

# Multi-Domain Solutions of PDEs Posed on Perforated Domains

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with: M. Boutilier and K. Brenner

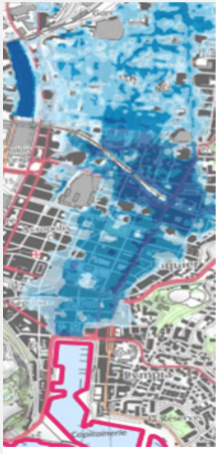
Preconditioning, Atlanta, 11 June 2024

## Motivation and model problem

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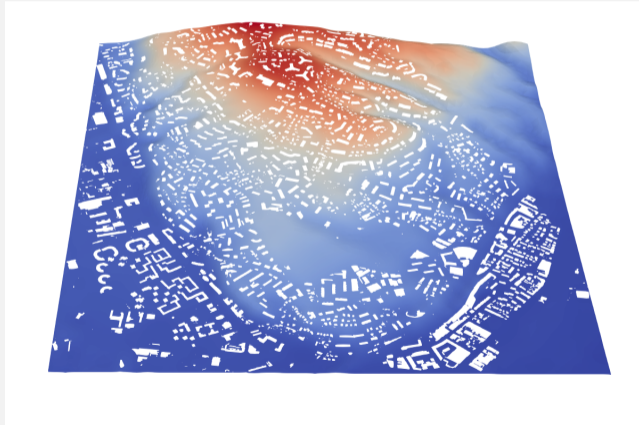
## Motivation: Urban Flood Modeling

- Efficiently solve problems on perforated domains.
- Expect corner singularities, triangle of varying magnitude, many degrees of freedom.



## Modelling a Realistic Problem

- Realistic topography ( $z_b(x, y)$ ) of Nice, France);
- Rainfall data (source term): Can be taken from previous flood events (rain gauge data);
- Flood maps: previous areas of flood risk.



# Nonlinear Problem: Diffusive Wave model and discretisation

## A non linear problem

$$\begin{cases} \partial_t u + \operatorname{div} \mathcal{F}(x, u, \nabla u) & = f, \text{ in } \Omega, \\ \mathcal{F}(x, u, \nabla u) \cdot \mathbf{n} & = 0, \text{ on } \partial\Omega \cap \partial\Omega_S, \\ u & = g, \text{ on } \partial\Omega \setminus \partial\Omega_S. \end{cases}$$

$$\mathcal{F}(x, u, \nabla u) = c_f \frac{h(u, z_b(\mathbf{x}))^\alpha}{\|\nabla u\|^{1-\gamma}} \nabla u,$$

- $z_b(\mathbf{x})$ : Bathymetry;
- $h(u, z_b(\mathbf{x})) = \max(u - z_b(\mathbf{x}), 0)$ : Water depth;
- $\alpha > 1, 0 < \gamma \leq 1$ .
- $c_f > 1$ : friction coefficient.

## Discretisation

Discretisation in space and time:

$$F(\mathbf{u}) := \frac{P}{\Delta t} (\mathbf{u} - \mathbf{u}^{\text{old}}) + K(\mathbf{u}) = 0, \quad (1)$$

where  $P$  is the (lumped) mass-matrix.

- Backward-Euler for time discretisation;
- $K(\mathbf{u})$  is discretisation of nonlinear term (FEM/FVM) and source term;
- Perform upwinding on  $h(u, z_b(\mathbf{x}))^\alpha$  term (due to degeneracy);
- Adaptive time-stepping may be necessary for Newton's method on this system.

## The purpose of this work

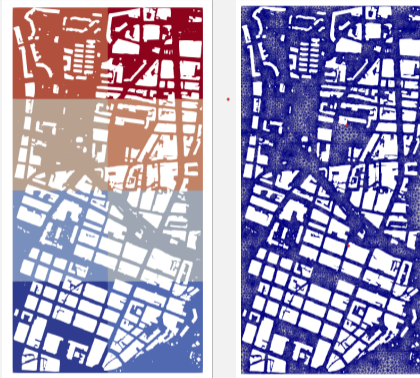
- Design a fast solution method for the **nonlinear multiscale** problem  $F(\mathbf{u}) = 0$ .
- Simulate the whole time-dependent problem.

