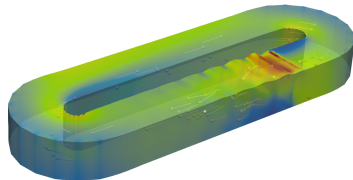
Navier-Stokes, Shallow water, etc



Paddle wheel

Water tank

Time: 73.6 s



Example: raceway ponds

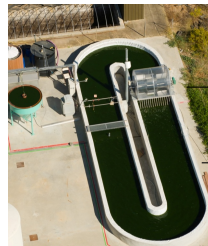
Maximize average growth rate $\bar{\mu}(X)$

Governing physics: hydrodynamics

Navier-Stokes, Shallow water, etc

Microalgae movement: advection-diffusion

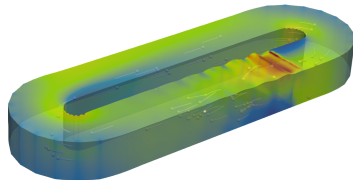
$$\partial_t X + \mathbf{u} \cdot \nabla X - \kappa \Delta X = f$$



Paddle wheel

Water tank

Time: 73.6 s



Example: raceway ponds

Maximize average growth rate $\bar{\mu}(X)$

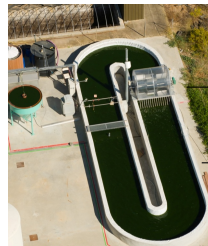
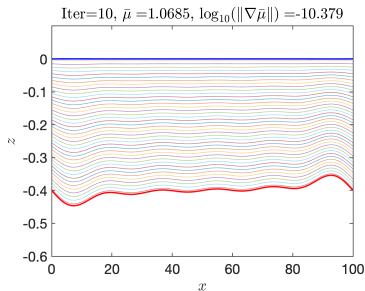
Governing physics: hydrodynamics

Navier-Stokes, Shallow water, etc

Microalgae movement: advection-diffusion

$$\partial_t X + \mathbf{u} \cdot \nabla X - \kappa \Delta X = f$$

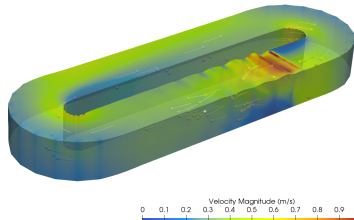
e.g., design best shape of the topography



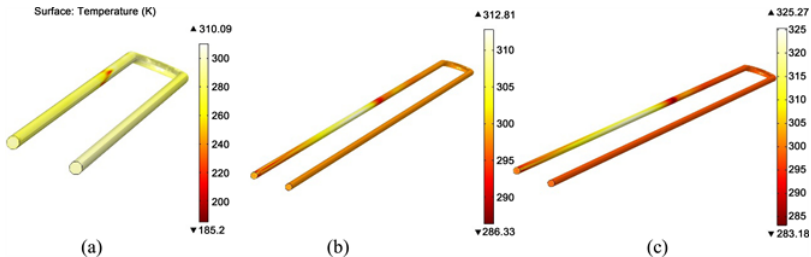
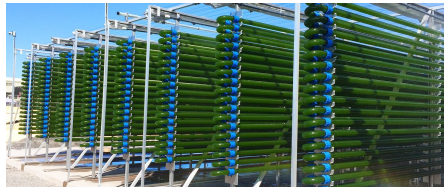
Paddle wheel

Water tank

Time: 73.6 s



Example: tubular photobioreactor



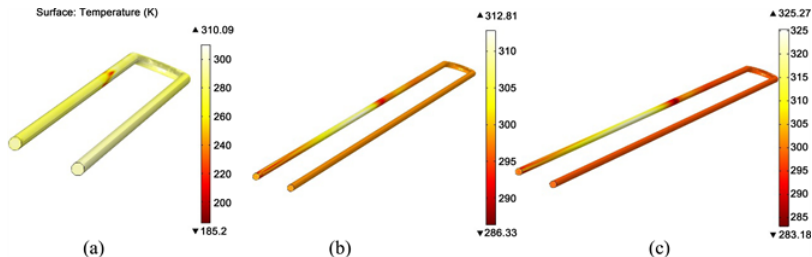
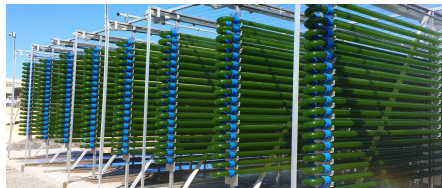
Example: tubular photobioreactor

Minimize temperature distribution w.r.t. target $\hat{y}$

$$J(y, u) = \frac{1}{2} \|y - \hat{y}\|_{L^2(Q)}^2 + \frac{\nu}{2} \|u\|_{L^2(Q)}^2$$

Governing equation: heat equation

$$\partial_t y - \kappa \Delta y = u$$



Model problem

Temperature control w.r.t. a target:

minimize the functional $J(y, u) := \frac{1}{2} \|y - \hat{y}\|_{L^2(Q)}^2 + \frac{\nu}{2} \|u\|_{L^2(Q)}^2$, $\hat{y} \in L^2(Q)$, $\nu > 0$

constrained by $\partial_t y - \Delta y = u$ in $Q := \Omega \times (0, T)$

completed with initial condition y_0 and boundary condition $y = 0$ on $\Sigma := \partial\Omega \times (0, T)$

Model problem

Temperature control w.r.t. a target:

minimize the functional $J(y, u) := \frac{1}{2} \|y - \hat{y}\|_{L^2(Q)}^2 + \frac{\nu}{2} \|u\|_{L^2(Q)}^2$, $\hat{y} \in L^2(Q)$, $\nu > 0$

constrained by $\partial_t y - \Delta y = u$ in $Q := \Omega \times (0, T)$

completed with initial condition y_0 and boundary condition $y = 0$ on $\Sigma := \partial\Omega \times (0, T)$

Lagrange multiplier p :

$$\mathcal{L}(y, p, u) = J(y, u) + \langle p, \partial_t y - \Delta y - u \rangle$$

Derive first-order optimality system

$$\begin{array}{llll} \partial_t y - \Delta y = u & \text{in } Q & \partial_t p + \Delta p = y - \hat{y} & \text{in } Q \\ y = 0 & \text{on } \Sigma & p = 0 & \text{on } \Sigma, \\ y = y_0 & \text{on } \Sigma_0 & p = 0 & \text{on } \Sigma_T \\ \nu u - p = 0 & \text{in } Q & & \end{array}$$

Model problem

Reduced optimality system (forward-backward)

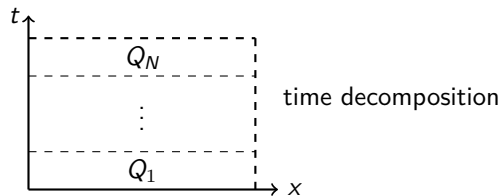
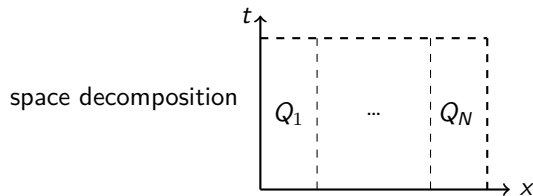
$$\begin{array}{llll} \partial_t y - \Delta y = \nu^{-1} p & \text{in } Q & \partial_t p + \Delta p = y - \hat{y} & \text{in } Q \\ y = 0 & \text{on } \Sigma & p = 0 & \text{on } \Sigma \\ y = y_0 & \text{on } \Sigma_0 & p = 0 & \text{on } \Sigma_T \end{array}$$

Model problem

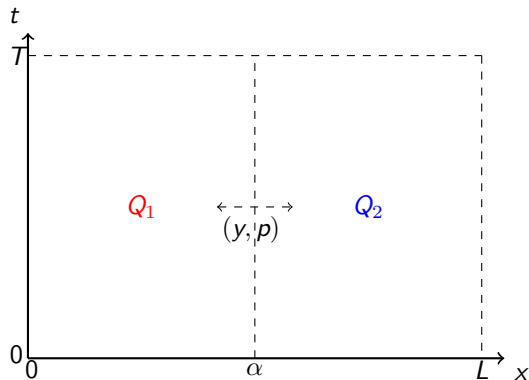
Reduced optimality system (forward-backward)

$$\begin{array}{ll} \partial_t y - \Delta y = \nu^{-1} p & \text{in } Q \\ y = 0 & \text{on } \Sigma \\ y = y_0 & \text{on } \Sigma_0 \end{array} \quad \begin{array}{ll} \partial_t p + \Delta p = y - \hat{y} & \text{in } Q \\ p = 0 & \text{on } \Sigma \\ p = 0 & \text{on } \Sigma_T \end{array}$$

Apply domain decomposition methods



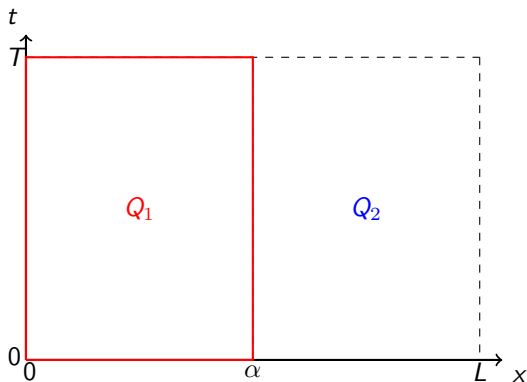
Space decomposition (Waveform method)



Subdomains: $Q_1 = (0, \alpha) \times (0, T)$ and
 $Q_2 = (\alpha, L) \times (0, T)$

$$\begin{aligned} \partial_t y - \partial_{xx} y &= \nu^{-1} p & \partial_t p + \partial_{xx} p &= y - \hat{y} \\ y(0, t) &= 0 & p(0, t) &= 0 \\ y(L, t) &= 0 & p(L, t) &= 0 \\ y(x, 0) &= y_0(x) & p(x, T) &= 0 \end{aligned}$$

Alternating Schwarz waveform



$$\text{In } Q_1 = (0, \alpha) \times (0, T)$$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$y_1^\ell(0, t) = 0$$

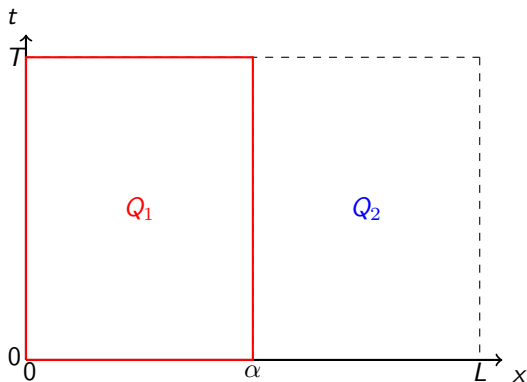
$$y_1^\ell(x, 0) = y_{1,0}(x)$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$p_1^\ell(0, t) = 0$$

$$p_1^\ell(x, T) = 0$$

Alternating Schwarz waveform



In $Q_1 = (0, \alpha) \times (0, T)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$y_1^\ell(0, t) = 0$$

$$y_1^\ell(\alpha, t) = y_2^{\ell-1}(\alpha, t)$$

$$y_1^\ell(x, 0) = y_{1,0}(x)$$

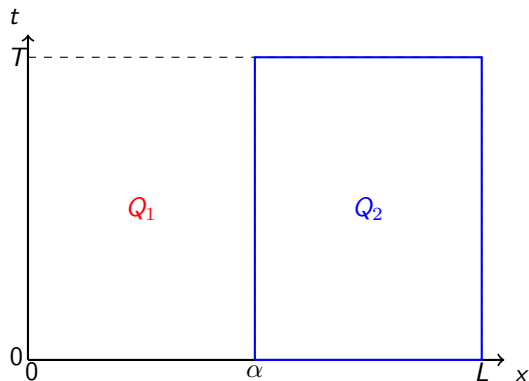
$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$p_1^\ell(0, t) = 0$$

$$p_1^\ell(\alpha, t) = p_2^{\ell-1}(\alpha, t)$$

$$p_1^\ell(x, T) = 0$$

Alternating Schwarz waveform



In $Q_1 = (0, \alpha) \times (0, T)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$y_1^\ell(0, t) = 0$$

$$y_1^\ell(\alpha, t) = y_2^{\ell-1}(\alpha, t)$$

$$y_1^\ell(x, 0) = y_{1,0}(x)$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$p_1^\ell(0, t) = 0$$

$$p_1^\ell(\alpha, t) = p_2^{\ell-1}(\alpha, t)$$

$$p_1^\ell(x, T) = 0$$

In $Q_2 = (\alpha, L) \times (0, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell$$

$$y_2^\ell(L, t) = 0$$

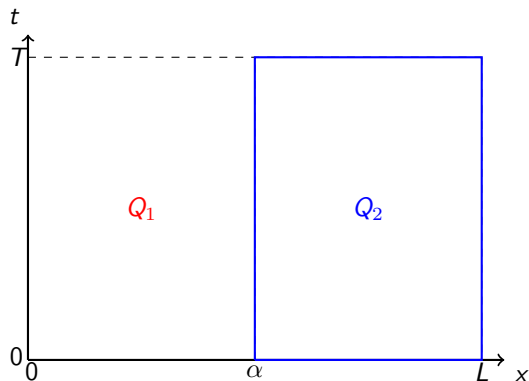
$$y_2^\ell(x, 0) = y_{2,0}(x)$$

$$\partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz waveform



In $Q_1 = (0, \alpha) \times (0, T)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$y_1^\ell(0, t) = 0$$

$$y_1^\ell(\alpha, t) = y_2^{\ell-1}(\alpha, t)$$

$$y_1^\ell(x, 0) = y_{1,0}(x)$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$p_1^\ell(0, t) = 0$$

$$p_1^\ell(\alpha, t) = p_2^{\ell-1}(\alpha, t)$$

$$p_1^\ell(x, T) = 0$$

In $Q_2 = (\alpha, L) \times (0, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell$$

$$y_2^\ell(\alpha, t) = y_1^\ell(\alpha, t)$$

$$y_2^\ell(L, t) = 0$$

$$y_2^\ell(x, 0) = y_{2,0}(x)$$

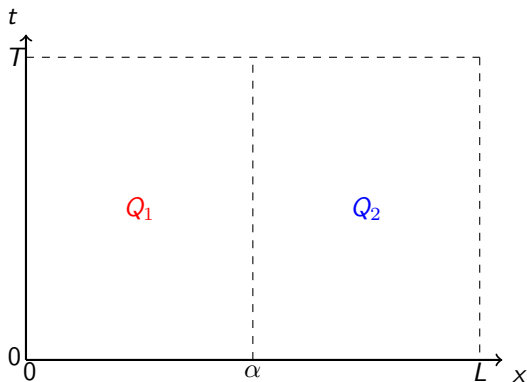
$$\partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$p_2^\ell(\alpha, t) = p_1^\ell(\alpha, t)$$

