

# Domain decomposition methods for SPDEs

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# Outline

**Underlying problem**

**Operator splitting & Domain decomposition**

**Convergence result**

**Numerical example**

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# Underlying problem

**Goal: Efficient numerical approximation of an SPDE**

Abstract problem

$$\begin{cases} du(t) + A(t, u(t)) dt = F(t, u(t)) dt \\ \quad \quad \quad \quad \quad \quad \quad + B(t, u(t)) dW(t), & t \in (0, T], \\ u(0) = u_0. \end{cases}$$

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More concrete example

$$\begin{cases} du(t, x) - 0.1(1 + \exp(-t))\Delta u(t, x) dt \\ \quad \quad \quad = \sin(u(t, x)) dt + u(t, x) dW(t, x), & t \in (0, T], x \in (0, 1)^2, \\ u(t, x) = 0, & t \in (0, T], x \in \partial(0, 1)^2, \\ u(0, x) = u_0(x), & x \in (0, 1)^2. \end{cases}$$

# Notation

- ▶  $H$  is separable Hilbert space, e.g.  $H = L^2(\mathcal{D})$ ,
- ▶  $V$  is a separable, reflexive Banach space, e.g.  $V = H_0^1(\mathcal{D})$ ,
- ▶ the spaces built a Gelfand triple  $V \hookrightarrow H \cong H^* \hookrightarrow V^*$ ,
- ▶ initial data  $u_0: \Omega \rightarrow H$  belongs to  $L^2(\Omega, H)$ ,
- ▶  $Q$ -Wiener process  $W$  such that
  - ▶  $Q$  being positive definite and self-adjoint,
  - ▶  $Q^{\frac{1}{2}} \in \mathcal{HS}(H)$ .

## Assumptions on the data, $A$

Assume that the family of operators  $A(t, \cdot): V \rightarrow V^*$  fulfills:

- ▶ monotone,
- ▶ coercive,
- ▶  $\nu$ -Hölder continuous in the  $t$ -component for  $\nu \in (0, \frac{1}{2}]$ ,
- ▶ Lipschitz continuous in the second component.

Example:  $A(t, \cdot) = -\Delta: H_0^1 \rightarrow H^{-1}$ .

## Assumptions on the data, $F$ and $B$

Assume that the family of operators  $F(t, \cdot): H \rightarrow H$ ,  
 $B(t, \cdot): H \rightarrow \mathcal{HS}(Q^{\frac{1}{2}}(H), H)$  fulfill:

- ▶  $\|F(t, 0)\|_H$  and  $\|B(t, 0)\|_{L_0^2}$  are bounded,
- ▶  $\nu$ -Hölder continuous in the first component for  $\nu \in (0, \frac{1}{2}]$ ,
- ▶ Lipschitz continuous in the second component.

Example:  $F(t, \cdot): L^2(\mathcal{D}) \rightarrow L^2(\mathcal{D})$ ,  $F(t, u) = \sqrt{t} \sin(u)$









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# Literature

## Operator splittings:

- ▶ Reaction-Diffusion problems: [Arrarás, in 't Hout, Hundsdorfer, Portero, 2017], [Hansen, Stillfjord, 2012], [Koch, Neuhauser, Thalhammer 2013], ...
- ▶ S(P)DEs: [Ableidinger, Buckwar, 2016], [Buckwar, et. al. 2022], [Bréhier, Goudenège, 2019], ...
- ▶ ...

## Domain decomposition:

- ▶ Deterministic PDEs: [Arrarás, Portero, 2015], [E., Hansen, 2018, 2022], [Mathew, Polyakov, Russo, Wang, 1998], [Vabishechevich, 2008], ...
- ▶ Random PDEs: [Mu Wang, 2019], [Sarkar, Benabbou, Ghanem, 2008].

# Advantages domain decomposition

## Advantages

- ▶ Cheaper sub-problems
- ▶ Problems can be implemented in parallel

## Plan

- ▶ So far: Domain decomposition for
  - ▶ deterministic time dependent PDEs,
  - ▶ random stationary PDEs
- ▶ Extend to time dependent SPDEs

# Operator splitting

Consider the stochastic evolution equation

$$du(t) + A(t, u(t)) dt = B(t, u(t)) dW(t).$$

# Operator splitting

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$$\begin{aligned} du(t) + (A_1(t, u(t)) + A_2(t, u(t))) dt \\ = (B_1(t, u(t)) + B_2(t, u(t))) dW(t). \end{aligned}$$

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Then we solve

$$du(t) + A_1(t, u(t)) dt = B_1(t, u(t)) dW(t)$$

and

$$du(t) + A_2(t, u(t)) dt = B_2(t, u(t)) dW(t)$$

individually and “glue” them together to a suitable solution.