## **Overlapping Schwarz: linear and nonlinear preconditioners**

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## Divide and conquer

- Partition of domain  $\Omega$  into subdomains  $\{\Omega_j\}_{j=1}^N$ ;
- Two levels of discretisation: 'Coarse' and 'Fine';
- Local subdomain solves can be done in parallel;
- Schwarz methods: use overlapping subdomains.



**Idea:** Solve model problem on each subdomain locally, with boundary conditions taken from adjacent subdomains.

# Linear preconditioning: Newton-Krylov-Schwarz (NKS)

## Newton's method to solve $F(\mathbf{u}) = 0$

Given initial  $\mathbf{u}^0$ , for outer iteration  $n = 0, \dots$ , to convergence,

- Solve for  $\delta^n : \nabla F(\mathbf{u}^n)\delta^n = F(\mathbf{u}^n)$ ,
- Update  $\mathbf{u}^{n+1} = \mathbf{u}^n - \delta^n$ .

At each Newton's iteration we need to solve a linear system:

$$J_n \delta_n = F_n$$

with  $J_n = \nabla F(\mathbf{u}^n)$ ,  $F_n = F(\mathbf{u}^n)$ .



X.-C. Cai, W. D. Gropp, D. E. Keyes, and M. D. Tidriri, Newton-Krylov-Schwarz methods in CFD, 1994

## DD preconditioning

Solve the preconditioned system

$$M^{-1}J_n\delta_n = M^{-1}F_n,$$

by a Krylov method, for some domain decomposition preconditioner  $M^{-1}$ .

- Does not change convergence/robustness of Newton's method.
- $M^{-1} \approx J_n^{-1} \rightarrow$  improves convergence of linear Krylov solver.



## Goal

Instead of  $F(\mathbf{u}) = 0$ , solve  $N(F(\mathbf{u})) = 0$  via Newton.

- $N(\mathbf{v}) = 0 \rightarrow \mathbf{v} = 0$ ;
- $N(F(\mathbf{v}))$  straightforward to compute.

## Use a fixed point iteration

$$\mathbf{u}^{n+1} = P(\mathbf{u}^n), \quad (2)$$

solve  $\mathcal{F}(\mathbf{u}) = P(\mathbf{u}) - \mathbf{u} = 0 \rightsquigarrow \mathcal{F}(\mathbf{u}) = 0$  is the preconditioned nonlinear system.

Nonlinear preconditioning can **improve convergence/robustness** of Newton's method and **localise** difficult nonlinearities.



X.-C. Cai and D. E. Keyes, *Nonlinearly preconditioned inexact newton algorithms*, SISC (2002)

# Nonlinear Restricted Additive Schwarz (NRAS) : $F(\mathbf{u}) = 0$

## Idea: use nonlinear Schwarz

- Decomposition into subdomains  $\Omega_j$
- Start from an initial guess  $\mathbf{u}^0$
- Perform local nonlinear subdomain solves

$$R_j F(R_j^T G_j(\mathbf{u}^n) + (I - R_j^T R_j)\mathbf{u}^n) = 0.$$

where  $R_j$  are restriction operators and  $D_j$  partition of unity matrices.

- “Glue” together local solutions  $G_j(\mathbf{u}^n)$

$$\mathbf{u}^{n+1} = \sum_j R_j^T D_j G_j(\mathbf{u}^n)$$

## Advantages

- Local subproblems are solved via Newton with negligible cost;
- Local solves can be done in parallel.
- Natural nonlinear solver based on the decomposition into subdomains.

**Downsides:** this is the non-linear equivalent of the iterative version of RAS, hence in general with a slow convergence.

## From NRAS to RASPEN

- Start with the nonlinear fixed point iteration

$$\mathbf{u}^{n+1} = \sum_j \mathbf{R}_j^T \mathbf{D}_j \mathbf{G}_j(\mathbf{u}^n)$$

which solves the nonlinear system


$$\mathcal{F}(\mathbf{u}) := \sum_j \mathbf{R}_j^T \mathbf{D}_j \mathbf{G}_j(\mathbf{u}) - \mathbf{u} = 0.$$

- Accelerate via Newton (the equivalent of GMRES in the nonlinear world)  $\rightsquigarrow$  the RASPEN method.

