



# DOMAIN DECOMPOSITION BASED PRECONDITIONING FOR A UNIQUE CONTINUATION PROBLEM SUBJECT TO THE WAVE EQUATION

BRUNELLI Filippo, DELAY Guillaume,  
NATAF Frédéric, PAROLIN Emile,  
TOURNIER Pierre-Henri

Sorbonne Université, LJLL  
Inria Paris

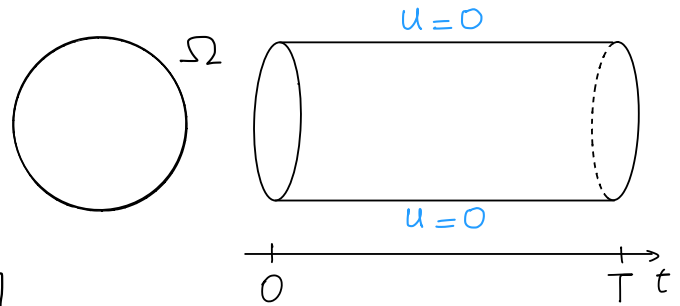
03/06/2026 - CANUM 2026

# THE MODEL PROBLEM

$\Omega \subseteq \mathbb{R}^d$ ,  $\partial\Omega$  smooth,  
 $T > 0$ ,

$$Q := \Omega \times [0, T],$$

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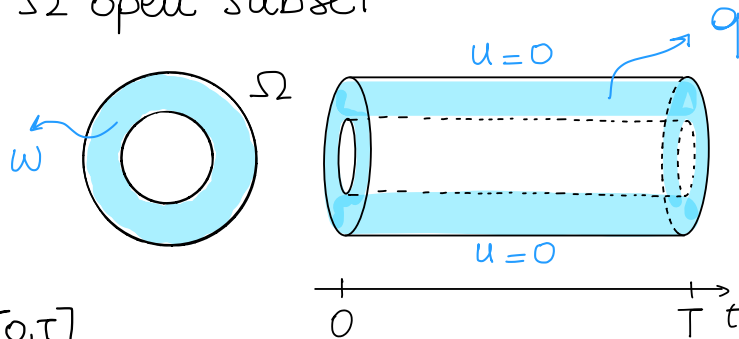
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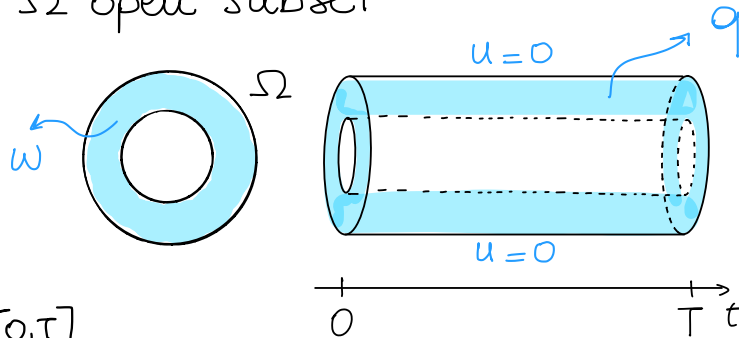


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$$Z := \{u \in C^0([0, T]; L^2(\Omega)) \cap C^1([0, T], H^{-1}(\Omega)), \square u \in L^2(0, T, H^{-1}(\Omega))\}.$$

$$W := \{u \in Z, \square u = 0\}$$

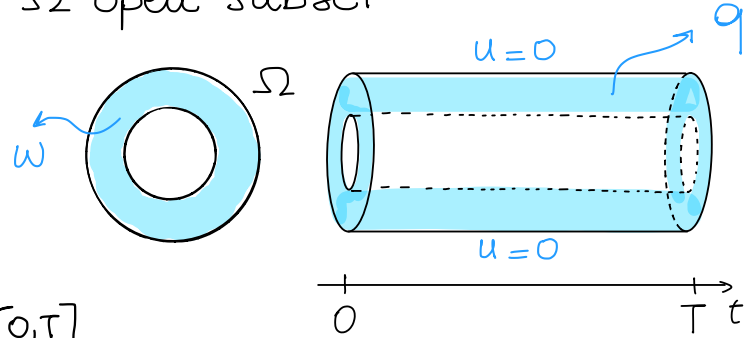
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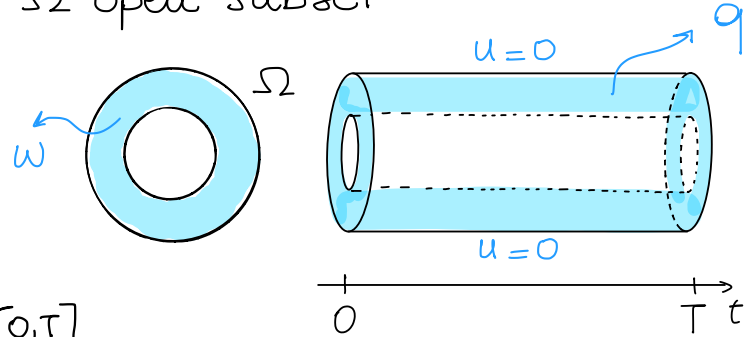
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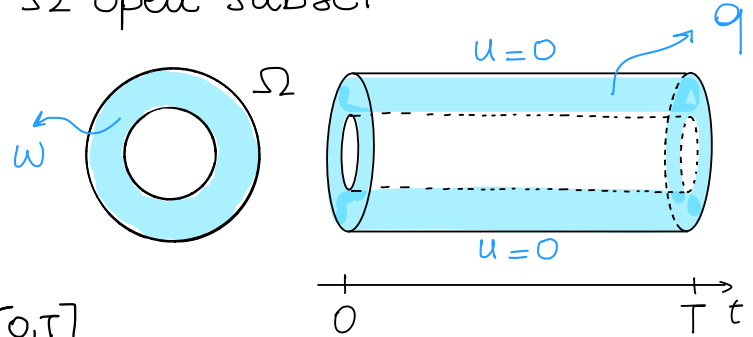
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$$(1) \quad \min_{u \in W} \mathcal{J}(u) := \frac{1}{2} \|u - u_d\|_{L^2(q)}^2.$$

We look for the  $u \in W$  that minimises the  $L^2$ -misfit with  $u_d \in L^2(q)$

# FIRST ORDER OPTIMALITY CONDITIONS

The PDE constraint is weakly imposed using a Lagrange multiplier  $\lambda \in Y := L^2(0, T, H_0^1(\Omega))$

$$\mathcal{L}(u, \lambda) := \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle_{H^{-1}, H_0^1}.$$

$$\downarrow \nabla \mathcal{L} = 0$$

Find  $u \in Z, \lambda \in Y$  s.t.

$$(2) \quad \begin{aligned} (u, v)_q + \int_0^T \langle \square v, \lambda \rangle_{H^{-1}, H_0^1} &= (u_d, v)_q & \forall v \in Z, \\ \int_0^T \langle \square u, \psi \rangle_{H^{-1}, H_0^1} &= 0 & \forall \psi \in Y. \end{aligned}$$

**Lemma** [Cîndea, Münch 2015]

Under GCC, Problem (2) is well-posed and the unique solution is the unique saddle point of  $\mathcal{L}$  and the unique minimiser in (1)

# MOTIVATION BEHIND THE WORK

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▽ Time lost its connotation of unidirectional variable.

⇒ The problem to solve is globally coupled in space AND time.