$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz waveform



In $Q_1 = (0, \alpha) \times (0, T)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$y_1^\ell(0, t) = 0$$

$$y_1^\ell(\alpha, t) = y_2^{\ell-1}(\alpha, t)$$

$$y_1^\ell(x, 0) = y_{1,0}(x)$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$p_1^\ell(0, t) = 0$$

$$p_1^\ell(\alpha, t) = p_2^{\ell-1}(\alpha, t)$$

$$p_1^\ell(x, T) = 0$$

In $Q_2 = (\alpha, L) \times (0, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell$$

$$y_2^\ell(\alpha, t) = y_1^\ell(\alpha, t)$$

$$y_2^\ell(L, t) = 0$$

$$y_2^\ell(x, 0) = y_{2,0}(x)$$

$$\partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

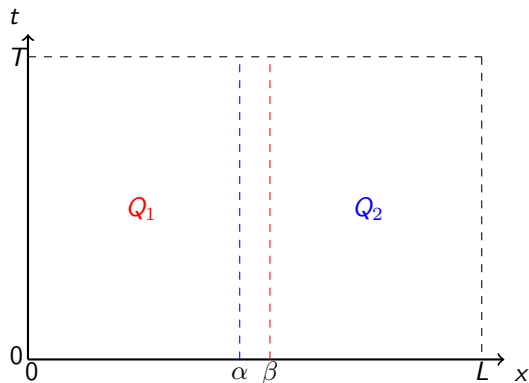
$$p_2^\ell(\alpha, t) = p_1^\ell(\alpha, t)$$

$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, T) = 0$$

Does not converge without overlap !

Alternating Schwarz waveform



Now converges with overlap !

In $Q_1 = (0, \beta) \times (0, T)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$y_1^\ell(0, t) = 0$$

$$y_1^\ell(\beta, t) = y_2^{\ell-1}(\beta, t)$$

$$y_1^\ell(x, 0) = y_{1,0}(x)$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$p_1^\ell(0, t) = 0$$

$$p_1^\ell(\beta, t) = p_2^{\ell-1}(\beta, t)$$

$$p_1^\ell(x, T) = 0$$

In $Q_2 = (\alpha, L) \times (0, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell$$

$$y_2^\ell(\alpha, t) = y_1^\ell(\alpha, t)$$

$$y_2^\ell(L, t) = 0$$

$$y_2^\ell(x, 0) = y_{2,0}(x)$$

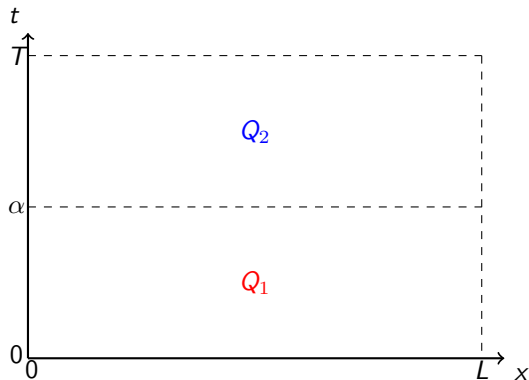
$$\partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$p_2^\ell(\alpha, t) = p_1^\ell(\alpha, t)$$

$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, T) = 0$$

Time decomposition



Subdomains: $Q_1 = (0, L) \times (0, \alpha)$ and
 $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y - \partial_{xx} y = \nu^{-1} p \quad \partial_t p + \partial_{xx} p = y - \hat{y}$$

$$y(0, t) = 0$$

$$p(0, t) = 0$$

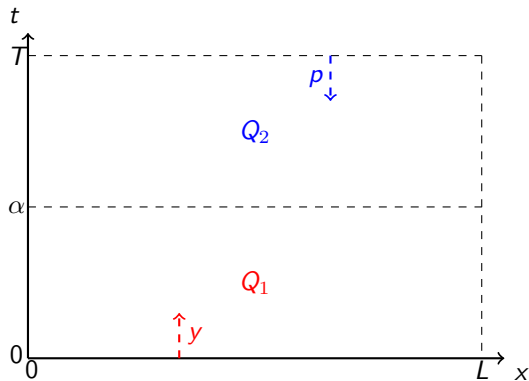
$$y(L, t) = 0$$

$$p(L, t) = 0$$

$$y(x, 0) = y_0(x)$$

$$p(x, T) = 0$$

Time decomposition



Subdomains: $Q_1 = (0, L) \times (0, \alpha)$ and $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y - \partial_{xx} y = \nu^{-1} p \quad \partial_t p + \partial_{xx} p = y - \hat{y}$$

$$y(0, t) = 0$$

$$p(0, t) = 0$$

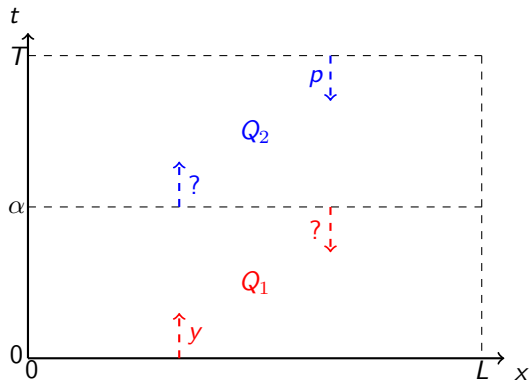
$$y(L, t) = 0$$

$$p(L, t) = 0$$

$$y(x, 0) = y_0(x)$$

$$p(x, T) = 0$$

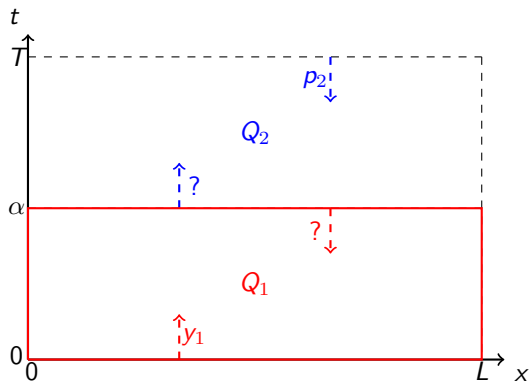
Time decomposition



Subdomains: $Q_1 = (0, L) \times (0, \alpha)$ and $Q_2 = (0, L) \times (\alpha, T)$

$$\begin{aligned} \partial_t y - \partial_{xx} y &= \nu^{-1} p & \partial_t p + \partial_{xx} p &= y - \hat{y} \\ y(0, t) &= 0 & p(0, t) &= 0 \\ y(L, t) &= 0 & p(L, t) &= 0 \\ y(x, 0) &= y_0(x) & p(x, T) &= 0 \end{aligned}$$

Alternating Schwarz in time



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

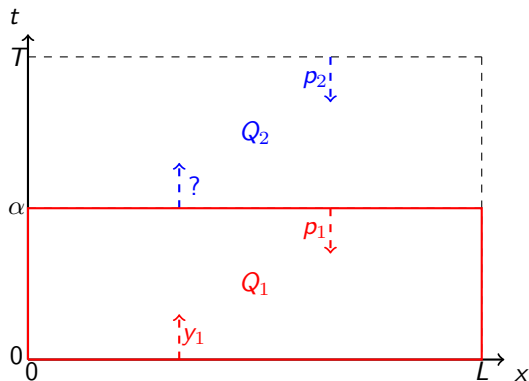
$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

Alternating Schwarz in time



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

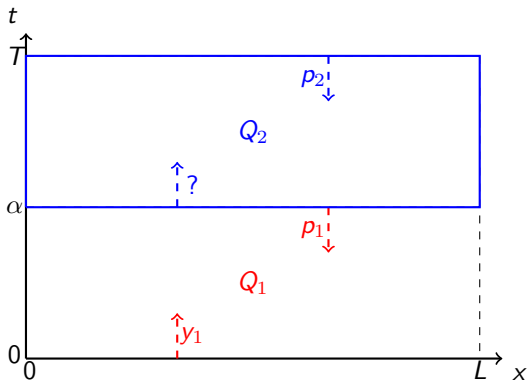
$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$p_1^\ell(x, \alpha) = p_2^{\ell-1}(x, \alpha)$$

Alternating Schwarz in time


$$\text{In } Q_1 = (0, L) \times (0, \alpha)$$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$p_1^\ell(x, \alpha) = p_2^{\ell-1}(x, \alpha)$$

$$\text{In } Q_2 = (0, L) \times (\alpha, T)$$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

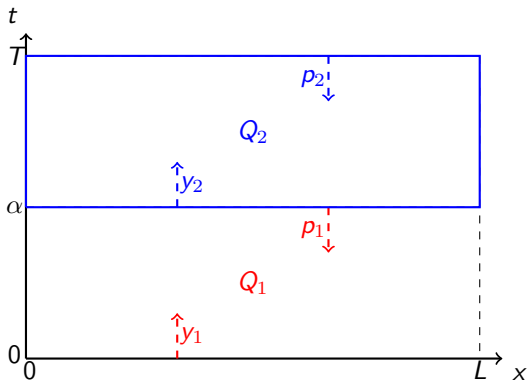
$$p_2^\ell(0, t) = 0$$

$$y_2^\ell(L, t) = 0$$

$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time


$$\text{In } Q_1 = (0, L) \times (0, \alpha)$$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$p_1^\ell(x, \alpha) = p_2^{\ell-1}(x, \alpha)$$

$$\text{In } Q_2 = (0, L) \times (\alpha, T)$$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

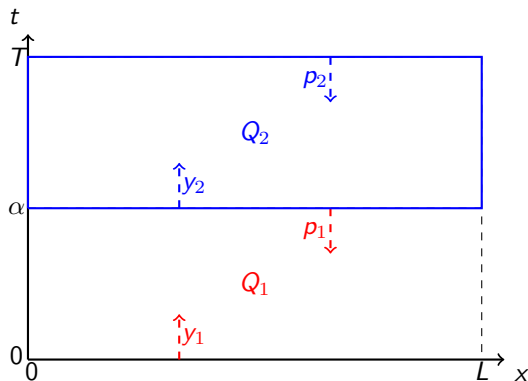
$$y_2^\ell(L, t) = 0$$

$$p_2^\ell(L, t) = 0$$

$$y_2^\ell(x, \alpha) = y_1^\ell(x, \alpha)$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time



How about the convergence ?