## Main features

- **Acceleration** of convergent fixed point iteration;
- Computation of **exact Jacobian**  $\nabla \mathcal{F}$ , or specifically the matrix-vector product  $\nabla \mathcal{F} \mathbf{v}$  for some  $\mathbf{v}$ .

$$\begin{aligned} \nabla \mathcal{F}(\mathbf{u}^n) &= \nabla(\mathbf{u}^n - \sum_j \mathbf{R}_j^T \mathbf{D}_j \mathbf{G}_j(\mathbf{u}^n)) \\ &= \sum_j \mathbf{R}_j^T \mathbf{D}_j [\mathbf{R}_j \nabla \mathbf{F}(\mathbf{u}^n) \mathbf{R}_j^T]^{-1} \mathbf{R}_j \nabla \mathbf{F}(\mathbf{u}^n) \end{aligned}$$


 V. Dolean, M. J. Gander, W. Kheriji, F. Kwok, and R. Masson, Nonlinear preconditioning: How to use a nonlinear schwarz method to precondition newton's method, SISC (2016).

The algorithm, for each time step, is given by:

## One-level algorithm

Given initial  $\mathbf{u}^0$ , for outer iteration  $n = 0, \dots$ , do until convergence

- Solve local subproblems  $R_j F(R_j^T G_j(\mathbf{u}^n) + (I - R_j^T R_j)\mathbf{u}^n) = 0$  for  $G_j(\mathbf{u}^n)$  (Newton);
- Glue local solutions  $\hat{\mathbf{u}}^n = \sum_j R_j^T D_j G_j(\mathbf{u}^n)$ ;
- Set  $\mathcal{F}(\mathbf{u}) = \mathbf{u} - \hat{\mathbf{u}}^n$ ;
- Solve  $\nabla \mathcal{F}(\mathbf{u}^n) \Delta^n = \mathcal{F}(\mathbf{u}^n)$ .
- Update  $\mathbf{u}^{n+1} = \mathbf{u}^n - \Delta^n$ .

 V. Dolean, M. J. Gander, W. Kheriji, F. Kwok, and R. Masson, Nonlinear preconditioning: How to use a nonlinear schwarz method to precondition newton's method, SISC (2016).

## Advantages of coarse spaces

- Allow for global communication between all subdomains.
- Are necessary for scalability for large number of subdomains for preconditioning of linear/nonlinear problems.

## Aim

- Robustness with respect to perforation size/location (even along subdomain interfaces);
- Robustness with respect to the number of subdomains  $N$ .

Two options:

- Coarse space based on the FAS like in Dolean et al. (2016)
- Here, we choose the coarse space specially tailored to perforated domains.



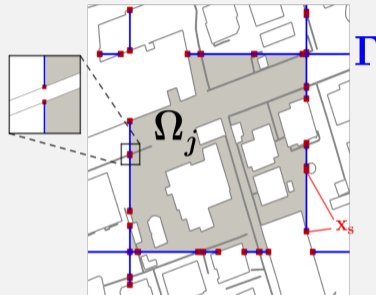
V. Dolean, M. J. Gander, W. Kheriji, F. Kwok, and R. Masson, *Nonlinear preconditioning: How to use a nonlinear Schwarz method to precondition newton's method*, SISC (2016).

## The construction of the coarse space

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## Coarse grid nodes for coarse space basis functions

- Coarse grid nodes arise at the intersection of nonoverlapping skeleton with a perforation boundary;
- $(\phi_s)_{s \in \{1, \dots, N_x\}}$  : Locally harmonic basis functions for each coarse grid node.
- # of coarse grid nodes is automatically generated.
- Continuously, the coarse space is given by  $V_H = \text{span}\{\phi_s\}$ .
- Think of as 'enriching' Multi-scale FEM (MsFEM) coarse space.

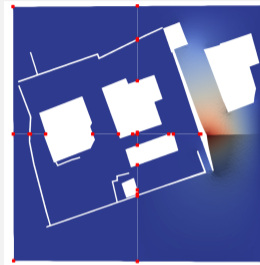


M. Boutilier, K. Brenner, and V. Dolean, Robust methods for multiscale coarse approximations of diffusion models in perforated domains, APNUM, 2024.

## Basis functions: Harmonic local solutions

For all nonoverlapping  $(\Omega'_j)_{j \in \{1, \dots, N\}}$  and  $s = 1, \dots, N_x$ , to obtain  $\phi_{s,j} = \phi_s|_{\Omega'_j}$ , solve

$$\begin{cases} \Delta \phi_{s,j} = 0 & \text{in } \Omega'_j, \\ \frac{\partial \phi_{s,j}}{\partial n} = 0 & \text{on } \partial \Omega'_j \cap \partial \Omega_s, \\ \phi_{s,j} = g_s & \text{on } \partial \Omega'_j \setminus \partial \Omega_s. \end{cases}$$



$g_s : \Gamma \rightarrow [0, 1]$  as: for  $i = 1, \dots, N_x$ ,

$$g_s(\mathbf{x}_i) = \begin{cases} 1, & s = i, \\ 0, & s \neq i, \end{cases}$$

- $g_s$  is linearly extended on the remainder of  $\Gamma$ .
- Can also include higher-order polynomials on coarse edges.