▽ We switched to a mixed formulation.

⇒ Additional degrees of freedom for the Lagrange multiplier and indefinite saddle point system.

▽ It is an inverse problem

⇒ We expect strong ill-conditioning at the discrete level

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Linear system  
of large size

+

bad spectral  
properties

⇒

Both direct and  
iterative methods  
struggle

# DECOMPOSE - THEN - OPTIMIZE APPROACH (ES)

[Coquet, Gauder, Vieira, 2025, arXiv]

DECOMPOSE  
then  
OPTIMIZE

- 1) Decompose the domain  $\Omega$
- 2) Write the PDE in a multi domain fashion  
+  
Introduce continuity conditions at the subdomains' interfaces
- 3) Treat each constraint using some optimization technique

OPTIMIZE  
then  
DECOMPOSE

- 1) Write the first order optimality conditions
- 2) Decompose the domain  $\Omega$
- 3) Apply some DD paradigm on the resulting system

# DECOMPOSITION OF THE TIME AXIS

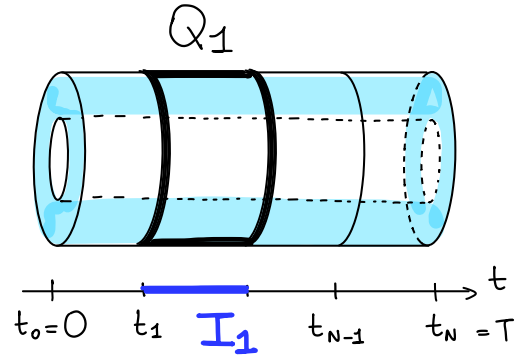
Let us subdivide  $[0, T]$  into  $N$  subintervals

$$I_n := [t_n, t_{n+1}], \quad n \in \llbracket 0, N-1 \rrbracket$$

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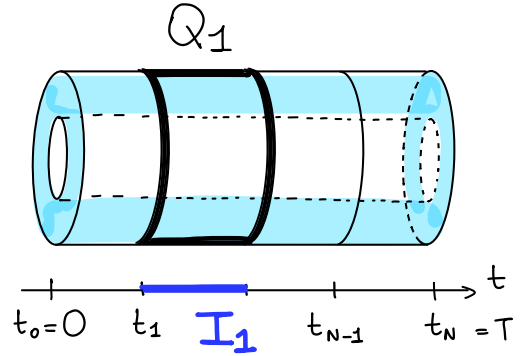
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$$\underline{Z}^\circ := \left\{ \underline{U} \in \underline{Z}, \text{ s.t. } \begin{array}{l} \llbracket \underline{U} \rrbracket_n := u_n(t_n) - u_{n-1}(t_n) = 0 \text{ in } L^2(\Omega), \\ \llbracket \partial_t \underline{U} \rrbracket_n := \partial_t u_n(t_n) - \partial_t u_{n-1}(t_n) = 0 \text{ in } H^{-1}(\Omega) \end{array} \right\}.$$

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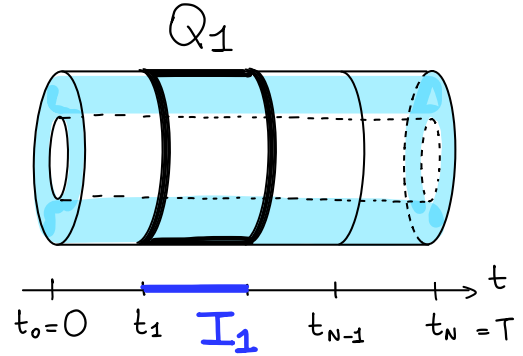
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CONTINUITY  
CONDITIONS  
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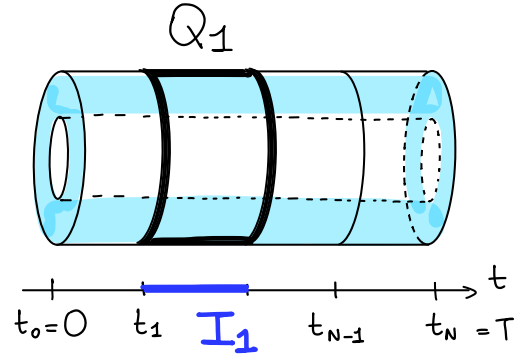
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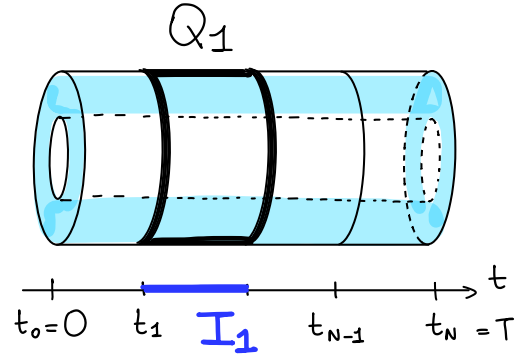
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CONTINUITY  
CONDITIONS  
IN TIME

regularity  
+  
continuity  
conditions

locally  
enforced PDE

# MULTIDOMAIN FORMULATION

(Decompose -then-optimise approach)

$$W^0 := \{ \underline{u} \in \mathbf{Z}^0, \square u_n = 0, \forall n = 0, \dots, N-1 \},$$

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multidomain PDE  
constrained  
minimisation

## Lemma

The multidomain formulation is equivalent to the starting constraint minimisation problem (1)

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- $\square u_n = 0, \forall n = 0, \dots, N-1 \Rightarrow$  Lagrange multipliers
- $[\underline{U}]_n := u_n(t_n) - u_{n-1}(t_n) = 0$   
 $\underline{\Delta} = (\lambda_n)_n \in Y$
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$\Rightarrow$  "Least-squares"  
penalisation

# "LEAST-SQUARES" PENALISATION

For  $\sigma > 0$ , let us consider the following Lagrangian function

$$\mathcal{L}_\sigma^P(\underline{U}, \underline{\Delta}) := \frac{1}{2} \sum_{n=0}^{N-1} \|u_n - u_{d,n}\|_{q_n}^2$$

$$+ \frac{1}{2\sigma} \sum_{n=1}^{N-1} \left\{ \|\llbracket U \rrbracket_n\|_{L^2(\Omega)}^2 + \|\partial_t \llbracket U \rrbracket_n\|_{H^{-1}(\Omega)}^2 \right\}$$

for  $\underline{U} \in \mathbb{Z}$ ,  $\underline{\Delta} \in \mathbb{Y}$

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NO continuity  
in time

penalisation  
parameter

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$N$  local  
multipliers

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Find  $(\underline{U}_\sigma, \underline{\Delta}_\sigma) \in \mathbf{Z} \times \mathbf{Y}$  s.t.