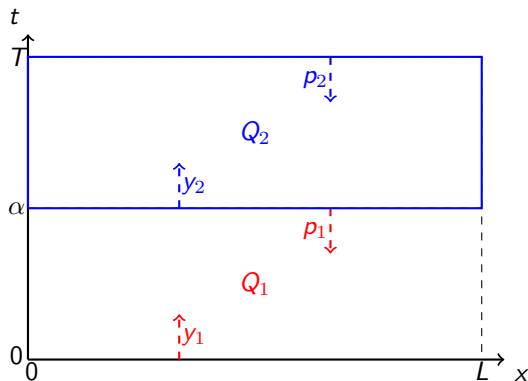
In $Q_1 = (0, L) \times (0, \alpha)$

$$\begin{aligned} \partial_t y_1^\ell - \partial_{xx} y_1^\ell &= \nu^{-1} p_1^\ell & \partial_t p_1^\ell + \partial_{xx} p_1^\ell &= y_1^\ell - \hat{y}_1 \\ y_1^\ell(0, t) &= 0 & p_1^\ell(0, t) &= 0 \\ y_1^\ell(L, t) &= 0 & p_1^\ell(L, t) &= 0 \\ y_1^\ell(x, 0) &= y_0(x) & p_1^\ell(x, \alpha) &= p_2^{\ell-1}(x, \alpha) \end{aligned}$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\begin{aligned} \partial_t y_2^\ell - \partial_{xx} y_2^\ell &= \nu^{-1} p_2^\ell & \partial_t p_2^\ell + \partial_{xx} p_2^\ell &= y_2^\ell - \hat{y}_2 \\ y_2^\ell(0, t) &= 0 & p_2^\ell(0, t) &= 0 \\ y_2^\ell(L, t) &= 0 & p_2^\ell(L, t) &= 0 \\ y_2^\ell(x, \alpha) &= y_1^\ell(x, \alpha) & p_2^\ell(x, T) &= 0 \end{aligned}$$

Alternating Schwarz in time



How about the convergence ?
It **does converge** without overlap !

In $Q_1 = (0, L) \times (0, \alpha)$

$$\begin{aligned} \partial_t y_1^\ell - \partial_{xx} y_1^\ell &= \nu^{-1} p_1^\ell & \partial_t p_1^\ell + \partial_{xx} p_1^\ell &= y_1^\ell - \hat{y}_1 \\ y_1^\ell(0, t) &= 0 & p_1^\ell(0, t) &= 0 \\ y_1^\ell(L, t) &= 0 & p_1^\ell(L, t) &= 0 \\ y_1^\ell(x, 0) &= y_0(x) & p_1^\ell(x, \alpha) &= p_2^{\ell-1}(x, \alpha) \end{aligned}$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\begin{aligned} \partial_t y_2^\ell - \partial_{xx} y_2^\ell &= \nu^{-1} p_2^\ell & \partial_t p_2^\ell + \partial_{xx} p_2^\ell &= y_2^\ell - \hat{y}_2 \\ y_2^\ell(0, t) &= 0 & p_2^\ell(0, t) &= 0 \\ y_2^\ell(L, t) &= 0 & p_2^\ell(L, t) &= 0 \\ y_2^\ell(x, \alpha) &= y_1^\ell(x, \alpha) & p_2^\ell(x, T) &= 0 \end{aligned}$$

Alternating Schwarz in time - Intuition

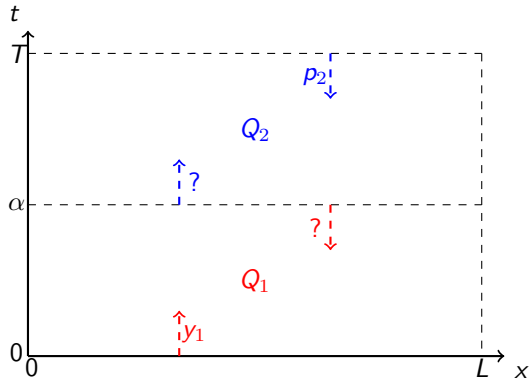
Relations

$$p_j^\ell = \nu(\partial_t y_j^\ell - \partial_{xx} y_j^\ell)$$

$$y_j^\ell = \partial_t p_j^\ell + \partial_{xx} p_j^\ell + \hat{y}_j$$

In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$



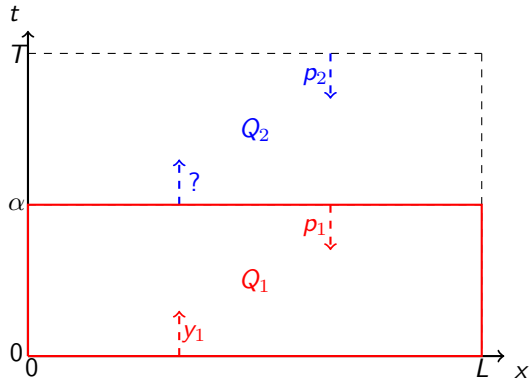
In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

Alternating Schwarz in time - Intuition

Relations

$$p_j^\ell = \nu(\partial_t y_j^\ell - \partial_{xx} y_j^\ell)$$



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$p_1^\ell(x, \alpha) = p_2^{\ell-1}(x, \alpha)$$

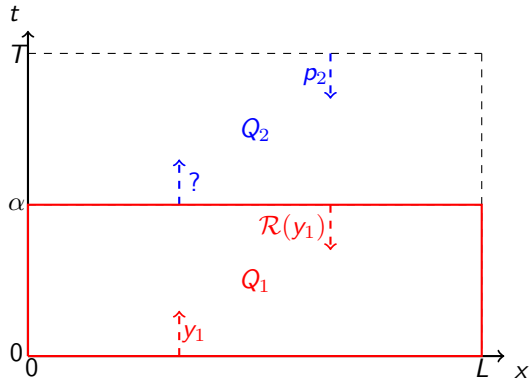
In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

Alternating Schwarz in time - Intuition

Relations

$$p_j^\ell = \nu(\partial_t y_j^\ell - \partial_{xx} y_j^\ell)$$



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$(\partial_t - \partial_{xx}) y_1^\ell(x, \alpha) = (\partial_t - \partial_{xx}) y_2^{\ell-1}(x, \alpha)$$

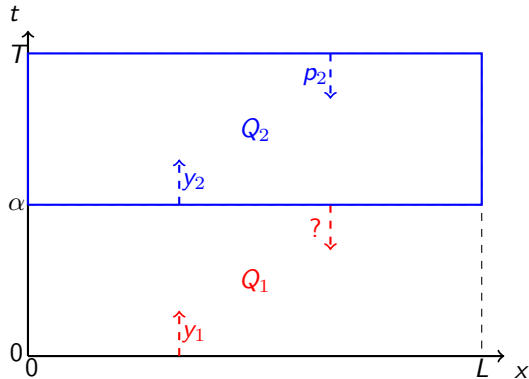
In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

Alternating Schwarz in time - Intuition

Relations

$$y_j^\ell = \partial_t p_j^\ell + \partial_{xx} p_j^\ell + \hat{y}_j$$



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

$$y_2^\ell(L, t) = 0$$

$$p_2^\ell(L, t) = 0$$

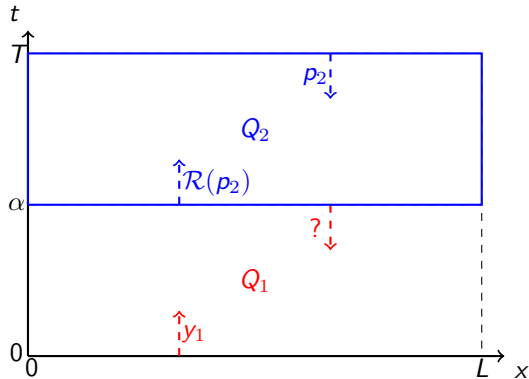
$$y_2^\ell(x, \alpha) = y_1^\ell(x, \alpha)$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time - Intuition

Relations

$$y_j^\ell = \partial_t p_j^\ell + \partial_{xx} p_j^\ell + \hat{y}_j$$



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

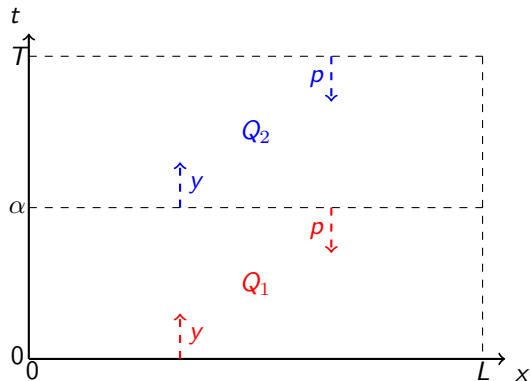
$$y_2^\ell(L, t) = 0$$

$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, T) = 0$$

$$(\partial_t + \partial_{xx}) p_2^\ell(x, \alpha) = (\partial_t + \partial_{xx}) p_1^\ell(x, \alpha)$$

Alternating Schwarz in time - Observation



- **Dirichlet** condition transform to **Robin** type condition !

In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$p_1^\ell(x, \alpha) = p_2^{\ell-1}(x, \alpha)$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

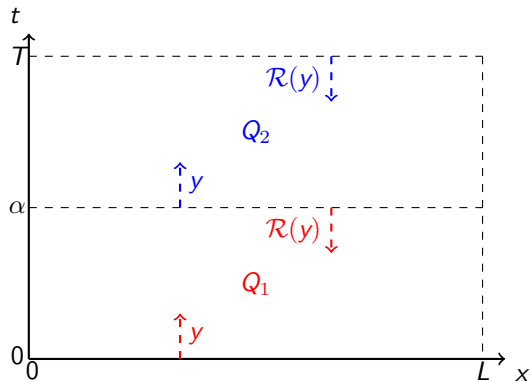
$$y_2^\ell(L, t) = 0$$

$$p_2^\ell(L, t) = 0$$

$$y_2^\ell(x, \alpha) = y_1^\ell(x, \alpha)$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time - Observation



- **Dirichlet** condition transform to **Robin** type condition !
- Forward-backward **might be** less important ?

$$\text{In } Q_1 = (0, L) \times (0, \alpha)$$

$$-\partial_{tt}y_1^\ell + \Delta^2 y_1^\ell + \nu^{-1}y_1^\ell = \nu^{-1}\hat{y}_1$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$(\partial_t - \partial_{xx})y_1^\ell(x, \alpha) = (\partial_t - \partial_{xx})y_2^{\ell-1}(x, \alpha) + \text{B.C.}$$

$$\text{In } Q_2 = (0, L) \times (\alpha, T)$$

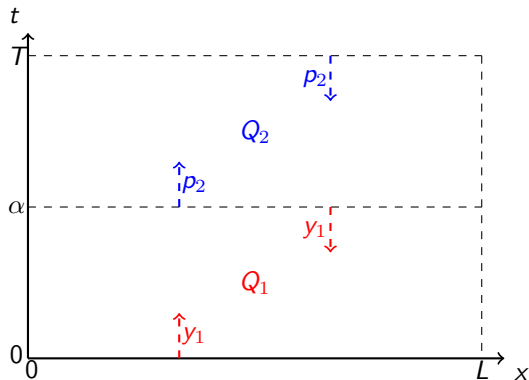
$$-\partial_{tt}y_2^\ell + \Delta^2 y_2^\ell + \nu^{-1}y_2^\ell = \nu^{-1}\hat{y}_2$$

$$y_2^\ell(x, \alpha) = y_1^\ell(x, \alpha)$$

$$(\partial_t - \partial_{xx})y_2^{\ell-1}(x, T) = 0 + \text{B.C.}$$

Alternating Schwarz in time - Variants

Exchange p and y



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$y_1^\ell(x, \alpha) = y_2^{\ell-1}(x, \alpha)$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

$$y_2^\ell(L, t) = 0$$

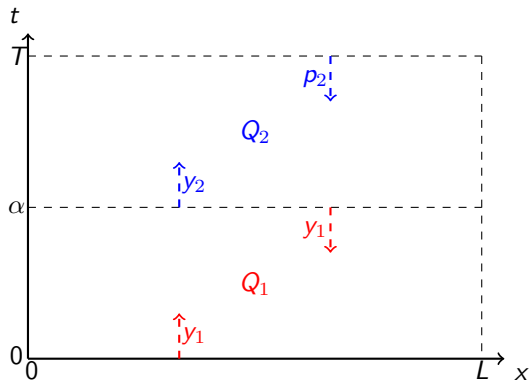
$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, \alpha) = p_1^\ell(x, \alpha)$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time - Variants

Only with y



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$y_1^\ell(x, \alpha) = y_2^{\ell-1}(x, \alpha)$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

$$y_2^\ell(L, t) = 0$$

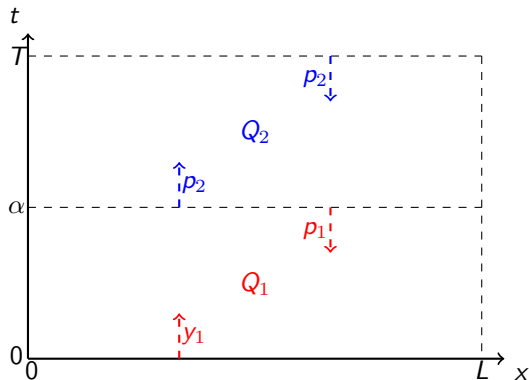
$$p_2^\ell(L, t) = 0$$

$$y_2^\ell(x, \alpha) = y_1^\ell(x, \alpha)$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time - Variants

Only with p



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$p_1^\ell(x, \alpha) = p_2^{\ell-1}(x, \alpha)$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

$$y_2^\ell(L, t) = 0$$

$$p_2^\ell(L, t) = 0$$

$$p_2^\ell(x, \alpha) = p_1^\ell(x, \alpha)$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time - Convergence