 M. Boutilier, K. Brenner, and V. Dolean, Robust methods for multiscale coarse approximations of diffusion models in perforated domains, APNUM, 2024.



## Two-level RASPEN with "multiscale" coarse space

We add the coarse correction multiplicatively and discretely.

### Two-level algorithm

Given initial  $\mathbf{u}^0$ , for outer iteration  $n = 0, \dots$ , do until convergence

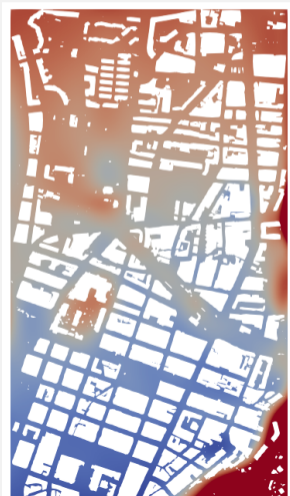
- Solve local subproblems  $R_j F(R_j^T G_j(\mathbf{u}^n) + (I - R_j^T R_j)\mathbf{u}^n) = 0$  for  $G_j(\mathbf{u}^n)$  (Newton);
- Glue local solutions  $\hat{\mathbf{u}}^n = \sum_j R_j^T D_j G_j(\mathbf{u}^n)$ ;
- Set  $\mathcal{F}(\mathbf{u}) = \mathbf{u} - \hat{\mathbf{u}}^n$ ;
- Solve coarse problem  $R_0 F(\hat{\mathbf{u}}^n - R_0^T \mathbf{c}_0^n) = 0$  for  $\mathbf{c}_0^n$ ;
- Set  $\mathcal{F}(\mathbf{u}^n) = \mathbf{u}^n - \hat{\mathbf{u}}^n + R_0^T \mathbf{c}_0^n$ ;
- Solve  $\nabla \mathcal{F}(\mathbf{u}^n) \Delta^n = \mathcal{F}(\mathbf{u}^n)$ .
- Update  $\mathbf{u}^{n+1} = \mathbf{u}^n - \Delta^n$ .

## Numerical Results

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## Setup example model problem

- Excessive water flow coming from Paillon river in Nice, France;
- Dirichlet boundary conditions with initial condition  $u_0 > z_b$  at leftmost boundary (river).
- $\alpha = \frac{3}{2}, \gamma = 1, 0$  source term,  $c_f = 30$ .



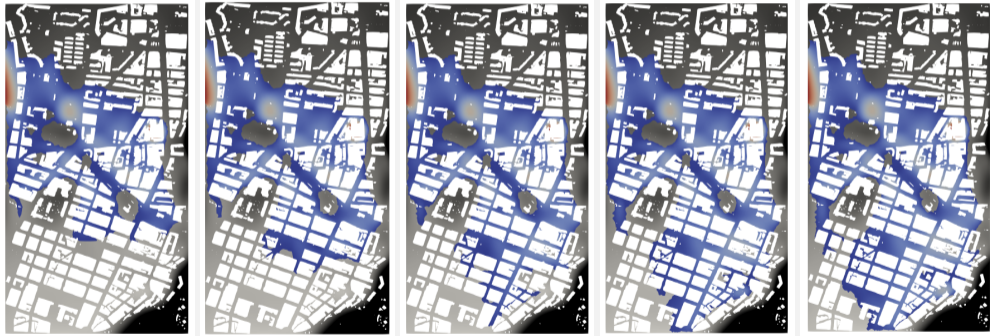
## Solution over time

- $z_b$ : black and white (darker = higher elevation)
- $h$  (water depth): colour



## Solution over time

- $z_b$ : black and white (darker = higher elevation)
- $h$  (water depth): colour



We can also perform typical linear DD methods on the linearised system at each Newton iteration (SNK: Schwarz-Newton-Krylov)