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$$(4) \quad \sum_{n=0}^{N-1} (u_n, v_n)_{q_n} + \frac{1}{\sigma} \mathbf{s}(\underline{U}_\sigma, \underline{V}) + \mathbf{b}(\underline{V}, \underline{\Delta}_\sigma) = \mathbf{L}(\underline{V}) \quad \forall \underline{V} \in \mathbf{Z},$$

$$\mathbf{b}(\underline{U}_\sigma, \underline{\Psi}) = 0 \quad \forall \underline{\Psi} \in \mathbf{Y}.$$

$$\mathbf{s}(\underline{U}_\sigma, \underline{V}) := \sum_{n=1}^{N-1} \left\{ (\llbracket U_\sigma \rrbracket_n, \llbracket V \rrbracket_n)_\Omega + \langle \llbracket \partial_t U_\sigma \rrbracket_n, \llbracket \partial_t V \rrbracket_n \rangle_{H^{-1}(\Omega)} \right\}$$

Coupling between neighbouring slabs  $\Rightarrow$  "block-tridiagonal" structure

# "LEAST-SQUARES" PENALISATION

For  $\sigma > 0$ , let us consider the following Lagrangian function

$$\mathcal{L}_\sigma^P(\underline{U}, \underline{\Delta}) := \frac{1}{2} \sum_{n=0}^{N-1} \|u_n - u_{d,n}\|_{q_n}^2$$

$$+ \frac{1}{2\sigma} \sum_{n=1}^{N-1} \left\{ \|\llbracket U \rrbracket_n\|_{L^2(\Omega)}^2 + \|\llbracket \partial_t U \rrbracket_n\|_{H^{-1}(\Omega)}^2 \right\}$$

for  $\underline{U} \in \mathbf{Z}$ ,  $\underline{\Delta} \in \mathbf{Y}$

$$+ \sum_{n=0}^{N-1} \int_{I_n} \langle \square u_n, \lambda_n \rangle_{H^{-1}, H_0^1}$$

$$\downarrow \nabla \mathcal{L}_\sigma^P = 0$$

Find  $(\underline{U}_\sigma, \underline{\Delta}_\sigma) \in \mathbf{Z} \times \mathbf{Y}$  s.t.

$$(4) \quad \sum_{n=0}^{N-1} (u_n, v_n)_{q_n} + \frac{1}{\sigma} \mathbf{s}(\underline{U}_\sigma, \underline{V}) + \mathbf{b}(\underline{V}, \underline{\Delta}_\sigma) = \mathbf{L}(\underline{V}) \quad \forall \underline{V} \in \mathbf{Z},$$

$$\mathbf{b}(\underline{U}_\sigma, \underline{\Psi}) = 0 \quad \forall \underline{\Psi} \in \mathbf{Y}.$$

## Lemma

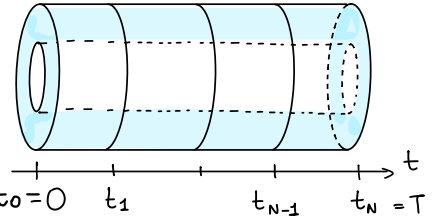
For  $\sigma > 0$ , problem (4) is well-posed and the unique solution is the unique minimiser of the penalised functional.

## Lemma

Given  $\underline{U}$  solution of (3),  $\underline{U}_\sigma \rightarrow \underline{U}$  in  $\mathbf{Z}$  for  $\sigma \rightarrow 0$ .

# SPLITTING THE COUPLING TERM

Notation:  $y, z \in L^2(\Omega)$   
 $\partial_t y, \partial_t z \in H^{-1}(\Omega) \Rightarrow \langle y, z \rangle := (y, z)_\Omega + \langle \partial_t y, \partial_t z \rangle_{H^{-1}}$



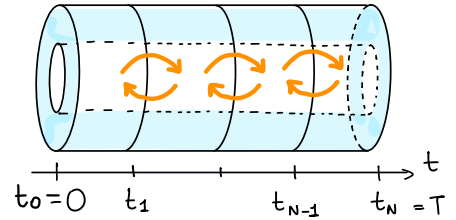
$$\mathbf{S}(\underline{U}, \underline{V}) = \sum_{n=1}^{N-1} \langle \llbracket U \rrbracket_n, \llbracket V \rrbracket_n \rangle$$

inter-subdomain coupling

$$= \sum_{n=1}^{N-1} \left\{ \begin{aligned} &\langle \mathcal{U}_{n-1}(t_n), \mathcal{V}_{n-1}(t_n) \rangle + \langle \mathcal{U}_n(t_n), \mathcal{V}_{n-1}(t_n) \rangle \\ &+ \langle \mathcal{U}_{n-1}(t_n), \mathcal{V}_n(t_n) \rangle + \langle \mathcal{U}_n(t_n), \mathcal{V}_n(t_n) \rangle \end{aligned} \right\}$$

# SPLITTING THE COUPLING TERM

Notation:  $y, z \in L^2(\Omega)$   
 $\partial_t y, \partial_t z \in H^{-1}(\Omega) \Rightarrow \langle y, z \rangle := (y, z)_\Omega + \langle \partial_t y, \partial_t z \rangle_{H^{-1}}$



*inter-subdomain coupling*

$$\mathbf{S}(\underline{U}, \underline{V}) = \sum_{n=1}^{N-1} \langle \llbracket \underline{U} \rrbracket_n, \llbracket \underline{V} \rrbracket_n \rangle$$

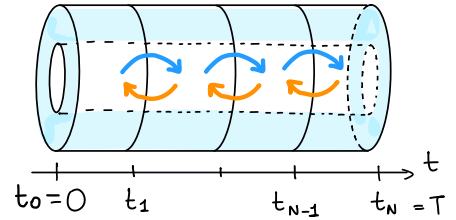
$$= \sum_{n=1}^{N-1} \left\{ \langle \mathcal{U}_{n-1}(t_n), \mathcal{V}_{n-1}(t_n) \rangle + \langle \mathcal{U}_n(t_n), \mathcal{V}_{n-1}(t_n) \rangle \right. \\ \left. + \langle \mathcal{U}_{n-1}(t_n), \mathcal{V}_n(t_n) \rangle + \langle \mathcal{U}_n(t_n), \mathcal{V}_n(t_n) \rangle \right\}$$

$$\Rightarrow \mathbf{S}(\underline{U}, \underline{V}) = \mathbf{S}_d(\underline{U}, \underline{V}) - \mathbf{S}_c(\underline{U}, \underline{V}).$$

*Jacobi-like splitting*

# SPLITTING THE COUPLING TERM

Notation:  $y, z \in L^2(\Omega)$   
 $\partial_t y, \partial_t z \in H^{-1}(\Omega) \Rightarrow \langle y, z \rangle := (y, z)_\Omega + \langle \partial_t y, \partial_t z \rangle_{H^{-1}}$



inter-subdomain coupling

$$\mathbf{S}(\underline{U}, \underline{V}) = \sum_{n=1}^{N-1} \langle \llbracket \underline{U} \rrbracket_n, \llbracket \underline{V} \rrbracket_n \rangle$$

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$$\Rightarrow \mathbf{S}(\underline{U}, \underline{V}) = \mathbf{S}_d(\underline{U}, \underline{V}) - \mathbf{S}_c(\underline{U}, \underline{V}).$$

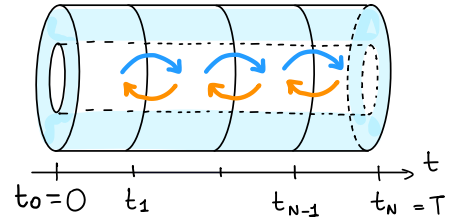
Jacobi-like splitting

$$\Rightarrow \mathbf{S}(\underline{U}, \underline{V}) = \mathbf{S}_f(\underline{U}, \underline{V}) - \mathbf{S}_b(\underline{U}, \underline{V}).$$