Dirichlet type transmission condition:

name	SD_1	SD_2	SD_3	SD_4
Q_1	p	y	y	p
Q_2	y	p	y	p

Alternating Schwarz in time - Convergence

Dirichlet type transmission condition:

name	SD ₁	SD ₂	SD ₃	SD ₄
Q_1	p	y	y	p
Q_2	y	p	y	p

Analysis: eigendecomposition in space

$$\rho_{SD_1} = \max_{\lambda_m \in \Lambda} \frac{1}{\nu(\sigma_m \coth(a_m) + \lambda_m)(\sigma_m \coth(b_m) + \lambda_m)}$$

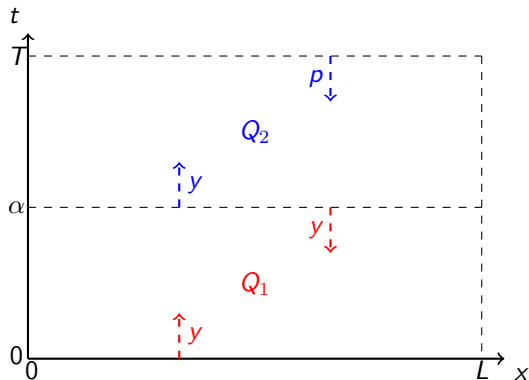
$$\rho_{SD_2} = \max_{\lambda_m \in \Lambda} \nu(\sigma_m \coth(a_m) + \lambda_m)(\sigma_m \coth(b_m) + \lambda_m)$$

$$\rho_{SD_3} = 1$$

$$\rho_{SD_4} = 1$$

Alternating Schwarz in time - Convergence

Only with y



In $Q_1 = (0, L) \times (0, \alpha)$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell \quad \partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(0, t) = 0$$

$$p_1^\ell(0, t) = 0$$

$$y_1^\ell(L, t) = 0$$

$$p_1^\ell(L, t) = 0$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$y_1^\ell(x, \alpha) = y_2^{\ell-1}(x, \alpha)$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$\partial_t y_2^\ell - \partial_{xx} y_2^\ell = \nu^{-1} p_2^\ell \quad \partial_t p_2^\ell + \partial_{xx} p_2^\ell = y_2^\ell - \hat{y}_2$$

$$y_2^\ell(0, t) = 0$$

$$p_2^\ell(0, t) = 0$$

$$y_2^\ell(L, t) = 0$$

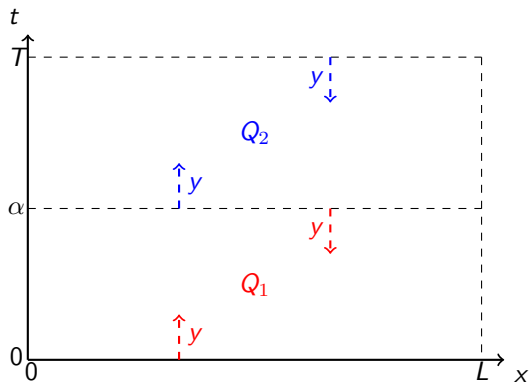
$$p_2^\ell(L, t) = 0$$

$$y_2^\ell(x, \alpha) = y_1^\ell(x, \alpha)$$

$$p_2^\ell(x, T) = 0$$

Alternating Schwarz in time - Convergence

Only with y



In $Q_1 = (0, L) \times (0, \alpha)$

$$-\partial_{tt}y_1^\ell + \Delta^2 y_1^\ell + \nu^{-1}y_1^\ell = \nu^{-1}\hat{y}_1$$

$$y_1^\ell(x, 0) = y_0(x)$$

$$y_1^\ell(x, \alpha) = y_2^{\ell-1}(x, \alpha) + \text{B.C.}$$

In $Q_2 = (0, L) \times (\alpha, T)$

$$-\partial_{tt}y_2^\ell + \Delta^2 y_2^\ell + \nu^{-1}y_2^\ell = \nu^{-1}\hat{y}_2$$

$$y_2^\ell(x, \alpha) = y_1^\ell(x, \alpha)$$

$$(\partial_t - \partial_{xx})y_2^{\ell-1}(x, T) = 0 + \text{B.C.}$$

Alternating Schwarz in time - Convergence

Dirichlet type transmission condition:

name	SD ₁	SD ₂	SD ₃	SD ₄
Q_1	p	y	y	p
Q_2	y	p	y	p

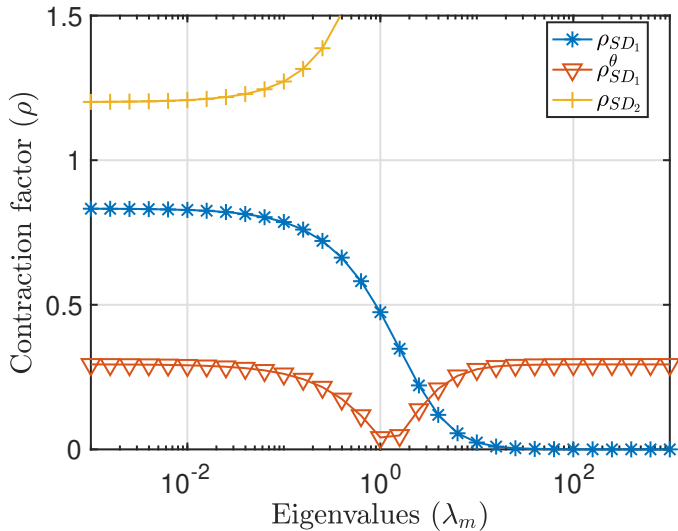
Analysis: eigendecomposition in space

$$\begin{aligned}\rho_{SD_1} &= \max_{\lambda_m \in \Lambda} \frac{1}{\nu(\sigma_m \coth(a_m) + \lambda_m)(\sigma_m \coth(b_m) + \lambda_m)} \\ \rho_{SD_2} &= \max_{\lambda_m \in \Lambda} \nu(\sigma_m \coth(a_m) + \lambda_m)(\sigma_m \coth(b_m) + \lambda_m) \\ \rho_{SD_3} &= 1 \\ \rho_{SD_4} &= 1\end{aligned}$$

Relaxation: e.g., $p_1^\ell(x, \alpha) = p_2^{\ell-1}(x, \alpha)$ to $p_1^\ell(x, \alpha) = f^\ell(x)$ with

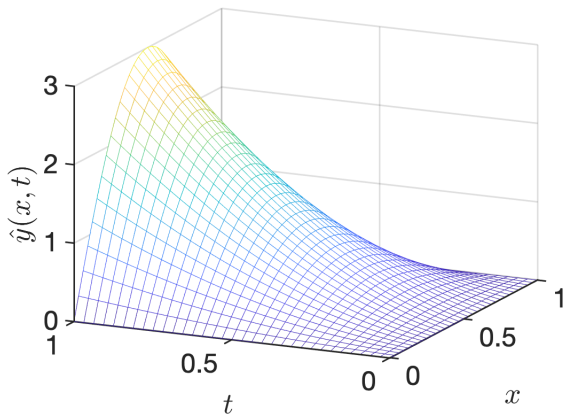
$$f^{\ell+1}(x) = (1 - \theta)f^\ell(x) + \theta p_2^\ell(x, \alpha) \quad \theta \in (0, 1)$$

Alternating Schwarz in time - Contraction



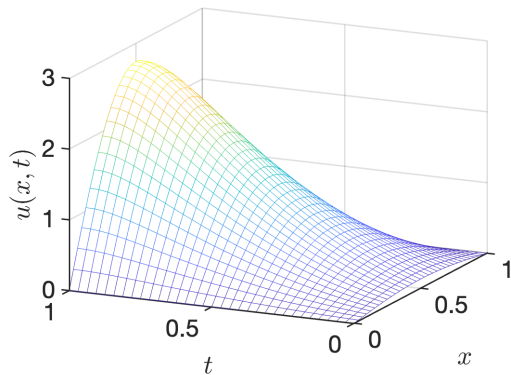
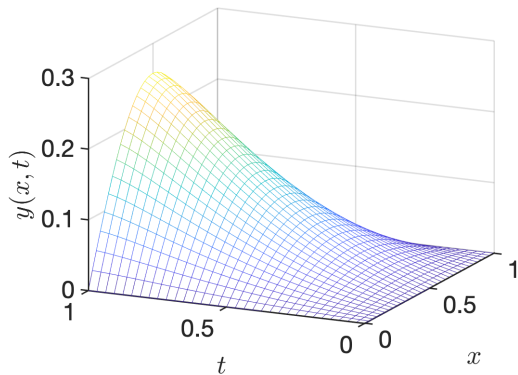
Numerical experiments

Target function: $\hat{y}(x, t) = \sin(\pi x)(2t^2 + t)$



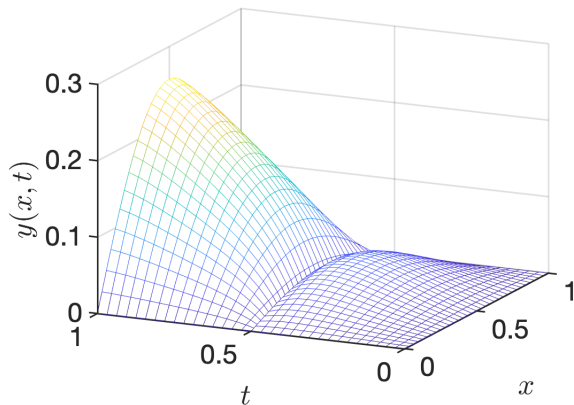
Numerical experiments

Numerical solutions: Crank-Nicolson with mesh size $h_t = h_x = \frac{1}{32}$ and penalization parameter: $\nu = 0.1$