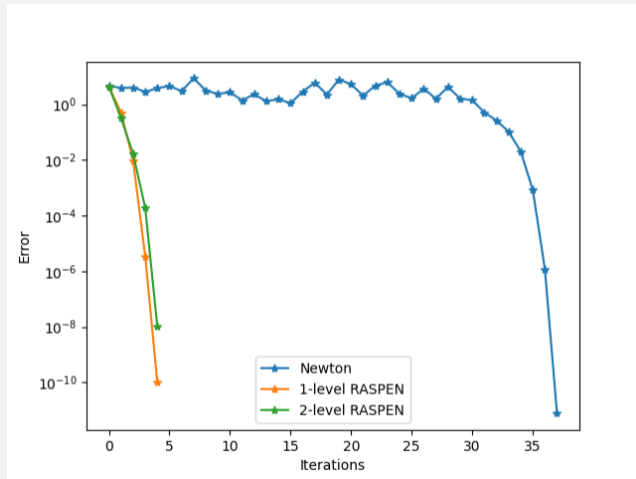
$$M_{\text{RAS},2}^{-1} J_n \mathbf{u} = M_{\text{RAS}2}^{-1} \mathbf{F}_n.$$

where

$$M_{\text{RAS},2}^{-1} = \sum_{j=1}^N R_j^T D_j (R_j J_n R_j^T)^{-1} R_j + R_0^T (R_0 J_n R_0^T)^{-1} R_0.$$

**Question:** Will our coarse space (designed for Poisson equation) work well for the linearized Newton system?  $\rightsquigarrow$  Perform scalability tests.

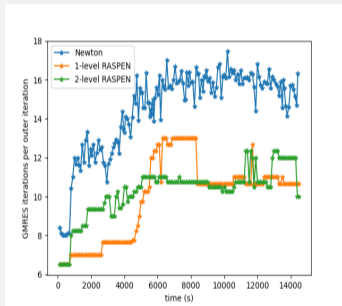
## Numerical Results: First time step



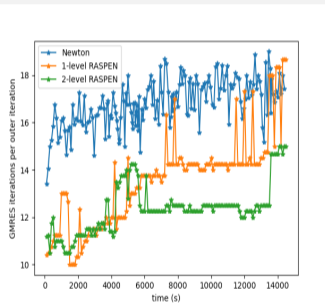
- 1 and 2 level RASPEN immediately enter region of quadratic convergence, quite insensitive to initial guess.

# Numerical results: GMRES iterations

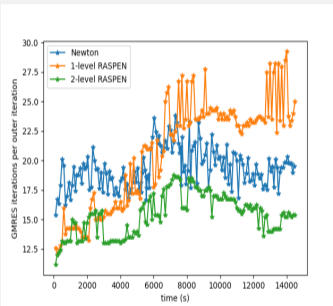
Average GMRES iterations per outer iteration for each time step.



$N = 2 \times 2$



$N = 4 \times 4$



$N = 8 \times 8$



## Numerical results: Outer iteration, GMRES iterations, and convergence curves

$N = 2 \times 2$	T. steps	Outer Its	GMRES Its
Newton	160	2216	19497
1-level RASPEN	146	440	2630
2-level RASPEN	146	532	3187(+1300 coarse)

$N = 4 \times 4$	T. steps	Outer Its	GMRES Its
Newton	160	2159	23989
1-level RASPEN	146	554	5026
2-level RASPEN	146	576	4507(+1719 coarse)

$N = 8 \times 8$	T. steps	Outer Its	GMRES Its
Newton	160	2266	30811
1-level RASPEN	146	602	9668
2-level RASPEN	146	609	5627(+2255 coarse)

- 1-level RASPEN: GMRES iterations do not scale with  $N$ .

- RASPEN is a very good alternative for our model problem, improving the convergence of Newton's method;
- Local time step reduction can be employed for problem subdomains to avoid a global time step reduction;
- For a larger number of subdomains, a coarse correction is necessary for scalability → our coarse space designed for the Poisson equation works well in this nonlinear case;
- As an alternative, the two-level RAS preconditioner with our coarse space provides a scalable number of Krylov iterations.



M. Boutilier, K. Brenner, and V. Dolean, Two-level Nonlinear Preconditioning Methods for Flood Models Posed on Perforated Domains, arXiv:2406.06189, 2024.

## Funding Acknowledgement

This work has been supported by ANR Project Top-up (ANR-20-CE46-0005).

We also thank Métropole Nice Côte d'Azur for the given data.



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Thanks for your attention!