## Different splittings

For

$$\begin{cases} du(t) + A(t, u(t)) dt = B(t, u(t)) dW(t), & t \in (0, T], \\ u(0) = u_0, \end{cases}$$

we consider the solution operator  $\mathcal{L}_{A,B}$  such that

$$\mathcal{L}_{A,B} u_0 = \mathcal{L}_{A_1+A_2, B_1+B_2} u_0 = u$$

- ▶ Lie splitting

$$\mathcal{L}_{A,B} \approx \mathcal{L}_{A_2, B_2} \circ \mathcal{L}_{A_1, B_1}$$

- ▶ Sum splitting

$$\mathcal{L}_{A,B} \approx \frac{1}{2} \left( \mathcal{L}_{2A_2, 2B_2} + \mathcal{L}_{2A_1, 2B_1} \right)$$

- ▶ Strang splitting

$$\mathcal{L}_{A,B} \approx \mathcal{L}_{\frac{1}{2}A_1, \frac{1}{2}B_1} \circ \mathcal{L}_{A_2, B_2} \circ \mathcal{L}_{\frac{1}{2}A_1, \frac{1}{2}B_1}$$

# Domain decomposition integrators

## Example: Stochastic heat equation

For

- ▶  $T > 0$ ,
- ▶ open, connected and bounded set  $\mathcal{D} \subseteq \mathbf{R}^d$ ,

consider the problem

$$\begin{cases} du(t, x) - \Delta u(t, x) dt = B(u) dW(t, x), & (t, x) \in (0, T) \times \mathcal{D}, \\ u(t, x) = 0, & (t, x) \in (0, T) \times \partial\mathcal{D}, \\ u(0, x) = u_0(x), & x \in \mathcal{D}. \end{cases}$$

## Suitable operators: Original problem

We find the weak formulation through

$$\int_{\mathcal{D}} duv \, dx - \int_{\mathcal{D}} (\nabla \cdot \nabla u) v \, dx \, dt = \int_{\mathcal{D}} B(u)v \, dx \, dW$$

for  $v \in C_0^\infty(\mathcal{D})$ .

## Suitable operators: Original problem

We find the weak formulation through

$$\int_{\mathcal{D}} duv \, dx + \int_{\mathcal{D}} \nabla u \cdot \nabla v \, dx \, dt = \int_{\mathcal{D}} B(u)v \, dx \, dW$$

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for  $v \in C_0^\infty(\mathcal{D})$ . For  $H = L^2(\mathcal{D})$  and  $V = H_0^1(\mathcal{D})$ , we introduce  $A: V \rightarrow V^*$  and  $B(u) \, dW \in H$  as

$$\begin{aligned} \langle Au, v \rangle_{V^* \times V} &= \int_{\mathcal{D}} \nabla u \cdot \nabla v \, dx, \quad u, v \in V, \\ (B(u) \, dW, v) &= \int_{\mathcal{D}} B(u)v \, dx \, dW, \quad u, v \in H. \end{aligned}$$

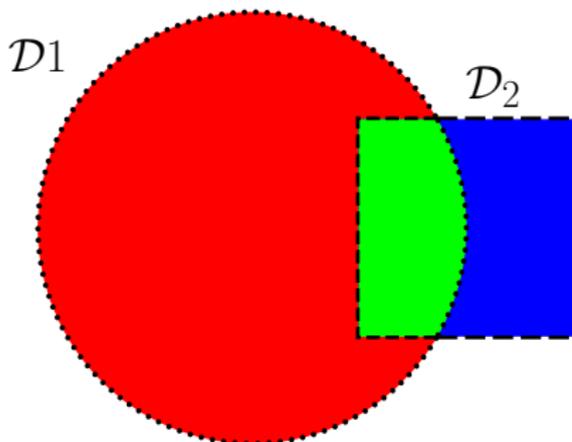
Then the SPDE can be written as an abstract equation

$$\begin{cases} du(t) + Au(t) \, dt = B(u(t)) \, dW(t) & \text{in } V^*, \quad t \in (0, T], \\ u(0) = u_0 & \text{in } H. \end{cases}$$