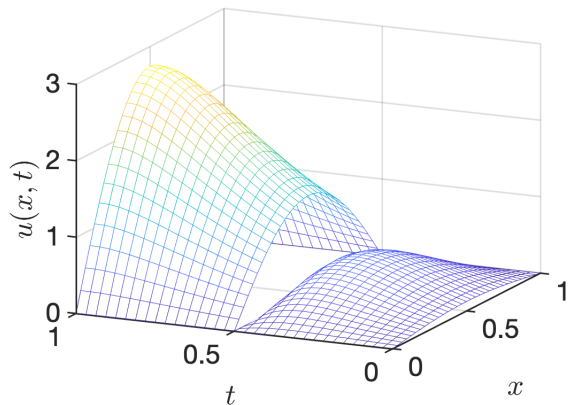


Numerical experiments

Iteration = 1

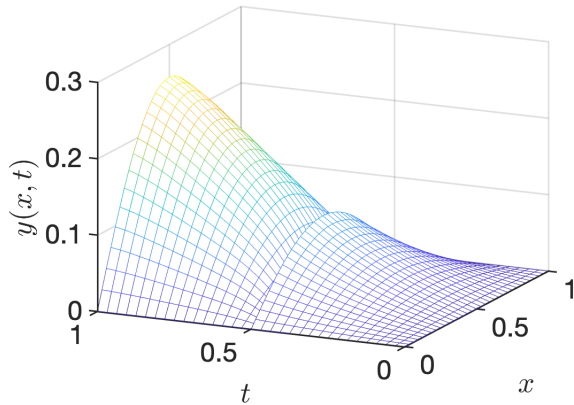


Iteration = 1

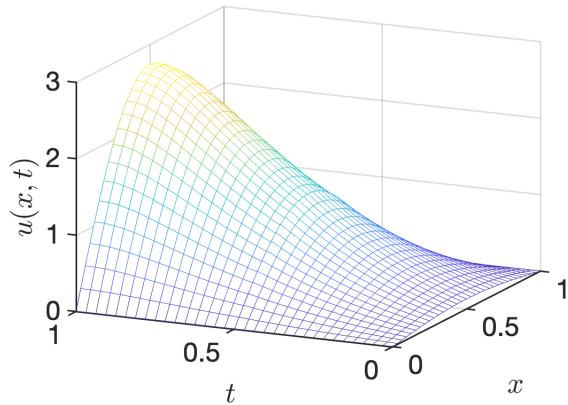


Numerical experiments

Iteration = 2

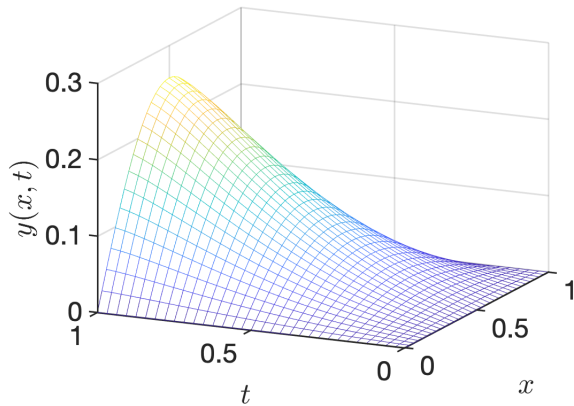


Iteration = 2

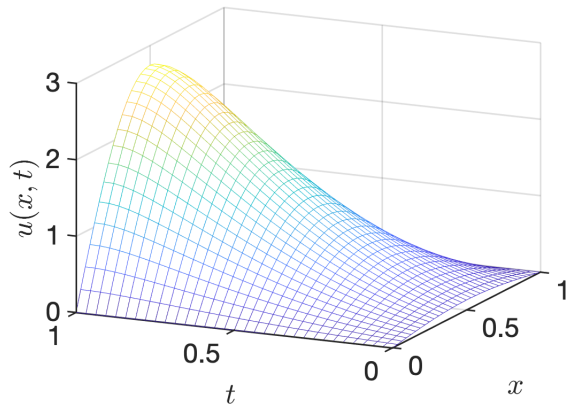


Numerical experiments

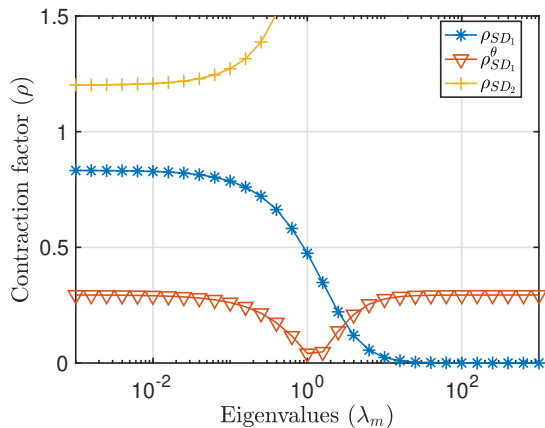
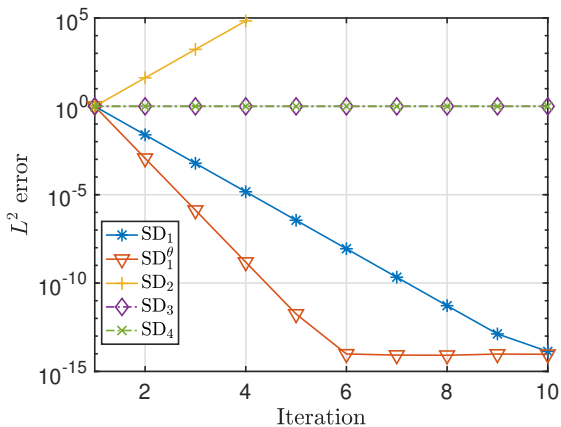
Iteration = 5



Iteration = 5



Numerical experiments - Convergence behavior



How about Neumann condition ?

Neumann type transmission condition:

name	SN ₁	SN ₂	SN ₃	SN ₄
Q_1	$\partial_t p$	$\partial_t y$	$\partial_t y$	$\partial_t p$
Q_2	$\partial_t y$	$\partial_t p$	$\partial_t y$	$\partial_t p$

How about Neumann condition ?

Neumann type transmission condition:

name	SN ₁	SN ₂	SN ₃	SN ₄
Q_1	$\partial_t p$	$\partial_t y$	$\partial_t y$	$\partial_t p$
Q_2	$\partial_t y$	$\partial_t p$	$\partial_t y$	$\partial_t p$

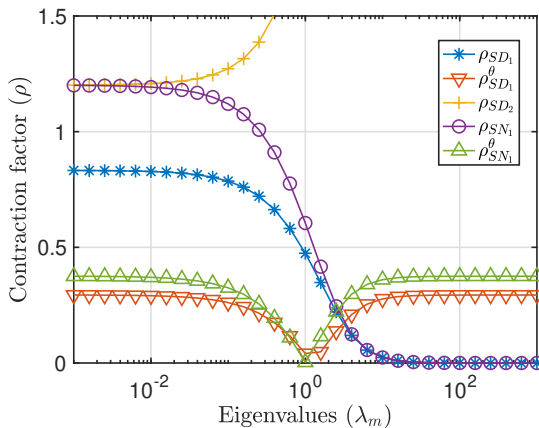
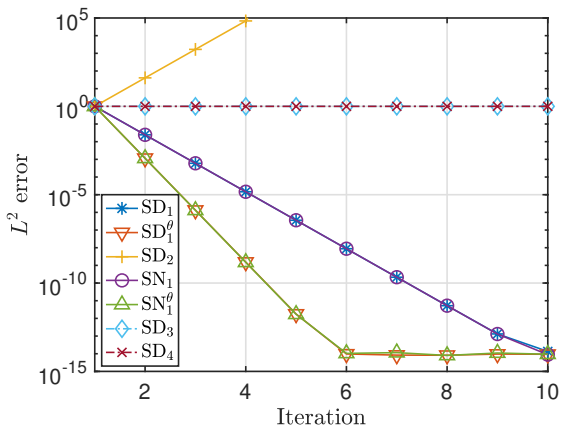
Analysis: eigendecomposition in space:

$$\begin{aligned}\rho_{\text{SN}_1} &= \max_{\lambda_m \in \Lambda} \frac{1}{\nu(\sigma_m \tanh(a_m) + \lambda_m)(\sigma_i \tanh(b_m) + \lambda_m)} \\ \rho_{\text{SN}_2} &= \max_{\lambda_m \in \Lambda} \nu(\sigma_m \tanh(a_m) + \lambda_m)(\sigma_m \tanh(b_m) + \lambda_m) \\ \rho_{\text{SN}_3} &= 1 \\ \rho_{\text{SN}_4} &= 1\end{aligned}$$

Relaxation: e.g., $\partial_t p_1^\ell(x, \alpha) = \partial_t p_2^{\ell-1}(x, \alpha)$ to $\partial_t p_1^\ell(x, \alpha) = f^\ell(x)$ with

$$f^{\ell+1}(x) = (1 - \theta)f^\ell(x) + \theta \partial_t p_2^\ell(x, \alpha) \quad \theta \in (0, 1)$$

How about Neumann condition ?



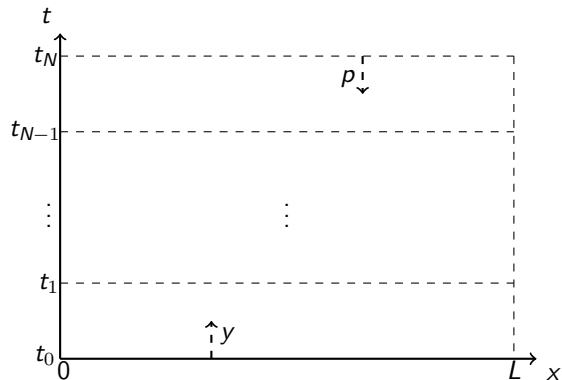
What we learned

Three main messages:

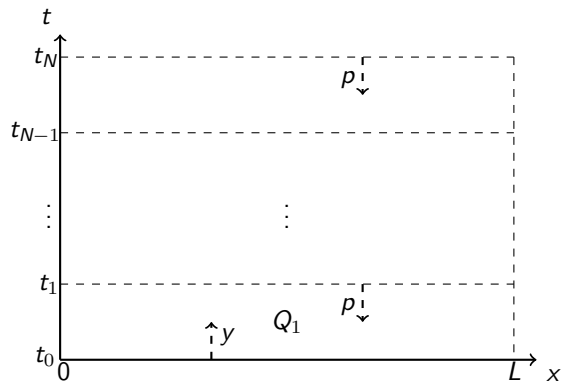
- Time decomposition is **different** from space decomposition
- Forward-backward structure is **important** for alternating Schwarz to converge without overlap
- Relaxation **can improve** the convergence, but is **not necessary** for the method to converge

Many other variants: 3 DN, 3 ND, 9 NN, 36 OS, ...

Many subdomains and scalability



Many subdomains and scalability



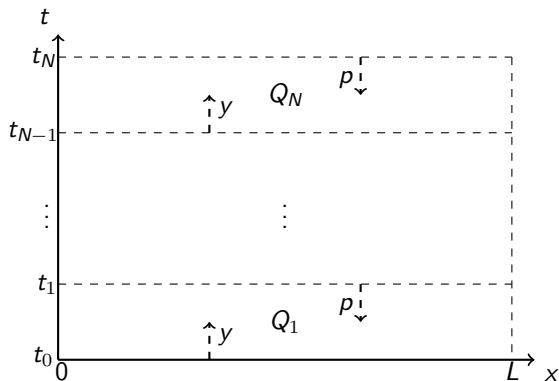
$$\text{In } Q_1 = (0, L) \times (t_0, t_1)$$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(x, t_0) = y_0(x), \quad p_1^\ell(x, t_1) = p_2^{\ell-1}(x, t_1)$$

Many subdomains and scalability



$$\text{In } Q_1 = (0, L) \times (t_0, t_1)$$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(x, t_0) = y_0(x), \quad p_1^\ell(x, t_1) = p_2^{\ell-1}(x, t_1)$$

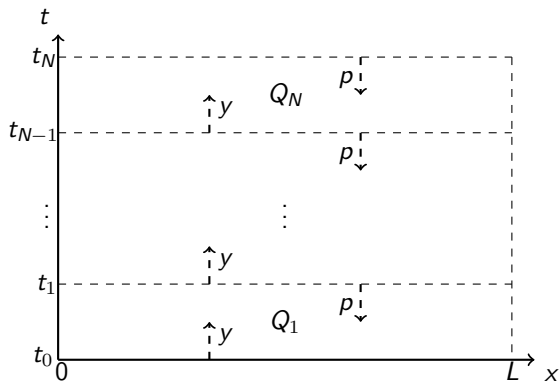
$$\text{In } Q_N = (0, L) \times (t_{N-1}, T)$$

$$\partial_t y_N^\ell - \partial_{xx} y_N^\ell = \nu^{-1} p_N^\ell$$

$$\partial_t p_N^\ell + \partial_{xx} p_N^\ell = y_N^\ell - \hat{y}_N$$

$$y_N^\ell(x, t_{N-1}) = y_N^{\ell-1}(x, t_{N-1}), \quad p_N^\ell(x, t_N) = 0$$

Many subdomains and scalability



$$\text{In } Q_1 = (0, L) \times (t_0, t_1)$$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(x, t_0) = y_0(x), \quad p_1^\ell(x, t_1) = p_2^{\ell-1}(x, t_1)$$

$$\text{In } Q_n = (0, L) \times (t_{n-1}, t_n)$$

$$\partial_t y_n^\ell - \partial_{xx} y_n^\ell = \nu^{-1} p_n^\ell$$

$$\partial_t p_n^\ell + \partial_{xx} p_n^\ell = y_n^\ell - \hat{y}_n$$

$$y_n^\ell(x, t_{n-1}) = y_n^{\ell-1}(x, t_{n-1}), \quad p_n^\ell(x, t_n) = p_{n+1}^{\ell-1}(x, t_n)$$

$$\text{In } Q_N = (0, L) \times (t_{N-1}, T)$$

$$\partial_t y_N^\ell - \partial_{xx} y_N^\ell = \nu^{-1} p_N^\ell$$

$$\partial_t p_N^\ell + \partial_{xx} p_N^\ell = y_N^\ell - \hat{y}_N$$

$$y_N^\ell(x, t_{N-1}) = y_N^{\ell-1}(x, t_{N-1}), \quad p_N^\ell(x, t_N) = 0$$

A comment on Decompose-then-optimize

Decompose first optimization
problem

$$\min_{y_n^\ell, u_n^\ell} \frac{1}{2} \|y_n^\ell - \hat{y}_n\|_{L^2(Q_n)}^2 + \frac{\nu}{2} \|u_n^\ell\|_{L^2(Q_n)}^2 \\ + y_n^\ell(t_n)^\top p_{n+1}^{\ell-1}(t_n)$$

s.t.