Gauss-Seidel-like splitting

# SPLITTING THE COUPLING TERM

Notation:  $y, z \in L^2(\Omega)$   
 $\partial_t y, \partial_t z \in H^{-1}(\Omega) \Rightarrow \langle y, z \rangle := (y, z)_{\Omega} + \langle \partial_t y, \partial_t z \rangle_{H^{-1}}$



inter-subdomain coupling

$$\mathbf{S}(\underline{U}, \underline{V}) = \sum_{n=1}^{N-1} \langle \llbracket \underline{U} \rrbracket_n, \llbracket \underline{V} \rrbracket_n \rangle$$

$$= \sum_{n=1}^{N-1} \left\{ \langle \mathcal{U}_{n-1}(t_n), \mathcal{V}_{n-1}(t_n) \rangle + \langle \mathcal{U}_n(t_n), \mathcal{V}_{n-1}(t_n) \rangle \right. \\ \left. + \langle \mathcal{U}_{n-1}(t_n), \mathcal{V}_n(t_n) \rangle + \langle \mathcal{U}_n(t_n), \mathcal{V}_n(t_n) \rangle \right\}$$

$$\Rightarrow \mathbf{S}(\underline{U}, \underline{V}) = \mathbf{S}_d(\underline{U}, \underline{V}) - \mathbf{S}_c(\underline{U}, \underline{V}).$$

Jacobi-like splitting

$$\Rightarrow \mathbf{S}(\underline{U}, \underline{V}) = \mathbf{S}_f(\underline{U}, \underline{V}) - \mathbf{S}_b(\underline{U}, \underline{V}).$$

Gauss-Seidel-like splitting

For  $\delta \geq 0$

$$\Rightarrow \mathbf{S}(\underline{U}, \underline{V}) = (1+\delta) \mathbf{S}_d(\underline{U}, \underline{V}) - [\delta \mathbf{S}_d(\underline{U}, \underline{V}) + \mathbf{S}_c(\underline{U}, \underline{V})]$$

JOR-like splitting

$$\Rightarrow \mathbf{S}(\underline{U}, \underline{V}) = \mathbf{S}_f(\underline{U}, \underline{V}) + \delta \mathbf{S}_d(\underline{U}, \underline{V}) - [\delta \mathbf{S}_d(\underline{U}, \underline{V}) + \mathbf{S}_b(\underline{U}, \underline{V})].$$

SOR-like splitting

# SPLITTING ALGORITHMS

Algorithm 1 [JOR-like]

▷ Find  $(\underline{U}_\sigma^{\ell+1}, \underline{\Delta}_\sigma^{\ell+1}) \in \mathbf{Z} \times \mathbf{Y}$  s.t.

$$\mathbf{a}(\underline{U}_\sigma^{\ell+1}, \underline{V}) + \frac{1+\delta}{\sigma} \mathbf{s}_d(\underline{U}_\sigma^{\ell+1}, \underline{V}) + \mathbf{b}(\underline{V}, \underline{\Delta}_\sigma^{\ell+1}) = \mathbf{L}(\underline{V}) + \frac{\delta}{\sigma} \mathbf{s}_d(\underline{U}_\sigma^\ell, \underline{V}) + \frac{1}{\sigma} \mathbf{s}_c(\underline{U}_\sigma^\ell, \underline{V}),$$

$$\mathbf{b}(\underline{U}_\sigma^{\ell+1}, \underline{\Psi}) = \mathbf{0},$$

$$\forall \underline{V} \in \mathbf{Z}, \forall \underline{\Psi} \in \mathbf{Y}$$

# SPLITTING ALGORITHMS

Algorithm 1 [JOR-like]

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$$\mathbf{b}(\underline{u}_\sigma^{\ell+1}, \underline{\Psi}) = 0,$$

$$\forall \underline{v} \in \mathbf{Z}, \forall \underline{\Psi} \in \mathbf{Y}$$

Algorithm 2 [SOR-like]

▷ Find  $(\underline{u}_\sigma^{\ell+1}, \underline{\Delta}_\sigma^{\ell+1}) \in \mathbf{Z} \times \mathbf{Y}$  s.t.

$$\mathbf{a}(\underline{u}_\sigma^{\ell+1}, \underline{v}) + \frac{1}{\sigma} \mathbf{s}_f^{(\ell+1)}(\underline{u}_\sigma^{\ell+1}, \underline{v}) + \frac{\gamma}{\sigma} \mathbf{s}_d^{(\ell+1)}(\underline{u}_\sigma^{\ell+1}, \underline{v}) + \mathbf{b}(\underline{v}, \underline{\Delta}_\sigma^{\ell+1}) = \mathbf{L}(\underline{v}) + \frac{\gamma}{\sigma} \mathbf{s}_d^{(\ell)}(\underline{u}_\sigma^{\ell}, \underline{v})$$
$$+ \frac{1}{\sigma} \mathbf{s}_b^{(\ell)}(\underline{u}_\sigma^{\ell}, \underline{v}),$$
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# SPLITTING ALGORITHMS

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$$\mathbf{b}(\underline{u}_\sigma^{\ell+1}, \underline{\Psi}) = 0,$$

$$\forall \underline{v} \in \mathbf{Z}, \forall \underline{\Psi} \in \mathbf{Y}$$

Algorithm 2 [SOR-like]

▷ Find  $(\underline{u}_\sigma^{\ell+1}, \underline{\Delta}_\sigma^{\ell+1}) \in \mathbf{Z} \times \mathbf{Y}$  s.t.

$$\mathbf{a}(\underline{u}_\sigma^{\ell+1}, \underline{v}) + \frac{1}{\sigma} \mathbf{s}_f(\underline{u}_\sigma^{\ell+1}, \underline{v}) + \frac{\gamma}{\sigma} \mathbf{s}_d(\underline{u}_\sigma^{\ell+1}, \underline{v}) + \mathbf{b}(\underline{v}, \underline{\Delta}_\sigma^{\ell+1}) = \mathbf{L}(\underline{v}) + \frac{\gamma}{\sigma} \mathbf{s}_d(\underline{u}_\sigma^\ell, \underline{v})$$

$$+ \frac{1}{\sigma} \mathbf{s}_b(\underline{u}_\sigma^\ell, \underline{v}),$$

$$\mathbf{b}(\underline{u}_\sigma^{\ell+1}, \underline{\Psi}) = 0,$$

Theorem [Convergence]

For  $\sigma > 0$ , Algorithms 1 and 2 converge  $\forall \gamma > 0$ , namely  $\|\underline{u}_\sigma^\ell - \underline{u}_\sigma\|_{\mathbf{Z}} \rightarrow 0$  for  $\ell \rightarrow \infty$ , where  $\underline{u}_\sigma$  is the solution of (4).

In particular, Algorithm 2 converges also for  $\gamma = 0$ .