# Decomposition of unity

- ▶ We consider the overlapping subdomains  $\mathcal{D}_\ell$ ,  $\ell = 1, \dots, s$ , with  $\bigcup_{\ell=1}^s \mathcal{D}_\ell = \mathcal{D}$ .
- ▶ We introduce a partition of unity  $(\chi_\ell)_{\ell=1, \dots, s}$  in  $W^{1, \infty}(\mathcal{D})$ .
- ▶ We assume that

$$\sum_{\ell=1}^s \chi_\ell = 1, \quad \text{supp}(\chi_\ell) \subseteq \mathcal{D}_\ell, \quad \chi_\ell(x) \in (0, 1] \text{ for } x \in \mathcal{D}_\ell.$$



## Suitable operators: Original problem

We find the weak formulation through

$$\int_{\mathcal{D}} duv \, dx + \int_{\mathcal{D}} \nabla u \cdot \nabla v \, dt \, dx = \int_{\mathcal{D}} (B(u) \, dW)v \, dx$$

for  $v \in C_0^\infty(\mathcal{D})$ . For  $H = L^2(\mathcal{D})$  and  $V = H_0^1(\mathcal{D})$ , we introduce  $A: V \rightarrow V^*$  and  $B(u) \, dW \in H$  as

$$\begin{aligned} \langle Au, v \rangle_{V^* \times V} &= \int_{\mathcal{D}} \nabla u \cdot \nabla v \, dx, \quad u, v \in V, \\ (B(u) \, dW, v) &= \int_{\mathcal{D}} (B(u) \, dW)v \, dx, \quad v \in V. \end{aligned}$$

Then the SPDE can be written as an abstract equation

$$\begin{cases} du(t) + Au(t) \, dt = B(u(t)) \, dW(t) & \text{in } V^*, \quad t \in (0, T], \\ u(0) = u_0 & \text{in } H. \end{cases}$$

## Suitable operators: Decomposed problem

We find the weak formulation through

$$\int_{\mathcal{D}} duv \, dx + \int_{\mathcal{D}_\ell} \chi_\ell \nabla u \cdot \nabla v \, dx \, dt = \int_{\mathcal{D}_\ell} \chi_\ell B(u)v \, dx \, dW.$$

for  $v \in C_0^\infty(\mathcal{D})$ . For  $H = L^2(\mathcal{D})$  and  $V_\ell \supset H_0^1(\mathcal{D})$ , we introduce  $A_\ell: V_\ell \rightarrow V_\ell^*$  and  $B_\ell(u) \, dW \in H$  as

$$\langle A_\ell u, v \rangle_{V_\ell^* \times V_\ell} = \int_{\mathcal{D}_\ell} \chi_\ell \nabla u \cdot \nabla v \, dx, \quad u, v \in V_\ell,$$

$$(B_\ell(u) \, dW, v) = \int_{\mathcal{D}_\ell} \chi_\ell B(u)v \, dx \, dW, \quad u, v \in H.$$

Then the SPDE can be written as an abstract equation

$$\begin{cases} du(t) + A_\ell u(t) \, dt = B_\ell(u(t)) \, dW(t) & \text{in } V_\ell^*, \quad t \in (0, T], \\ u(0) = u_0 & \text{in } H. \end{cases}$$









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# Explicit error bounds

Assume that

- ▶  $u \in C^\nu([0, T]; L^2(V, \Omega))$ ,  $\nu \in (0, 1/2)$ ,
- ▶ regularity assumption on the decomposition,
- ▶ eigenvalues of  $Q$  fulfill:  $\sum_{k>K} q_k < \varepsilon$ .

Then the error  $e^n = u(t_n) - U^n$  can be bound by

$$\begin{aligned} \mathbf{E}[\|e^n\|_H^2] &+ \sum_{i=1}^n \sum_{\ell=1}^s \mathbf{E}[\|e_\ell^i - e_{\ell-1}^i\|_H^2] + \sum_{i=1}^n h \sum_{\ell=1}^s \mathbf{E}[\|e_\ell^i\|_{V_\ell}^2] \\ &\leq C(h^{2\nu} + \varepsilon). \end{aligned}$$

Note:

Same error rate as in [Mukam, Tambue, 2019], [Kruse, Weiske, 2021].