$$\begin{aligned} \partial_t y_n^\ell - \Delta y_n^\ell &= u_n^\ell && \text{in } Q_n \\ y_n^\ell &= g && \text{on } \Sigma \\ y_n^\ell &= y_{n-1}^{\ell-1} && \text{on } \Sigma_{n-1} \end{aligned}$$

with an extra term $y_n^\ell(t_n)^\top p_{n+1}^{\ell-1}(t_n)$

A comment on Decompose-then-optimize

Decompose first optimization problem

$$\min_{y_n^\ell, u_n^\ell} \frac{1}{2} \|y_n^\ell - \hat{y}_n\|_{L^2(Q_n)}^2 + \frac{\nu}{2} \|u_n^\ell\|_{L^2(Q_n)}^2$$

$$+ y_n^\ell(t_n)^\top p_{n+1}^{\ell-1}(t_n)$$

Then optimize

s.t. $\implies$

$$\partial_t y_n^\ell - \Delta y_n^\ell = u_n^\ell \quad \text{in } Q_n$$

$$y_n^\ell = g \quad \text{on } \Sigma$$

$$y_n^\ell = y_{n-1}^{\ell-1} \quad \text{on } \Sigma_{n-1}$$

with an extra term $y_n^\ell(t_n)^\top p_{n+1}^{\ell-1}(t_n)$

$$\text{In } Q_1 = (0, L) \times (t_0, t_1)$$

$$\partial_t y_1^\ell - \partial_{xx} y_1^\ell = \nu^{-1} p_1^\ell$$

$$\partial_t p_1^\ell + \partial_{xx} p_1^\ell = y_1^\ell - \hat{y}_1$$

$$y_1^\ell(x, t_0) = y_0(x), \quad p_1^\ell(x, t_1) = p_2^{\ell-1}(x, t_1)$$

$$\text{In } Q_n = (0, L) \times (t_{n-1}, t_n)$$

$$\partial_t y_n^\ell - \partial_{xx} y_n^\ell = \nu^{-1} p_n^\ell$$

$$\partial_t p_n^\ell + \partial_{xx} p_n^\ell = y_n^\ell - \hat{y}_n$$

$$y_n^\ell(x, t_{n-1}) = y_{n-1}^{\ell-1}(x, t_{n-1}), \quad p_n^\ell(x, t_n) = p_{n+1}^{\ell-1}(x, t_n)$$

$$\text{In } Q_N = (0, L) \times (t_{N-1}, T)$$

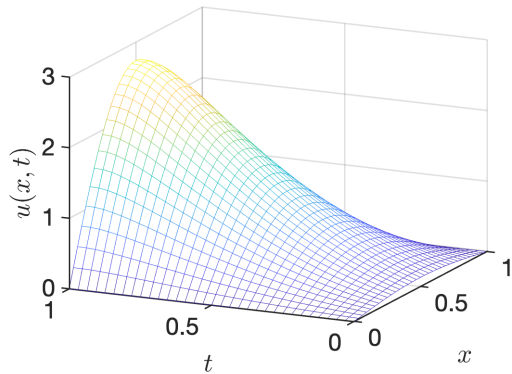
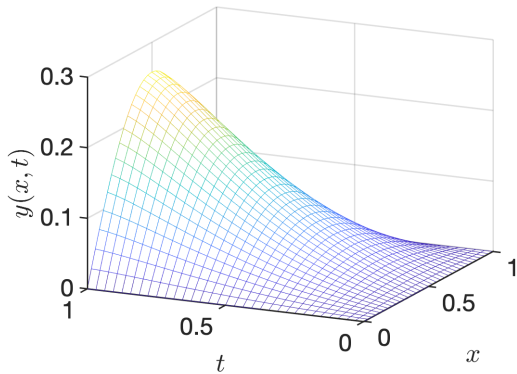
$$\partial_t y_N^\ell - \partial_{xx} y_N^\ell = \nu^{-1} p_N^\ell$$

$$\partial_t p_N^\ell + \partial_{xx} p_N^\ell = y_N^\ell - \hat{y}_N$$

$$y_N^\ell(x, t_{N-1}) = y_{N-1}^{\ell-1}(x, t_{N-1}), \quad p_N^\ell(x, t_N) = 0$$

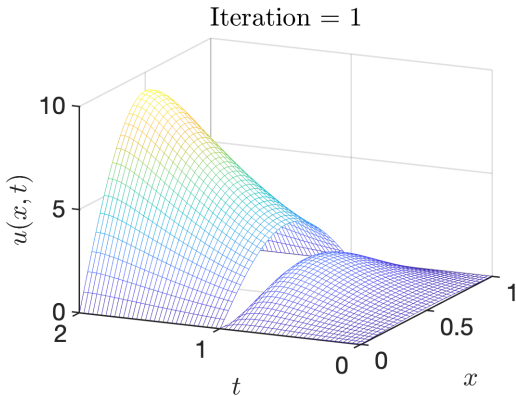
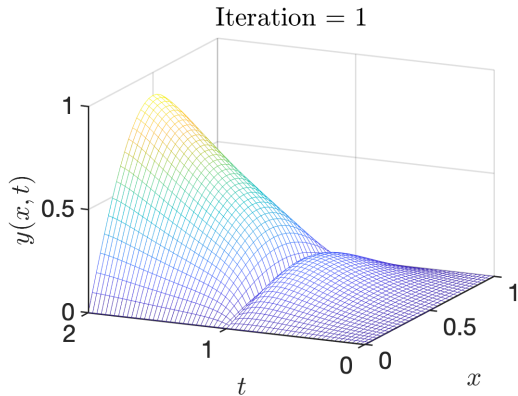
Weak scalability test

Numerical solutions: Same target $\hat{y}(x, t) = \sin(\pi x)(2t^2 + t)$ and same parameter settings



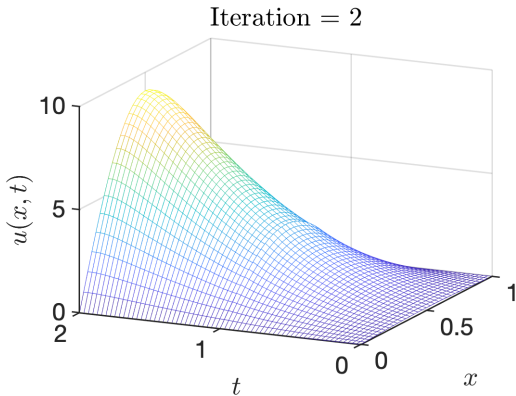
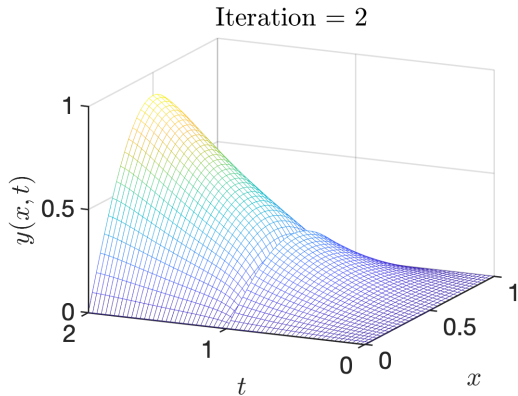
Weak scalability test

Two subdomains:



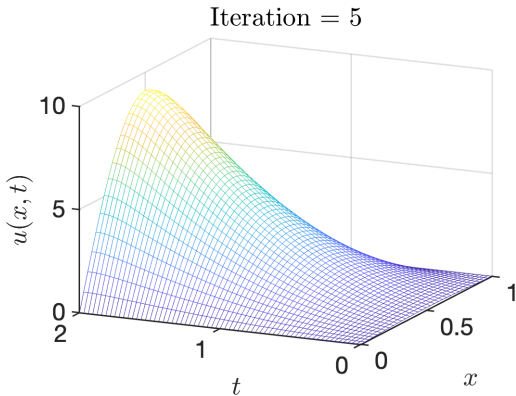
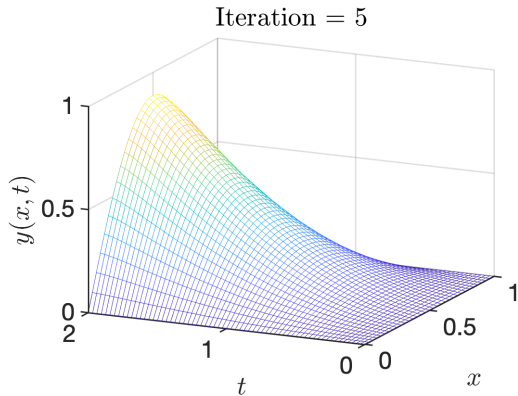
Weak scalability test

Two subdomains:



Weak scalability test

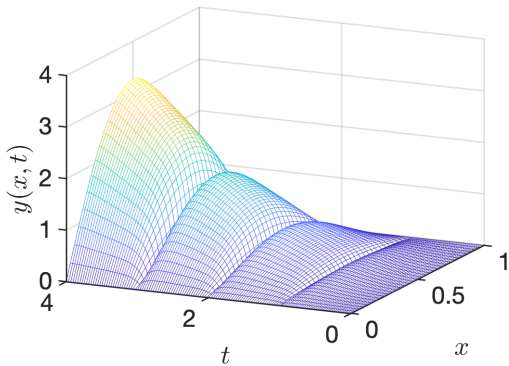
Two subdomains:



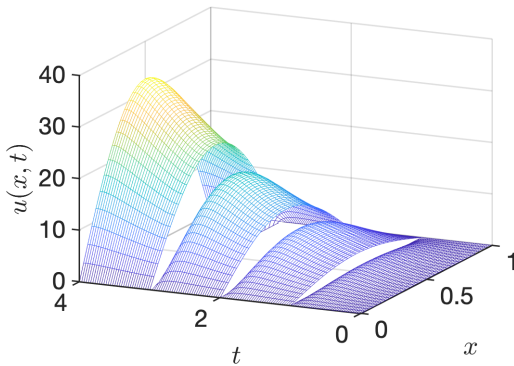
Weak scalability test

Four subdomains:

Iteration = 1

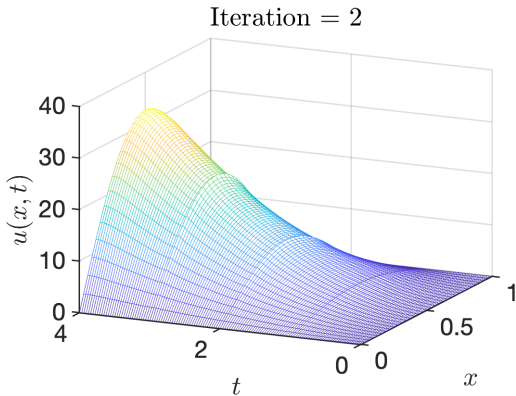
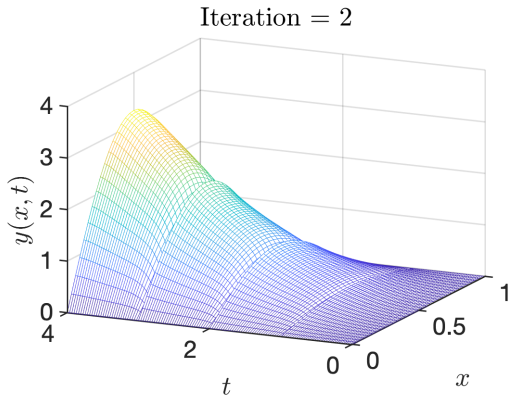


Iteration = 1



Weak scalability test

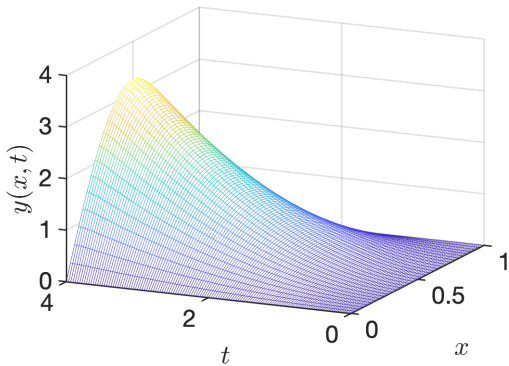
Four subdomains:



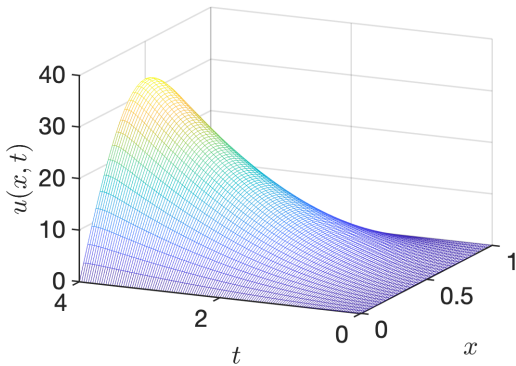
Weak scalability test

Four subdomains:

Iteration = 5

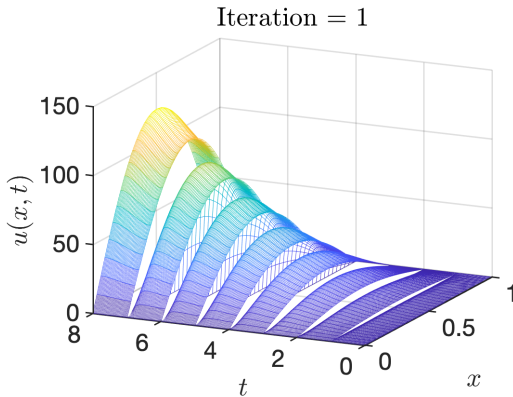
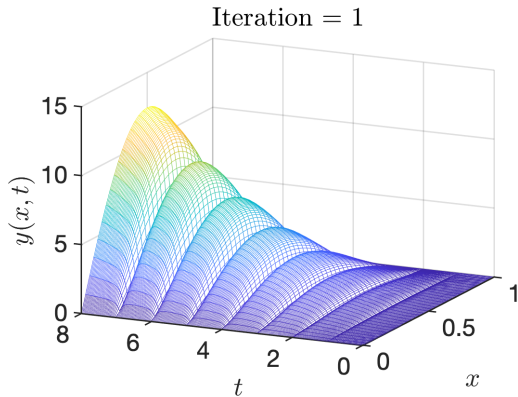