# SPLITTING ALGORITHMS

Algorithm 1 [JOR-like]

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$$\mathbf{b}(\underline{U}_\sigma^{\ell+1}, \underline{\Psi}) = 0,$$

⇒ PARALLEL

$\forall \underline{V} \in \mathbf{Z}, \forall \underline{\Psi} \in \mathbf{Y}$

Algorithm 2 [SOR-like]

▷ Find  $(\underline{U}_\sigma^{\ell+1}, \underline{\Delta}_\sigma^{\ell+1}) \in \mathbf{Z} \times \mathbf{Y}$  s.t.

$$\mathbf{a}(\underline{U}_\sigma^{\ell+1}, \underline{V}) + \frac{1}{\sigma} \mathbf{s}_f^{(\ell+1)}(\underline{U}_\sigma^{\ell+1}, \underline{V}) + \frac{\gamma}{\sigma} \mathbf{s}_d^{(\ell+1)}(\underline{U}_\sigma^{\ell+1}, \underline{V}) + \mathbf{b}(\underline{V}, \underline{\Delta}_\sigma^{\ell+1}) = \mathbf{L}(\underline{V}) + \frac{\gamma}{\sigma} \mathbf{s}_d^{(\ell)}(\underline{U}_\sigma^{\ell}, \underline{V}) + \frac{1}{\sigma} \mathbf{s}_b^{(\ell)}(\underline{U}_\sigma^{\ell}, \underline{V}),$$

$$\mathbf{b}(\underline{U}_\sigma^{\ell+1}, \underline{\Psi}) = 0,$$

⇒ SEQUENTIAL

Theorem [Convergence]

For  $\sigma > 0$ , Algorithms 1 and 2 converge  $\forall \gamma > 0$ , namely  $\|\underline{U}_\sigma^\ell - \underline{U}_\sigma\|_{\mathbf{Z}} \rightarrow 0$  for  $\ell \rightarrow \infty$ , where  $\underline{U}_\sigma$  is the solution of (4).

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# PRECONDITIONED FIXED POINT ITERATION

As for linear system of equation  $Au = b$

splitting  
algorithm



left-preconditioned  
fixed point iteration

$$Mu^{l+1} = Nu^l + b$$

with  $A = M - N$

$$u^{l+1} = u^l + M^{-1}(Au^l - b)$$

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## Algorithm 1 [JOR-like]

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$$\boxed{b}(\underline{u}_\sigma^{l+1}, \underline{\Psi}) = 0,$$

$\forall \underline{v} \in \mathbf{Z}, \forall \underline{\Psi} \in \mathbf{Y}.$

$M$

$N$

# PRECONDITIONED FIXED POINT ITERATION

As for linear system of equation  $Au = b$

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$$Mu^{l+1} = Nu^l + b$$

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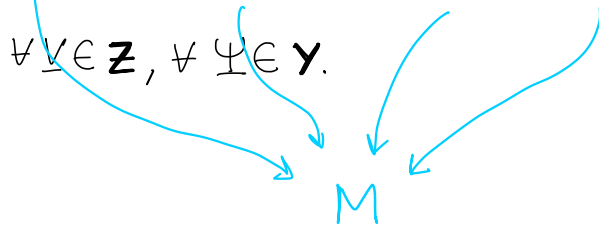
$$u^{l+1} = u^l + M^{-1}(Au^l - b)$$

## Algorithm 2 [SOR-like]

▷ Find  $(\underline{u}_\sigma^{l+1}, \underline{\Delta}_\sigma^{l+1}) \in \mathbf{Z} \times \mathbf{Y}$  s.t.

$$\boxed{a}(\underline{u}_\sigma^{l+1}, \underline{v}) + \boxed{\frac{1}{\sigma} \mathbf{s}_f}(\underline{u}_\sigma^{l+1}, \underline{v}) + \boxed{\frac{\delta}{\sigma} \mathbf{s}_d}(\underline{u}_\sigma^{l+1}, \underline{v}) + \boxed{\mathbf{b}}(\underline{v}, \underline{\Delta}_\sigma^{l+1}) = \mathbf{L}(\underline{v}) + \boxed{\frac{\delta}{\sigma} \mathbf{s}_d}(\underline{u}_\sigma^l, \underline{v}) + \boxed{\frac{1}{\sigma} \mathbf{s}_b}(\underline{u}_\sigma^l, \underline{v}),$$

$$\boxed{\mathbf{b}}(\underline{u}_\sigma^{l+1}, \underline{\Psi}) = 0,$$



# A RECIPE FOR THE DISCRETE LEVEL

Continuous level

Discrete level

$$u \in Z, \lambda \in Y$$

$$\mathcal{L}(u, \lambda) = \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle$$

↑ DECOMPOSE  
then  
↓ OPTIMIZE

$$\underline{u} \in \underline{Z}, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}_{MD}(\underline{u}, \underline{\Delta}) = \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 + \int_{I_n} \langle \square u_n, \lambda_n \rangle \right\}$$

↑  $\sigma \rightarrow 0$

$$\underline{u} \in \underline{Z}, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}^P(\underline{u}, \underline{\Delta}) = \mathcal{L}_{MD}(\underline{u}, \underline{\Delta})$$

$$+ \frac{1}{2\sigma} \sum_{n=1}^{N-1} \left\{ \|\llbracket \underline{u} \rrbracket_n\|_{\Omega}^2 + \|\llbracket \partial_t \underline{u} \rrbracket_n\|_{H^1}^2 \right\}$$

↑  $l \rightarrow \infty$

JOR/SOR-like algorithms  $\iff$  preconditioner

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↑  $l \rightarrow \infty$  ↓

$M_{\sigma}^{-1}$

JOR/SOR-like algorithms  $\iff$  preconditioner

# A RECIPE FOR THE DISCRETE LEVEL

## Continuous level

$$u \in Z, \lambda \in Y$$

$$\mathcal{L}(u, \lambda) = \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle$$

DECOMPOSE  
then  
OPTIMIZE

$$\underline{u} \in \underline{Z}^\circ, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}_{MD}(\underline{u}, \underline{\Delta}) = \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 \right.$$

$$\left. + \int_{I_n} \langle \square u_n, \lambda_n \rangle \right\}$$

$$\underline{u} \in \underline{Z}, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}^P(\underline{u}, \underline{\Delta}) = \mathcal{L}_{MD}(\underline{u}, \underline{\Delta})$$

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JOR/SOR-like  
algorithms

$\Leftrightarrow$  preconditioner

## Discrete level

$$u_h \in Z_h, \lambda_h \in Y_h$$

$$\mathcal{L}_h(u_h, \lambda_h) := \frac{1}{2} \|u_h - u_d\|_q^2 + \mathcal{S}(u_h, u_h) + b(u_h, \lambda_h) - \mathcal{S}^*(\lambda_h, \lambda_h)$$

$h \rightarrow 0$   
 $\Leftrightarrow$

# A RECIPE FOR THE DISCRETE LEVEL

## Continuous level

$$u \in Z, \lambda \in Y$$

$$\mathcal{L}(u, \lambda) = \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle$$

↑  
DECOMPOSE  
then  
OPTIMIZE

$$\underline{u} \in \underline{Z}^\circ, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}_{MD}(\underline{u}, \underline{\Delta}) = \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 \right.$$

$$\left. + \int_{I_n} \langle \square u_n, \lambda_n \rangle \right\}$$

↑  $\sigma \rightarrow 0$

$$\underline{u} \in \underline{Z}, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}^P(\underline{u}, \underline{\Delta}) = \mathcal{L}_{MD}(\underline{u}, \underline{\Delta})$$

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↑  $l \rightarrow \infty$

JOR/SOR-like  
algorithms

↔ preconditioner

$h \rightarrow 0$   
↔

## Discrete level

$$u_h \in Z_h, \lambda_h \in Y_h$$

stabilisation  
term for  $u_h$

$$\mathcal{L}_h(u_h, \lambda_h) := \frac{1}{2} \|u_h - u_d\|_q^2 + \boxed{\mathcal{S}(u_h, u_h)}$$

$$+ \boxed{b(u_h, \lambda_h)} - \boxed{\mathcal{S}^*(\lambda_h, \lambda_h)}$$

constraint    stabilisation  
term for  $\lambda_h$

# A RECIPE FOR THE DISCRETE LEVEL

## Continuous level

$$u \in Z, \lambda \in Y$$

$$\mathcal{L}(u, \lambda) = \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle$$