## Regularity of the solution

The mild solution of the semi-linear problem fulfills

$$\sup_{t_1, t_2 \in [0, T], t_1 \neq t_2} \frac{(\mathbf{E} [\|u(t_1) - u(t_2)\|_{H_0^1}^2])^{\frac{1}{2}}}{|t_1 - t_2|^{\min(\frac{1}{2}, \frac{r}{2})}} < \infty$$

if additionally

- ▶ coefficients are time-independent ,
- ▶  $-A$  is the generator of an analytic semigroup of contractions (in particular linear),
- ▶ initial value lies in  $L^2(\Omega, H^2(\mathcal{D}) \cap H_0^1(\mathcal{D}))$ ,
- ▶ Sufficient regularity in diffusion term:  
 $\|A^{\frac{r}{2}} B(x)\|_{\mathcal{HS}(H)} \leq C(1 + \|x\|_{H_0^r})$

Compare [Kruse, Larsson, 2012].

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## Numerical example 1

For the domain  $\mathcal{D} = (0, 1)^2 \subset \mathbf{R}^2$ , we consider the SPDE

$$\begin{cases} du(t, x) - D(t)\Delta u(t, x) dt = F(t, x, u) dt + B(t, x, u) dW(t, x), & t \in (0, 1], x \in \mathcal{D}, \\ u(t, x) = 0, & t \in [0, 1], x \in \partial\mathcal{D}, \\ u(0, x) = u_0(x), & x \in \mathcal{D}. \end{cases}$$

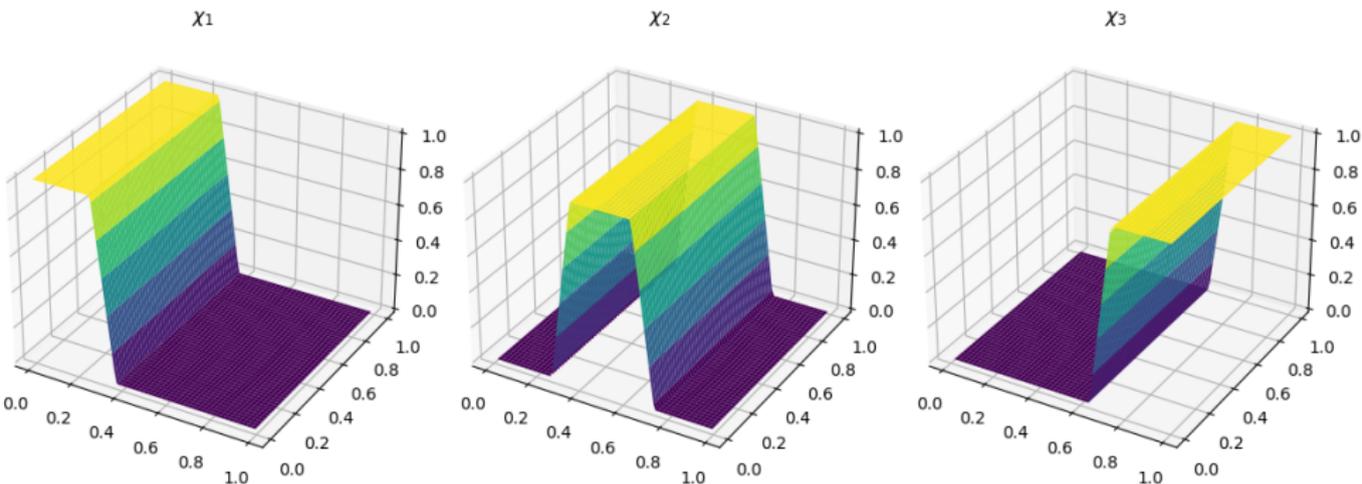
where

$$u_0(x) = 5x_1^2(x_1 - 1)^3x_2^2(x_2 - 1)^3, \quad D(t) = 0.1(1 + \exp(-t)) \\ F(t, x, u) = \sin(u), \quad B(t, x, u) = 1.$$

This means, we have a semi-linear equation with additive noise.