Iteration = 5



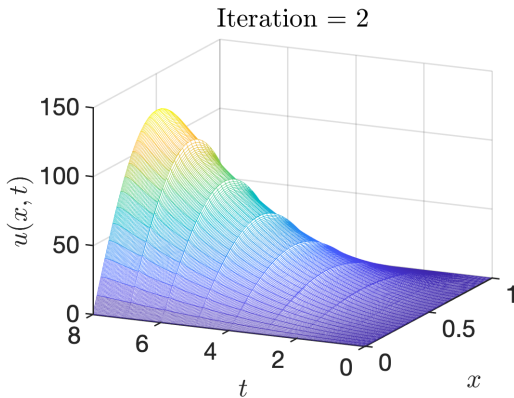
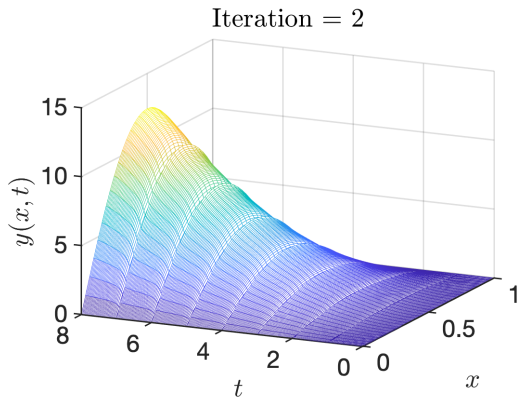
Weak scalability test

Eight subdomains:



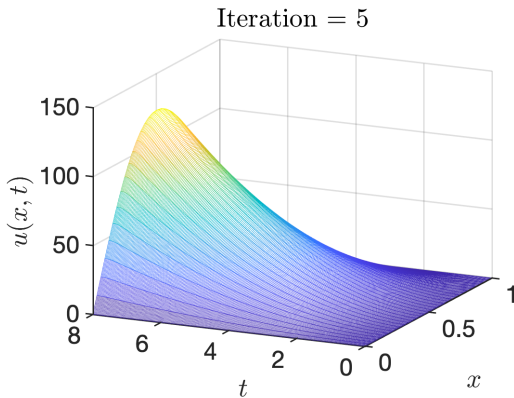
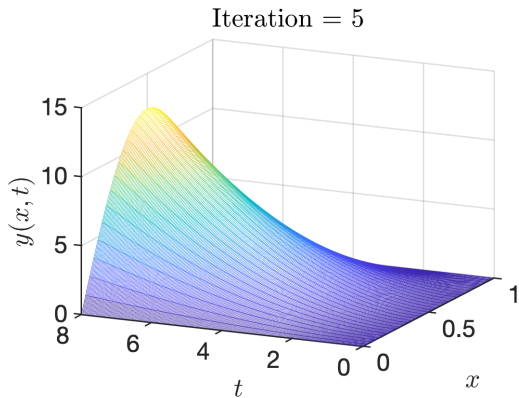
Weak scalability test

Eight subdomains:

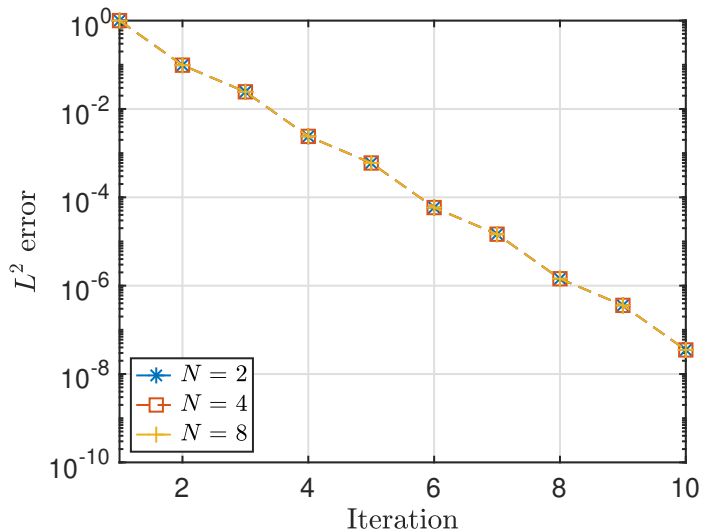


Weak scalability test

Eight subdomains:



Weak scalability test



Scalability analysis

Iteration matrix: eigendecomposition in space

$$T_{N,m} := \begin{bmatrix} T_{d,m} & T_{r,m} & & & \\ T_{l,m} & T_{d,m} & T_{r,m} & & \\ & \ddots & \ddots & \ddots & \\ & & T_{l,m} & T_{d,m} & T_{r,m} \\ & & & T_{l,m} & T_{d,m} \end{bmatrix}$$

block Toeplitz matrix with 2×2 block

$$T_{l,m} := \begin{bmatrix} 0 & 0 \\ 0 & C_{2,m}(\Delta t) \end{bmatrix}, \quad T_{r,m} := \begin{bmatrix} C_{2,m}(\Delta t) & 0 \\ 0 & 0 \end{bmatrix}$$

$$T_{d,m} := \begin{bmatrix} 0 & C_{1,m}(\Delta t) \\ -\nu C_{1,m}(\Delta t) & 0 \end{bmatrix}$$

$$C_{1,m}(\Delta t) := \frac{-\nu^{-1} \sinh(\sigma_m \Delta t)}{\sigma_m \cosh(\sigma_m \Delta t) + \lambda_m \sinh(\sigma_m \Delta t)}$$

$$C_{2,m}(\Delta t) := \frac{\sigma_m}{\sigma_m \cosh(\sigma_m \Delta t) + \lambda_m \sinh(\sigma_m \Delta t)}$$

Scalability analysis

Iteration matrix: eigendecomposition in space

$$T_{N,m} := \begin{bmatrix} T_{d,m} & T_{r,m} & & & \\ T_{l,m} & T_{d,m} & T_{r,m} & & \\ & \ddots & \ddots & \ddots & \\ & & T_{l,m} & T_{d,m} & T_{r,m} \\ & & & T_{l,m} & T_{d,m} \end{bmatrix}$$

block Toeplitz matrix with 2×2 block

Spectral radius of $T_{N,m}$:

$$\rho(T_{N,m}) \leq \|T_{N,m}\|$$

Goal: find a **Cst** such that $\|T_{N,m}\| \leq \mathbf{Cst} < 1$, and independent of N

$$T_{l,m} := \begin{bmatrix} 0 & 0 \\ 0 & C_{2,m}(\Delta t) \end{bmatrix}, \quad T_{r,m} := \begin{bmatrix} C_{2,m}(\Delta t) & 0 \\ 0 & 0 \end{bmatrix}$$

$$T_{d,m} := \begin{bmatrix} 0 & C_{1,m}(\Delta t) \\ -\nu C_{1,m}(\Delta t) & 0 \end{bmatrix}$$

$$C_{1,m}(\Delta t) := \frac{-\nu^{-1} \sinh(\sigma_m \Delta t)}{\sigma_m \cosh(\sigma_m \Delta t) + \lambda_m \sinh(\sigma_m \Delta t)}$$

$$C_{2,m}(\Delta t) := \frac{\sigma_m}{\sigma_m \cosh(\sigma_m \Delta t) + \lambda_m \sinh(\sigma_m \Delta t)}$$

First attempt: classical matrix norm, e.g.

$$\|T_{N,m}\|_{\infty} = \begin{cases} |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } 0 < \nu \leq 1, \\ \nu |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } \nu > 1, \end{cases}$$

Scalability analysis

First attempt: classical matrix norm, e.g.

$$\|T_{N,m}\|_{\infty} = \begin{cases} |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } 0 < \nu \leq 1, \\ \nu |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } \nu > 1, \end{cases}$$

Good news: independent of N

Bad news: larger than 1 for small ν

Observation: $\tilde{\rho}(m) = \nu (C_{1,m}(\Delta t))^2 + (C_{2,m}(\Delta t))^2 < 1$

Scalability analysis

First attempt: classical matrix norm, e.g.

$$\|T_{N,m}\|_{\infty} = \begin{cases} |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } 0 < \nu \leq 1, \\ \nu |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } \nu > 1, \end{cases}$$

Remedy: a modified infinity norm

$$|||T_{N,m}||| := \max_{i=1,\dots,2(N-1)} \left(\sum_{j=1}^{2(N-1)} (D^{-1} T_{N,m} D)_{ij}^2 \right)^{\frac{1}{2}}$$

Good news: independent of N

Bad news: larger than 1 for small ν

Observation: $\tilde{\rho}(m) = \nu (C_{1,m}(\Delta t))^2 + (C_{2,m}(\Delta t))^2 < 1$

$$D := \begin{bmatrix} D_d & & \\ & \ddots & \\ & & D_d \end{bmatrix}$$
$$D_d := \begin{bmatrix} 1 & 0 \\ 0 & \sqrt{\nu} \end{bmatrix}$$

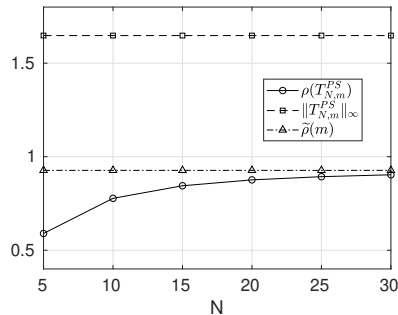
Scalability analysis

First attempt: classical matrix norm, e.g.

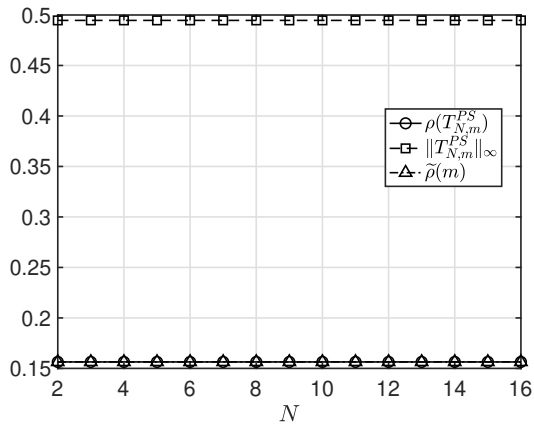
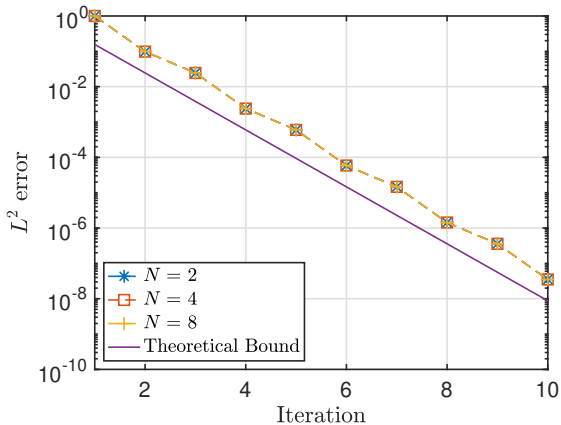
$$\|T_{N,m}\|_{\infty} = \begin{cases} |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } 0 < \nu \leq 1, \\ \nu |C_{1,m}(\Delta t)| + |C_{2,m}(\Delta t)|, & \text{if } \nu > 1, \end{cases}$$

Remedy: a modified infinity norm

$$|||T_{N,m}||| := \max_{i=1,\dots,2(N-1)} \left(\sum_{j=1}^{2(N-1)} (D^{-1}T_{N,m}D)_{ij}^2 \right)^{\frac{1}{2}}$$



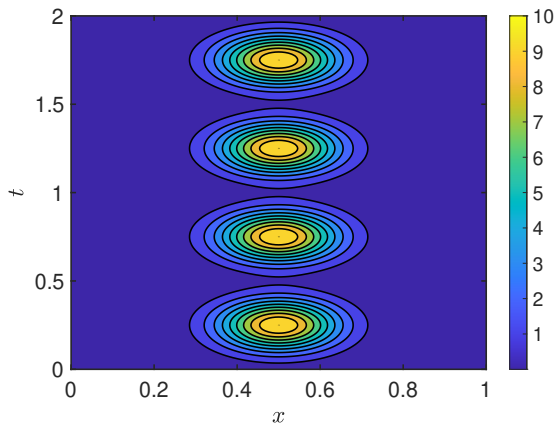
Scalability result



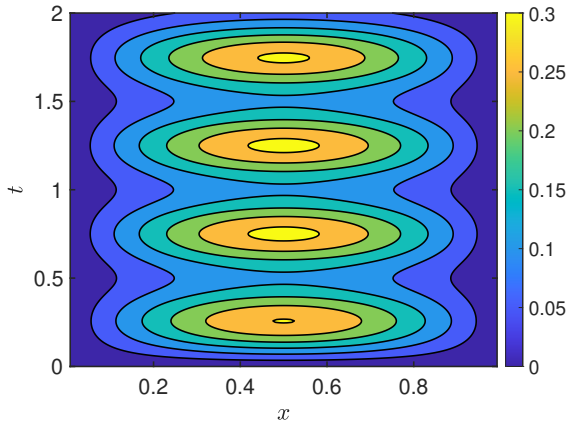
Periodic heating-cooling

Target function: $\hat{y}(x, t) = 10 \sum_{n=1}^N \exp \left(-50 \left(\left(x - \frac{L}{2} \right)^2 + \left(t - \frac{(2n-1)\Delta t}{2} \right)^2 \right) \right)$

