

Iterative solution of spatial network models by subspace decomposition

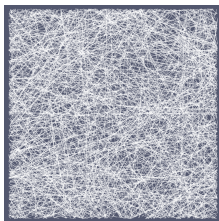
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Simulation of spatial network models at FCC



- We consider

$$Ku = f$$

- a simplified network model of an elliptic PDE (K is SPD)
- K is ill-conditioned (geometry and material data variation), only direct solver works
- The goal is to develop an iterative solver¹

¹Görtz-Hellman-M., Iterative solution of spatial network models by subspace decomposition, arXiv:2207.07488

- 1 **Graph Laplacian and model problem**
- 2 Network assumptions
- 3 Convergence of a semi-iterative solver
- 4 Numerical examples
- 5 Conclusions

Graph Laplacian

- Let $\mathcal{G} = (\mathcal{N}, \mathcal{E})$ be a graph of nodes and edges, $x \in \Omega \subset \mathbb{R}^d$
- The graph Laplacian L^g is SP(semi-)D, $L^g \mathbf{1} = 0$
- Let $\hat{V} : \mathcal{N} \rightarrow \mathbb{R}$ be scalar functions on \mathcal{N} . For $v, w \in \hat{V}$

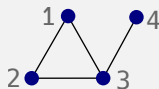
$$(v, w) = \sum_x v(x)w(x)$$

$$(L^g v, v) = \sum_{(x,y) \in \mathcal{E}} (v(x) - v(y))^2$$

$$L^g = \sum_x L_x^g$$

$$(L_x^g v, v) = \frac{1}{2} \sum_{y \sim x} (v(x) - v(y))^2$$

Example:



$$L^g = \begin{pmatrix} 2 & -1 & -1 & 0 \\ -1 & 2 & -1 & 0 \\ -1 & -1 & 3 & -1 \\ 0 & 0 & -1 & 1 \end{pmatrix}$$

$x \sim y$ denotes that x and y are connected by an edge

Weighted graph Laplacian

- A weighted graph Laplacian and diagonal mass matrix

$$(L_x v, v) = \frac{1}{2} \sum_{y \sim x} \frac{(v(x) - v(y))^2}{|x - y|}, \quad L = \sum L_x$$

$$(M_x v, v) := \frac{1}{2} v(x)^2 \sum_{y \sim x} |x - y|, \quad M = \sum M_x$$

- Consider the 1D mesh $0 = x_0 < x_1 < \dots < x_n = 1$.

$$(Lv, v) := \sum_{i=1}^n \frac{(v(x_i) - v(x_{i-1}))^2}{|x_i - x_{i-1}|}$$

- L is the P1-FEM stiffness matrix $(-\Delta)$ and M is the lumped mass matrix

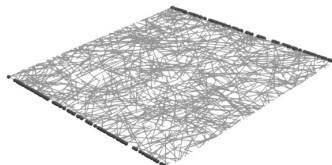
Model problem

Find $u \in V := \{v \in \hat{V} : v(x) = 0 \text{ for } x \in \Gamma_D\}$:

$$(Ku, v) = (f, v), \quad v \in V.$$

Assume: $(K\cdot, \cdot)$ is scalar product on V and

$$\alpha(Lv, v) \leq (Kv, v) \leq \beta(Lv, v), \quad \forall v \in V.$$



Example:

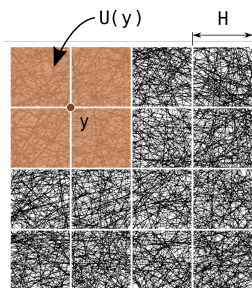


$$(Kv, v) = \sum_{(x,y) \in \mathcal{E}} \gamma_{xy} \frac{(v(x) - v(y))^2}{|x - y|}, \quad \alpha \leq \gamma_{xy} \leq \beta$$

- P1-FEM for 1D diffusion-reaction model on network with continuity and Kirchhoff flux constraint in junctions
- Structural model of a fibre network.