↑  
DECOMPOSE  
then  
OPTIMIZE  
↓

$$\underline{u} \in \underline{Z}^\circ, \underline{\Delta} \in \underline{Y}$$

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↑  
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↑  
 $l \rightarrow \infty$   
↓

JOR/SOR-like  
algorithms

↔ preconditioner

## Discrete level

$$u_h \in Z_h, \lambda_h \in Y_h$$

$$\mathcal{L}_h(u_h, \lambda_h) := \frac{1}{2} \|u_h - u_d\|_q^2 + S(u_h, u_h) + b(u_h, \lambda_h) - S^*(\lambda_h, \lambda_h)$$

↑  
DECOMPOSE  
then  
OPTIMIZE  
↓

$$\underline{u}_h \in \underline{Z}_h^\circ, \underline{\Delta}_h \in \underline{Y}_h$$

$$\mathcal{L}_h^{MD}(\underline{u}_h, \underline{\Delta}_h) := \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 + S_n(u_h, u_h) + b_n(u_h, \lambda_h) - S_n^*(\lambda_h, \lambda_h) \right\}$$

# A RECIPE FOR THE DISCRETE LEVEL

## Continuous level

$$u \in Z, \lambda \in Y$$

$$\mathcal{L}(u, \lambda) = \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle$$

DECOMPOSE  
then  
OPTIMIZE

$$\underline{u} \in \underline{Z}^\circ, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}_{MD}(\underline{u}, \underline{\Delta}) = \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 + \int_{I_n} \langle \square u_n, \lambda_n \rangle \right\}$$

$\sigma \rightarrow 0$

$$\underline{u} \in \underline{Z}, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}^p(\underline{u}, \underline{\Delta}) = \mathcal{L}_{MD}(\underline{u}, \underline{\Delta})$$

$$+ \frac{1}{2\sigma} \sum_{n=1}^{N-1} \left\{ \|\llbracket \underline{u} \rrbracket_n\|_{\Omega}^2 + \|\llbracket \partial_t \underline{u} \rrbracket_n\|_{H^1}^2 \right\}$$

JOR/SOR-like  
algorithms

$\Leftrightarrow$  preconditioner

$h \rightarrow 0$   
 $\Leftrightarrow$

## Discrete level

$$u_h \in Z_h, \lambda_h \in Y_h$$

$$\mathcal{L}_h(u_h, \lambda_h) := \frac{1}{2} \|u_h - u_d\|_q^2 + S(u_h, u_h) + b(u_h, \lambda_h) - S^*(\lambda_h, \lambda_h)$$

DECOMPOSE  
then  
OPTIMIZE

$$\underline{u}_h \in \underline{Z}_h^\circ, \underline{\Delta}_h \in \underline{Y}_h$$

$$\mathcal{L}_h^{MD}(\underline{u}_h, \underline{\Delta}_h) := \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 + S_n(u_h, u_h) + b_n(u_h, \lambda_h) - S_n^*(\lambda_h, \lambda_h) \right\}$$

regularity  
+  
continuity conditions

# A RECIPE FOR THE DISCRETE LEVEL

## Continuous level

$$u \in Z, \lambda \in Y$$

$$\mathcal{L}(u, \lambda) = \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle$$

DECOMPOSE  
then  
OPTIMIZE

$$\underline{u} \in \underline{Z}^\circ, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}_{MD}(\underline{u}, \underline{\Delta}) = \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 \right.$$

$$\left. + \int_{I_n} \langle \square u_n, \lambda_n \rangle \right\}$$

$\sigma \rightarrow 0$

$$\underline{u} \in \underline{Z}, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}^P(\underline{u}, \underline{\Delta}) = \mathcal{L}_{MD}(\underline{u}, \underline{\Delta})$$

$$+ \frac{1}{2\sigma} \sum_{n=1}^{N-1} \left\{ \|\llbracket \underline{u} \rrbracket_n\|_{\Omega}^2 + \|\llbracket a_t \underline{u} \rrbracket_n\|_{H^1}^2 \right\}$$

JOR/SOR-like  
algorithms

$\iff$  preconditioner

$h \rightarrow 0$   
 $\iff$

## Discrete level

$$u_h \in Z_h, \lambda_h \in Y_h$$

$$\mathcal{L}_h(u_h, \lambda_h) := \frac{1}{2} \|u_h - u_d\|_q^2 + S(u_h, u_h) + b(u_h, \lambda_h) - S^*(\lambda_h, \lambda_h)$$

DECOMPOSE  
then  
OPTIMIZE

$$\underline{u}_h \in \underline{Z}_h^\circ, \underline{\Delta}_h \in \underline{Y}_h$$

$$\mathcal{L}_h^{MD}(\underline{u}_h, \underline{\Delta}_h) := \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 + S_n(u_h, u_h) + b_n(u_h, \lambda_h) - S_n^*(\lambda_h, \lambda_h) \right\}$$

$\sigma \rightarrow 0$

$$\underline{u}_h \in \underline{Z}_h, \underline{\Delta}_h \in \underline{Y}_h$$

$$\mathcal{L}_h(\underline{u}_h, \underline{\Delta}_h) := \mathcal{L}_h^{MD} + \boxed{S_\sigma^{\downarrow\uparrow}(\underline{u}_h, \underline{u}_h)}$$

penalisation of  
the continuity  
conditions

# A RECIPE FOR THE DISCRETE LEVEL

## Continuous level

$$u \in Z, \lambda \in Y$$

$$\mathcal{L}(u, \lambda) = \frac{1}{2} \|u - u_d\|_q^2 + \int_0^T \langle \square u, \lambda \rangle$$

↑  
DECOMPOSE  
then  
OPTIMIZE

$$\underline{u} \in \underline{Z}^\circ, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}_{MD}(\underline{u}, \underline{\Delta}) = \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 \right.$$

$$\left. + \int_{I_n} \langle \square u_n, \lambda_n \rangle \right\}$$

↑  
 $\sigma \rightarrow 0$   
↓

$$\underline{u} \in \underline{Z}, \underline{\Delta} \in \underline{Y}$$

$$\mathcal{L}^P(\underline{u}, \underline{\Delta}) = \mathcal{L}_{MD}(\underline{u}, \underline{\Delta})$$

$$+ \frac{1}{2\sigma} \sum_{n=1}^{N-1} \left\{ \|\llbracket \underline{u} \rrbracket_n\|_{\Omega}^2 + \|\llbracket \partial_t \underline{u} \rrbracket_n\|_{H^1}^2 \right\}$$

↑  
 $l \rightarrow \infty$   
↓  
JOR/SOR-like  
algorithms

↔ preconditioner

## Discrete level

$$u_h \in Z_h, \lambda_h \in Y_h$$

$$\mathcal{L}_h(u_h, \lambda_h) := \frac{1}{2} \|u_h - u_d\|_q^2 + \mathcal{S}(u_h, u_h) + b(u_h, \lambda_h) - \mathcal{S}^*(\lambda_h, \lambda_h)$$