# Weight functions for domain decomposition

For the domain decomposition, we divide the domain into three part. For this we use the weight functions:



# Parameters for experiment

For the test we use the following parameters:

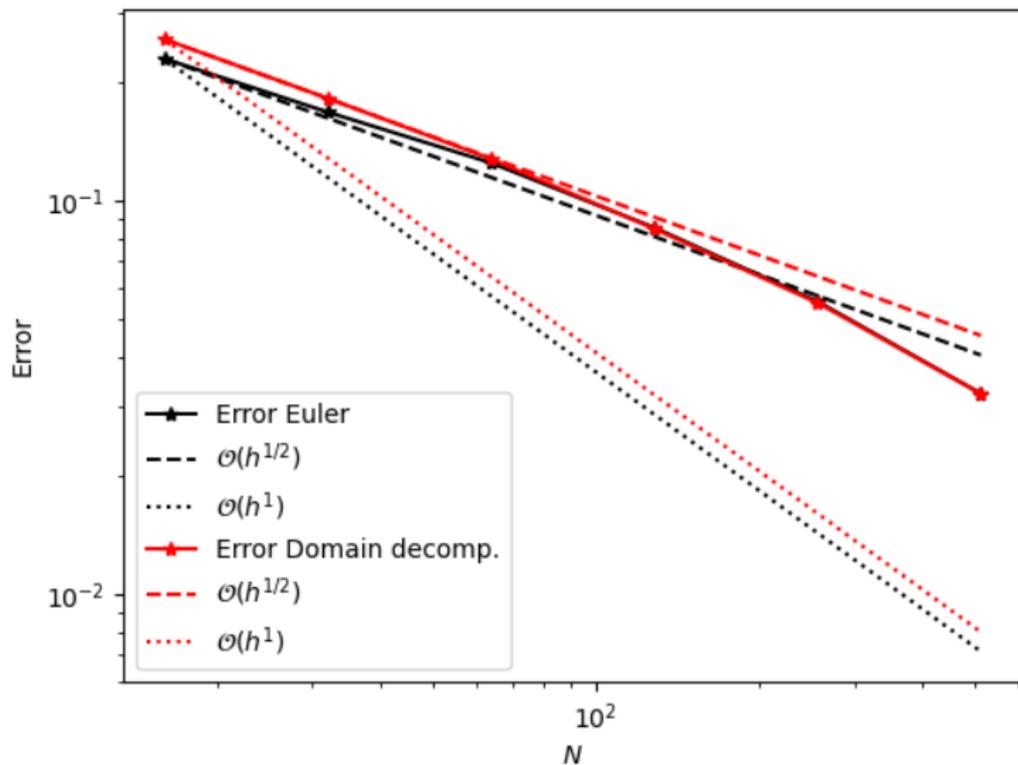
- ▶ Use weight functions  $\chi_1$ ,  $\chi_2$  and  $\chi_3$  from previous slide.
- ▶ Monte Carlo iterations: 10.
- ▶ For a family of Wiener processes  $(\beta_{i,j})_{i,j}$ , we use the truncated noise

$$W^K(t, x) = \sum_{i,j=1}^K (i+j)^{-\frac{1}{2}(2r+1+\varepsilon)} \beta_{i,j}(t) e_{i,j}(x)$$

with regularity and truncation parameters  $r = 0.1$  and  $K = 100$ .

- ▶ Spatial discretization parameter  $h_{\text{space}} = 1/100$ .

# Error plot



## Numerical example 2

For the domain  $\mathcal{D} = (0, 1)^2 \subset \mathbf{R}^2$ , we consider the SPDE

$$\begin{cases} du(t, x) - D(t)\Delta u(t, x) dt = F(t, x, u) dt + B(t, x, u) dW(t, x), & t \in (0, 1], x \in \mathcal{D}, \\ u(t, x) = 0, & t \in [0, 1], x \in \partial\mathcal{D}, \\ u(0, x) = u_0(x), & x \in \mathcal{D}. \end{cases}$$