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Multilevel solver: coarse scale representation



- \mathcal{T}_H is a mesh of squares
- \hat{V}_H is Q1-FEM with basis $\{\varphi_y\}_y$
- $V_H \subset \hat{V}_H$ satisfy the boundary conditions
- Clément type interpolation operator

$$\mathcal{I}_{HV} = \sum_{\text{free DoFs } y} \frac{(M_{U(y)} \mathbf{1}, v)}{(M_{U(y)} \mathbf{1}, \mathbf{1})} \varphi_y \in V_H$$

Lemma (Stability and approximability of \mathcal{I}_H)

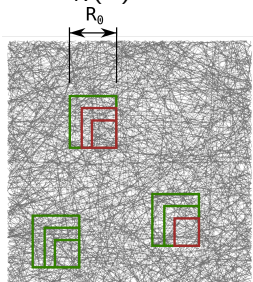
For all $v \in V$ and for $H > R_0$,

$$H^{-1} |v - \mathcal{I}_{HV}|_M + |\mathcal{I}_{HV}|_L \leq C |v|_L,$$

where $|\cdot|_M^2 = (M \cdot, \cdot)$, $|\cdot|_L^2 = (L \cdot, \cdot)$ and $C = C_d \mu \sqrt{\sigma}$.

Network locality, homogeneity and connectivity

- 1 All edges are shorter than $R_0 > 0$ (length scale)
- 2 Let $B_R(x)$ be a box at x of side length $2R$, with $H \geq R_0$,



$$1 \leq \frac{\max_x |1|_{M, B_H(x)}^2}{\min_x |1|_{M, B_H(x)}^2} \leq \sigma(R_0)$$

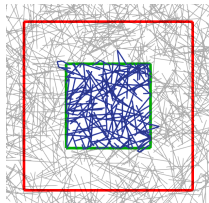
- 3 For all $x \in \Omega$ and $H > R_0$ there is a (uniform) $\mu(R_0) < \infty$ and $c_x \in \mathbb{R}$, such that the Poincaré-type inequality holds

$$|v - c_x|_{M, B_H(x)} \leq \mu H |v|_{L, B_{H+R_0}(x)}, \quad \forall v \in \hat{V}$$

Poincaré constant

Let $\mathcal{G}' = (\mathcal{N}', \mathcal{E}') \subset \mathcal{G}$ be connected and

- all nodes in $B_H(x)$ are included
- no nodes outside $B_{H+R_0}(x)$ are included



With L', M' defined on \mathcal{G}' we have

$$\lambda_1 = \inf \frac{(L'z, z)}{(M'z, z)} = \frac{(L'1, 1)}{(M'1, 1)} = 0, \quad \lambda_2 = \inf_{(M'1, z)=0} \frac{(L'z, z)}{(M'z, z)} > 0.$$

With $c_x = \frac{(M'1, v)}{(M'1, 1)}$ we have $(M'1, v - c_x) = 0$ so

$$|v - c_x|_{M, B_H(x)} \leq |v - c_x|_{M'} \leq \lambda_2^{-1/2} |v - c_x|_{L'} \leq \lambda_2^{-1/2} |v|_{L, B_{H+R_0}(x)}$$

λ_2 : measure connectivity² $\sim CH^{-2}$ if isoperimetric³ dim d .