↑  
DECOMPOSE  
then  
OPTIMIZE  
↓

$$\underline{u}_h \in \underline{Z}_h^\circ, \underline{\Delta}_h \in \underline{Y}_h$$

$$\mathcal{L}_h^{MD}(\underline{u}_h, \underline{\Delta}_h) := \sum_{n=0}^{N-1} \left\{ \frac{1}{2} \|u_n - u_{d,n}\|_{q_n}^2 + \mathcal{S}_n(u_h, u_h) + b_n(u_h, \lambda_h) - \mathcal{S}_n^*(\lambda_h, \lambda_h) \right\}$$

↑  
 $\sigma \rightarrow 0$   
↓

$$\underline{u}_h \in \underline{Z}_h, \underline{\Delta}_h \in \underline{Y}_h$$

$$\mathcal{L}_h(\underline{u}_h, \underline{\Delta}_h) := \mathcal{L}_h^{MD} + \mathcal{S}_\sigma^{\uparrow\downarrow}(\underline{u}_h, \underline{u}_h)$$

↑  
 $l \rightarrow \infty$   
↓

preconditioner ↔ JOR/SOR-like algorithms



# EXAMPLES OF DISCRETIZATIONS

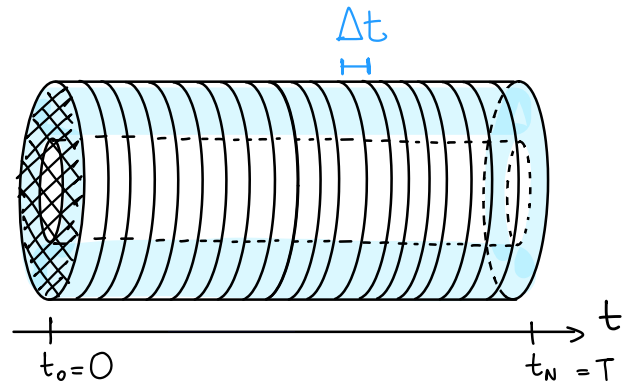
- ⇒ [Cîndea, Münch. 2015] Conforming space-time discretisation  $Z_h \subset Z, Y_h \subset Y$ , using  $C^1$ -conforming elements
- ⇒ [Burman et al. 2020] Non-conforming discretisation using FEM in space + finite differences in time
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Jumps at the time interfaces are already penalised

$$S_{\Delta t}^{\text{st}}(\underline{u}_h, \underline{v}_h) = \sum_{n=1}^{N-1} \left\{ \frac{1}{2\Delta t} ([\underline{u}]_n, [\underline{u}]_n)_{\Omega} + \frac{\Delta t}{2} (\nabla [\underline{u}]_n, \nabla [\underline{u}]_n)_{\Omega} + \frac{1}{2\Delta t} ([\partial_t \underline{u}]_n, [\partial_t \underline{u}]_n)_{\Omega} \right\}$$

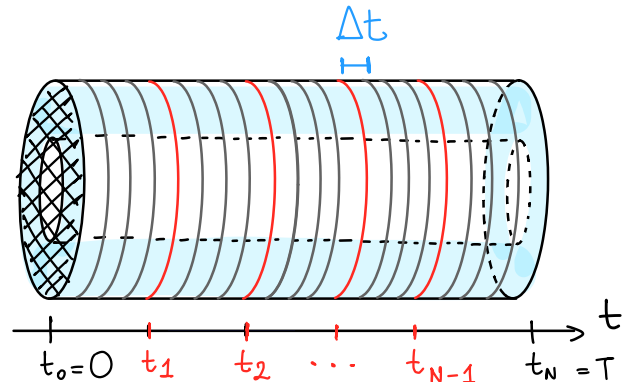


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Adding a domain decomposition framework

$$S_{\sigma}^{\mathcal{V}}(\underline{u}_h, \underline{v}_h) = \sum_{n=1}^{N-1} \left\{ \frac{1}{2(\Delta t + \sigma)} (\llbracket \underline{u} \rrbracket_n, \llbracket \underline{u} \rrbracket_n)_{\Omega} + \frac{(\Delta t + \sigma)}{2} (\nabla \llbracket \underline{u} \rrbracket_n, \nabla \llbracket \underline{u} \rrbracket_n)_{\Omega} + \frac{1}{2(\Delta t + \sigma)} (\llbracket \partial_t \underline{u} \rrbracket_n, \llbracket \partial_t \underline{u} \rrbracket_n)_{\Omega} \right\}$$



# NUMERICAL EXPERIMENTS

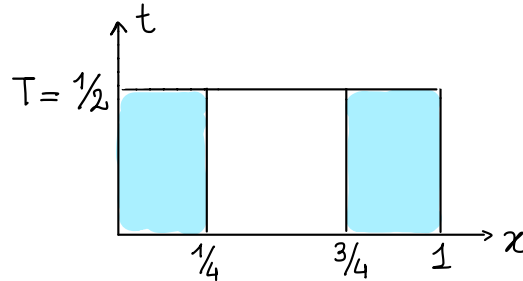
1D+time test case

● 9

$$\Omega = [0, 1]$$

$$\omega = [0, 1/4] \cup [3/4, 1]$$

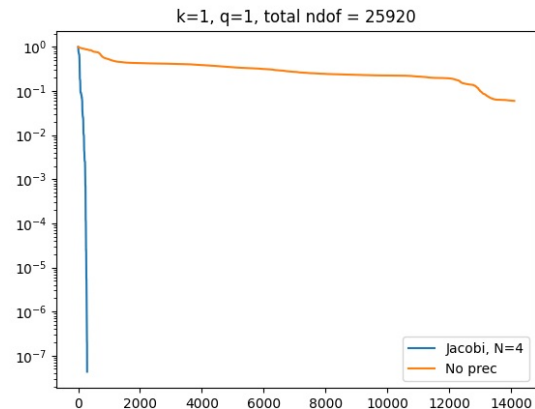
$$C = 1$$



We consider the discretisation from [Burman, Preuss, 2025].  
And we use the preconditioner from the fixed point  
in GMRES

⇒ Unpreconditioned GMRES  
cannot be applied

- space: FEM order 1
- time: DG order 1
- total ndofs: 25920



# NUMERICAL EXPERIMENTS

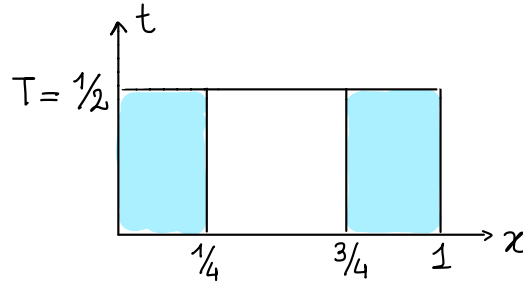
1D+time test case

●  $q$

$$\Omega = [0, 1]$$

$$\omega = [0, 1/4] \cup [3/4, 1]$$

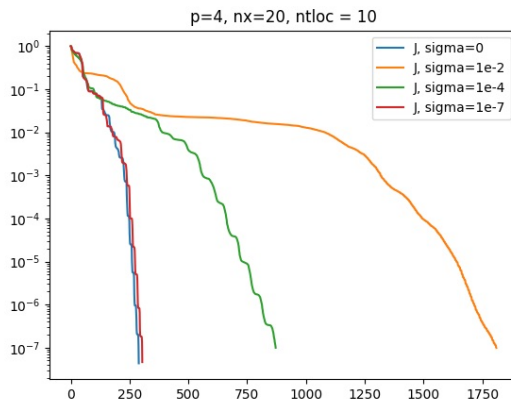
$$C = 1$$



We consider the discretisation from [Burman, Preuss, 2025].  
with  $S_{\sigma}^{\uparrow}$  at the subdomain interfaces