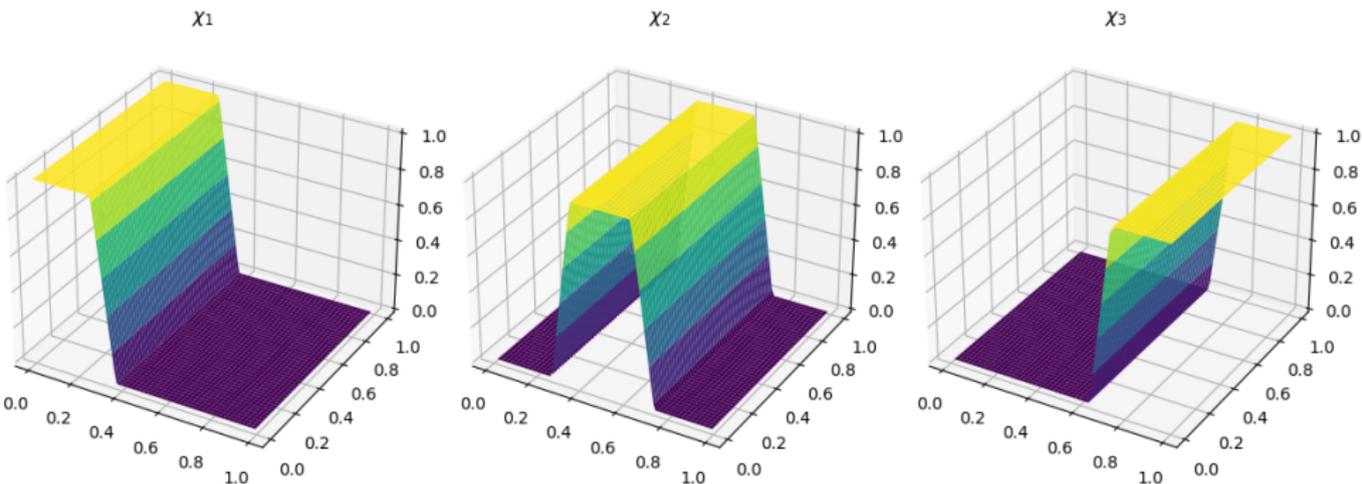
where

$$\begin{aligned} u_0(x) &= 5x_1^2(x_1 - 1)^3 x_2^2(x_2 - 1)^3, & D(t) &= 0.1 \\ F(t, x, u) &= (3 \sin(5x_1) + 2 \cos(7x_2))(\cos(t) + \sin(4t)) \\ B(t, x, u) &= u. \end{aligned}$$

This means, we have a linear equation with multiplicative noise.

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For the domain decomposition, we divide the domain into three part. For this we use the weight functions:



# Parameters for experiment

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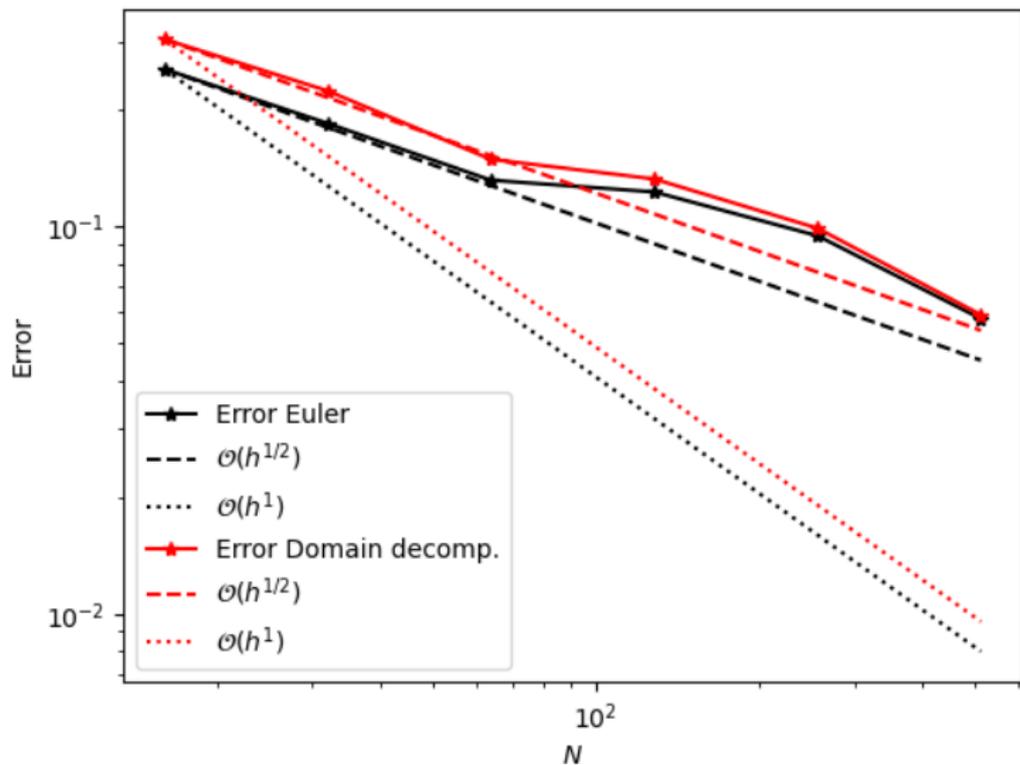
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**Thank you for your  
attention!**