²Cheeger 1970, Fiedler 1973

³F. Chung, Spectral graph theory, AMS, 1997

Example: Connectivity $\lambda_2^{-1/2} \approx \mu R$

Finite length fibers $r = 0.05$ and $|1|_M^2 = 1000$, $\Omega = [0, 1]^2$

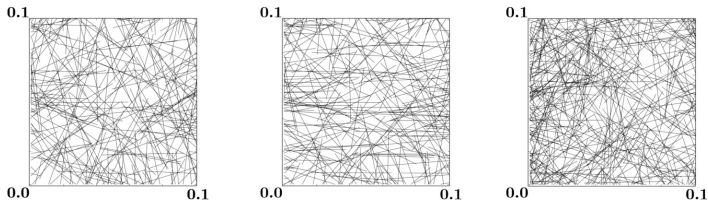
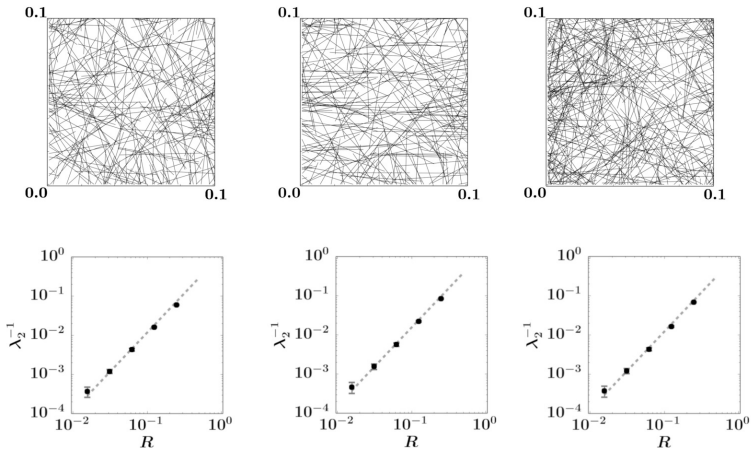


Table: (σ, μ) for different R

$R^{-1} = 4$	$R^{-1} = 8$	$R^{-1} = 16$	$R^{-1} = 32$	$R^{-1} = 64$
(1.04, 0.49)	(1.08, 0.53)	(1.27, 0.57)	(1.85, 0.675)	(3.42, 1.53)
(1.04, 0.59)	(1.08, 0.61)	(1.27, 0.69)	(1.87, 0.83)	(2.93, 1.35)
(1.04, 0.53)	(1.57, 0.54)	(2.13, 0.58)	(3.1, 0.76)	(6.86, 1.45)

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Finite length fibers $r = 0.05$ and $|1|_M^2 = 1000$, $\Omega = [0, 1]^2$



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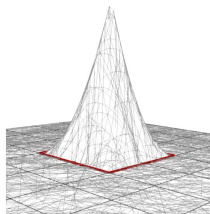
Subspace decomposition preconditioner⁴

Let $V_0 = V_H$. and

$$V_j = V(U(y_j)), \quad j = 1, \dots, m.$$

Define projections $P_j : V \rightarrow V_j$ by

$$(KP_j v, v_j) = (Kv, v_j), \quad \forall v_j \in V_j \subset V.$$



We add the projections to form

$$P = P_0 + P_1 + \dots + P_m.$$

- $P = BK$ is used as a preconditioner: $BKu = Bf$.
- We use the preconditioned conjugate gradient method.
- Involves direct solution of decoupled problems (**semi-iterative**).

⁴Kornhuber & Yserentant, MMS, 2016

Convergence analysis

Lemma (Spectral bound of P)

For $H > 2R_0$ it holds

$$C_1^{-1}|v|_K^2 \leq (K Pv, v) \leq C_2|v|_K^2, \quad \forall v \in V,$$

where $C_1 = C_d \beta \alpha^{-1} \sigma \mu^2$ and $C_2 = C_d \beta \alpha^{-1}$.

Interpolation bound is a crucial component of the proof.

Theorem (Convergence of PCG)

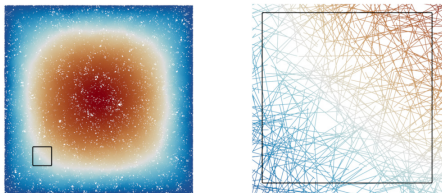
With $\sqrt{k} = \sqrt{C_1 C_2} = C_d \beta \alpha^{-1} \mu \sqrt{\sigma}$ and $H > 2R_0$ it holds

$$|u - u^{(\ell)}|_K \leq 2 \left(\frac{\sqrt{k} - 1}{\sqrt{k} + 1} \right)^\ell |u - u^{(0)}|_K.$$