⇒ We observe numerically  
that the best choice  
is  $\sigma = 0$

- space: FEM order 1
- time: DG order 1
- total ndofs: 25920



# NUMERICAL EXPERIMENTS

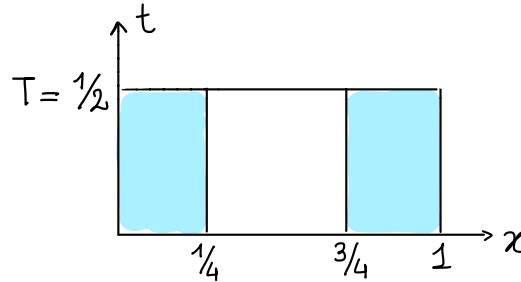
1D+time test case

● 9

$$\Omega = [0, 1]$$

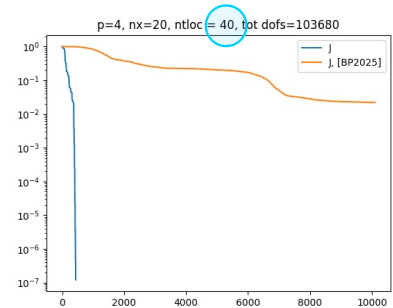
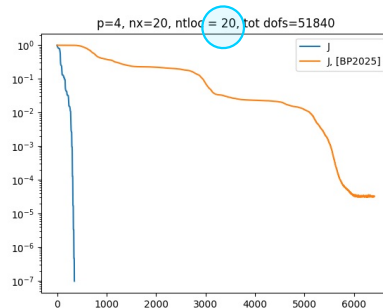
$$\omega = [0, 1/4] \cup [3/4, 1]$$

$$C = 1$$



We consider the discretisation from [Burman, Preuss, 2025].  
with  $S_{\sigma}^{\uparrow}$  at the subdomain interfaces with  $\sigma = 0$

⇒ We perform better than the monolithic preconditioners proposed in [Burman, Preuss, 2025]



# NUMERICAL EXPERIMENTS

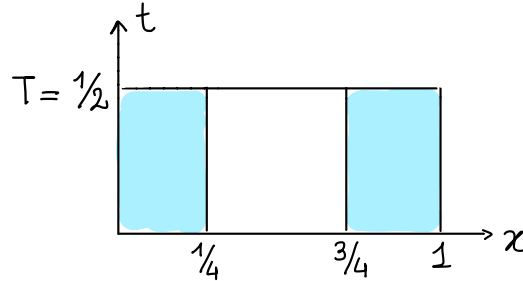
1D+time test case

● 9

$$\Omega = [0, 1]$$

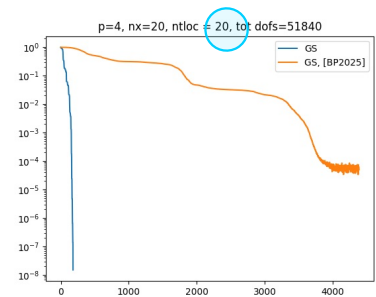
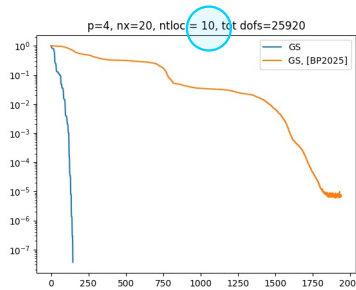
$$\omega = [0, 1/4] \cup [3/4, 1]$$

$$C = 1$$



We consider the discretisation from [Burman, Preuss, 2025].  
with  $S_{\sigma}^{\uparrow}$  at the subdomain interfaces with  $\sigma = 0$

⇒ We perform better  
than the monolithic  
preconditioners  
proposed in  
[Burman, Preuss, 2025]



# CONCLUSIONS & PROSPECTS

- ⇒ We proposed a decompose-then-optimize paradigm to define preconditioners, given some discretisation of a unique continuation problem subject to the wave equation.
- ⇒ The performance of the preconditioner depends on the penalisation parameter  $\sigma$

# CONCLUSIONS & PROSPECTS

- ⇒ We proposed a decompose-then-optimise paradigm to define preconditioners, given some discretisation of a unique continuation problem subject to the wave equation.
- ⇒ The performance of the preconditioner depends on the penalisation parameter  $\sigma$

## Future work

- ⇒ More complete numerical investigation on different discretisations
- ⇒ Introduce Lagrange multipliers for the continuity conditions instead of least-squares penalisation
- ⇒ Adding a decomposition with respect to space
- ⇒ Introducing a second level to speed up convergence and (hopefully) gain scalability

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# GEOMETRIC CONTROL CONDITION

[ Bardos, Lebeau, Rauch 1992 ]

We consider the case of stationary cylindrical observation domains

$$\omega(t) \equiv \omega, \quad q = \omega \times [0, T]$$

“Roughly speaking, it says that every geodesic propagating in  $\Omega$  at unit speed, and reflecting at the boundary according to the classical laws of geometrical optics, so-called generalized geodesics, should meet the open set  $\omega$  within time  $T$ .”

[ Le Rousseau et al. 2017 ]

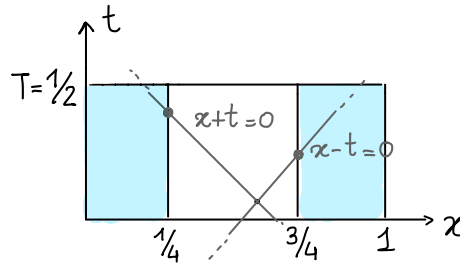
1D+time example

●  $q$

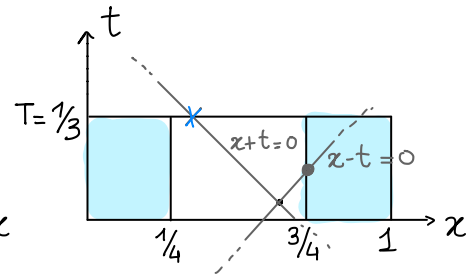
$$\Omega = [0, 1]$$

$$\omega = [0, 1/4] \cup [3/4, 1]$$

$$c = 1 \text{ m/s}$$



GCC verified



GCC not verified

## Assumption

The GCC holds for the triplet  $(\Omega, \omega, T)$ .

# OBSERVABILITY ESTIMATE

**Lemma** (Generalised observability inequality - GOI)

For  $\Omega$  with smooth boundary and  $(\Omega, \omega, T)$  satisfying the geometric control condition there exists a constant  $C = C(\Omega, \omega, T, c)$  such that  $\forall u \in Z$

$$\|u(0), \partial_t u(0)\|_{\mathbf{H}} \leq C \left( \|u\|_q + \|\square u\|_{L^2(0, T; H^{-1}(\Omega))} \right)$$

$\Rightarrow$  The proof uses a lifting argument from the classical result of [Bardos, Lebeau, Rauch 1992].

**Lemma** [Castro, Cindea, Münch, 2014]

The space  $Z$ , equipped with the scalar product

$$(u, v)_Z := (u, v)_q + \eta \int_0^T \langle \square u, \square v \rangle_{H^{-1}}$$

is an Hilbert space.