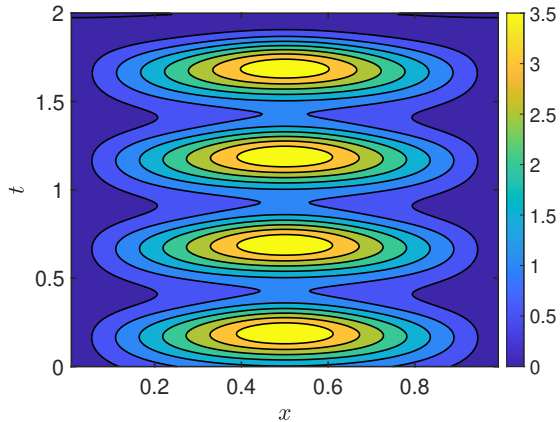
Illustration with four periods



Periodic heating-cooling

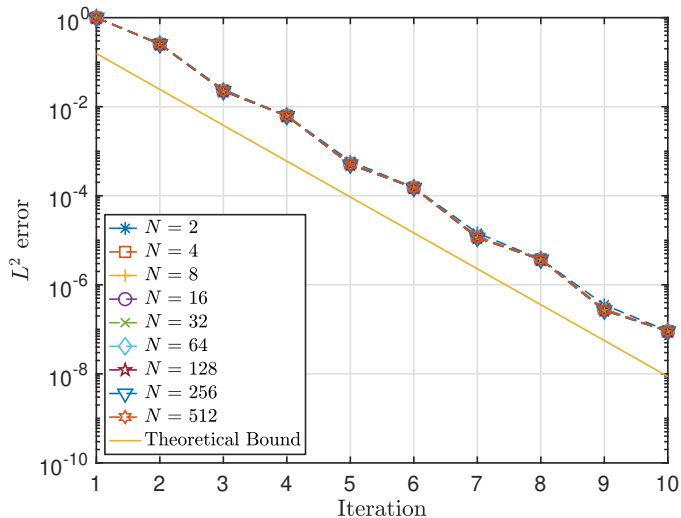


State solution y



Optimal control u

Periodic heating-cooling



Unconstrained

$$U_{\text{ad}} = L^2(Q)$$

Optimality condition:

$$\nu u - p = 0$$

Unconstrained

$$U_{\text{ad}} = L^2(Q)$$

Optimality condition:

$$\nu u - p = 0$$

Box constraint:

$$U_{\text{ad}} := \{v \in L^2(Q) : u_{\min} \leq v \leq u_{\max}\}$$

Optimality condition:

$$u = \mathbb{P}_{U_{\text{ad}}}(\nu^{-1}p) := \max\{u_{\min}, \min\{\nu^{-1}p, u_{\max}\}\}$$

Unconstrained

$$U_{\text{ad}} = L^2(Q)$$

Optimality condition:

$$\nu u - p = 0$$

Box constraint:

$$U_{\text{ad}} := \{v \in L^2(Q) : u_{\min} \leq v \leq u_{\max}\}$$

Optimality condition:

$$u = \mathbb{P}_{U_{\text{ad}}}(\nu^{-1}p) := \max\{u_{\min}, \min\{\nu^{-1}p, u_{\max}\}\}$$

Reduced optimality system:

$$\begin{array}{llll} \partial_t y - \Delta y = \mathbb{P}_{U_{\text{ad}}}(\nu^{-1}p) & \text{in } Q & \partial_t p + \Delta p = y - \hat{y} & \text{in } Q \\ y = g & \text{on } \Sigma & p = 0 & \text{on } \Sigma \\ y = y_0 & \text{on } \Sigma_0 & p = 0 & \text{on } \Sigma_T \end{array}$$

Box constraint - Two strategies

Key idea: Semismooth Newton for

$$F(u) := u - \mathbb{P}_{U_{\text{ad}}}(\nu^{-1}p) = 0$$

and time domain decomposition as inner solver

Box constraint - Two strategies

Semismooth Newton step:

$$u^{k+1} = u^k + \delta u^k$$

with δu^k solves $G(u^k)[\delta u^k] = -F(u^k)$

Box constraint - Two strategies

Semismooth Newton step:

$$u^{k+1} = u^k + \delta u^k$$

with δu^k solves $G(u^k)[\delta u^k] = -F(u^k)$

Apply time domain decomposition for correction

$$\partial_t \delta y^k - \Delta \delta y^k = \delta u^k \quad \text{in } Q$$

$$\delta y^k = 0 \quad \text{on } \Sigma$$

$$\delta y^k = 0 \quad \text{on } \Sigma_0$$

$$\partial_t \delta p^k + \Delta \delta p^k = \delta y^k \quad \text{in } Q$$

$$\delta p^k = 0 \quad \text{on } \Sigma$$

$$\delta p^k = 0 \quad \text{on } \Sigma_T$$

$$\delta u^k - \chi_{I^k} \nu^{-1} \delta p^k = -F(u^k) \quad \text{in } Q$$

Box constraint - Two strategies

Semismooth Newton step:

$$u^{k+1} = u^k + \delta u^k$$

with δu^k solves $G(u^k)[\delta u^k] = -F(u^k)$

Apply time domain decomposition for correction

$$\partial_t \delta y^k - \Delta \delta y^k = \delta u^k \quad \text{in } Q$$

$$\delta y^k = 0 \quad \text{on } \Sigma$$

$$\delta y^k = 0 \quad \text{on } \Sigma_0$$

$$\partial_t \delta p^k + \Delta \delta p^k = \delta y^k \quad \text{in } Q$$

$$\delta p^k = 0 \quad \text{on } \Sigma$$

$$\delta p^k = 0 \quad \text{on } \Sigma_T$$

$$\delta u^k - \chi_{I^k} \nu^{-1} \delta p^k = -F(u^k) \quad \text{in } Q$$

Key idea: first decomposing in time into N subintervals

$$\partial_t y_n^\ell - \Delta y_n^\ell = \mathbb{P}_{U_{\text{ad}}}(\nu^{-1} p_n^\ell) \quad \text{in } Q_n$$

$$y_n^\ell = 0 \quad \text{on } \Sigma$$

$$y_n^\ell = y_{n-1}^{\ell-1} \quad \text{on } \Sigma_{n-1}$$

$$\partial_t p_n^\ell + \Delta p_n^\ell = y_n^\ell - \hat{y}_n \quad \text{in } Q_n$$

$$p_n^\ell = 0 \quad \text{on } \Sigma$$

$$p_n^\ell = p_{n+1}^{\ell-1} \quad \text{on } \Sigma_n$$

Box constraint - Two strategies

Semismooth Newton step:

$$u^{k+1} = u^k + \delta u^k$$

with δu^k solves $G(u^k)[\delta u^k] = -F(u^k)$

Apply time domain decomposition for correction

$$\partial_t \delta y^k - \Delta \delta y^k = \delta u^k \quad \text{in } Q$$

$$\delta y^k = 0 \quad \text{on } \Sigma$$

$$\delta y^k = 0 \quad \text{on } \Sigma_0$$

$$\partial_t \delta p^k + \Delta \delta p^k = \delta y^k \quad \text{in } Q$$

$$\delta p^k = 0 \quad \text{on } \Sigma$$

$$\delta p^k = 0 \quad \text{on } \Sigma_T$$

$$\delta u^k - \chi_{I^k} \nu^{-1} \delta p^k = -F(u^k) \quad \text{in } Q$$

Key idea: first decomposing in time into N subintervals

$$\partial_t y_n^\ell - \Delta y_n^\ell = \mathbb{P}_{U_{\text{ad}}}(\nu^{-1} p_n^\ell) \quad \text{in } Q_n$$

$$y_n^\ell = 0 \quad \text{on } \Sigma$$

$$y_n^\ell = y_{n-1}^{\ell-1} \quad \text{on } \Sigma_{n-1}$$

$$\partial_t p_n^\ell + \Delta p_n^\ell = y_n^\ell - \hat{y}_n \quad \text{in } Q_n$$

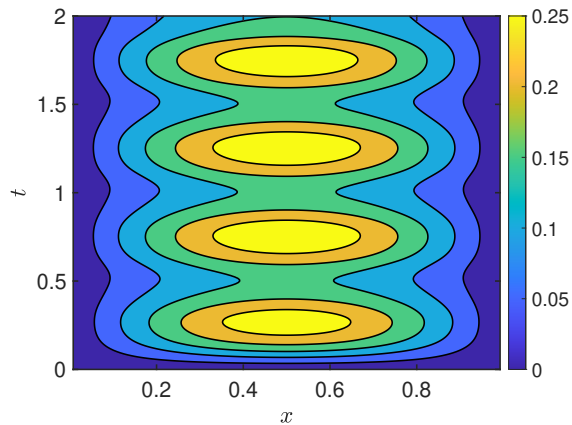
$$p_n^\ell = 0 \quad \text{on } \Sigma$$

$$p_n^\ell = p_{n+1}^{\ell-1} \quad \text{on } \Sigma_n$$

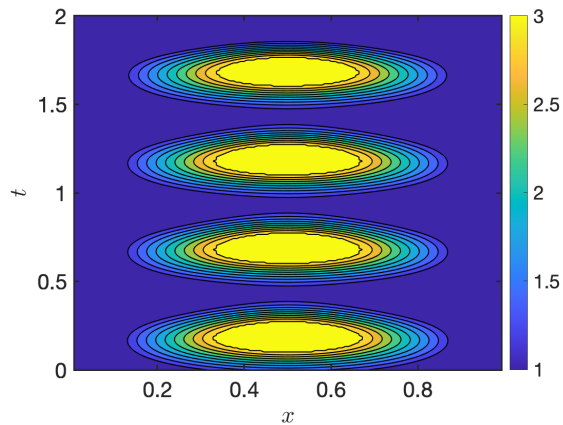
then apply semismooth Newton **within each time subinterval**.

Box constraint - Numerical results

Admissible set: $u_{\min} = 1$ and $u_{\max} = 3$



State solution y



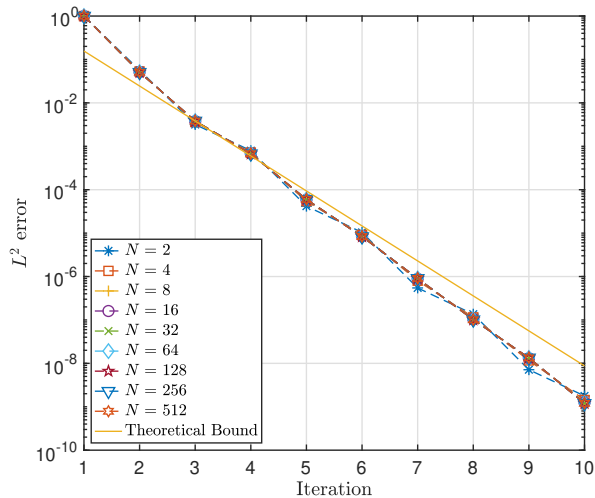
Optimal control u

Box constraint - Numerical results

Strategy 1

N	Outer SMNewton It.	Total PSM It.	Averaged PSM It.
8	3	35	11.7
16	3	46	15.3
32	3	71	23.6
64	3	77	25.7
128	3	77	25.7
256	3	77	25.7
512	3	77	25.7

Strategy 2



What we learned

Time domain decomposition:

- Time decomposition is **different** from space decomposition
- Forward-backward structure is **important** for alternating Schwarz to converge without overlap
- Relaxation **can improve** the convergence, but is **not necessary** for the method to converge

Scalability:

- Time Parallel Schwarz is **weak** scalable (excellent scalability for heat problems)
- **Two different strategies** to handle box constraint

What we learned

Time domain decomposition:

- Time decomposition is **different** from space decomposition
- Forward-backward structure is **important** for alternating Schwarz to converge without overlap
- Relaxation **can improve** the convergence, but is **not necessary** for the method to converge

Scalability:

- Time Parallel Schwarz is **weak** scalable (excellent scalability for heat problems)
- **Two different strategies** to handle box constraint

Conclusion: Domain decomposition provides a framework rather than just a solver !

What we learned

Time domain decomposition:

- Time decomposition is **different** from space decomposition
- Forward-backward structure is **important** for alternating Schwarz to converge without overlap
- Relaxation **can improve** the convergence, but is **not necessary** for the method to converge

Scalability:

- Time Parallel Schwarz is **weak** scalable (excellent scalability for heat problems)
- **Two different strategies** to handle box constraint

Conclusion: Domain decomposition provides a framework rather than just a solver !

Thank you for your attention ! 感谢您的聆听!