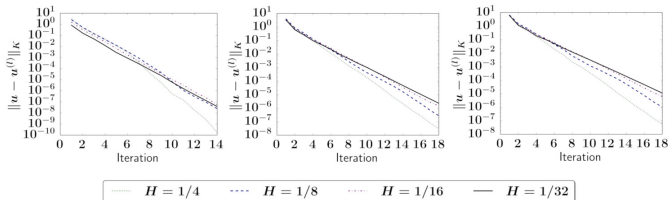
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Example: Convergence graph Laplacian

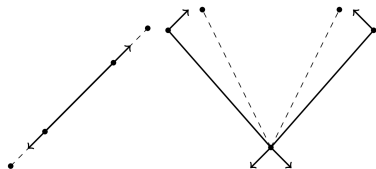
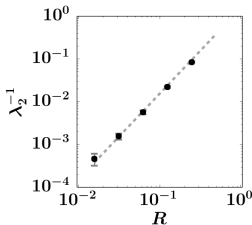
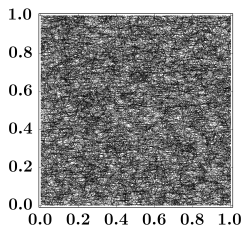
Consider $Ku = M1$ with homogeneous Dirichlet bc $|1|_M^2 = 1000$.



Grid $\gamma = 1$ (left), rand $\gamma = 1$ (center), rand $\gamma \in U([0.1, 1])$ (right)



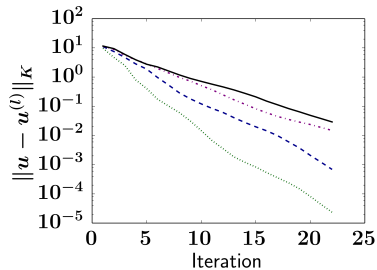
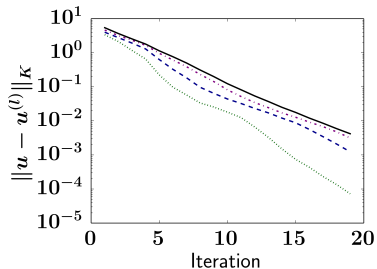
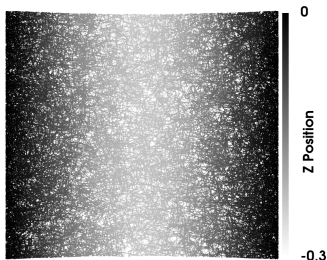
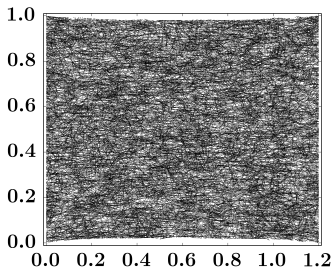
Example: A fibre network model⁵



- $2 \cdot 10^4$ fibres, biased angle (x-axis), length 0.05, $3 \cdot 10^5$ nodes, $\alpha = 0.05$, $\beta = 500$.
- Two forces in the model: edge extension and angular deviation.
- Find displacement u : $Ku = f$ (tensile, distributed load)
- Theory extends to vector valued setting (Korn, $K \sim L$)
- DD with $H = 1/4, 1/8, 1/16, 1/32$.

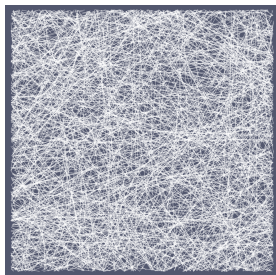
⁵Kettil et. al. *Numerical upscaling of discrete network models*, BIT 2020

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Conclusions



- The network should resemble a homogeneous material on coarse scales $H > H_0$
- Direct solver on fine scales (localized, in parallel)
- Poincaré type inequality plays a crucial role in the analysis
- Numerical PDE methods/analysis can be applied to spatial network models